Chaotic Attractors of Control Systems

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Abstract. The paper is dedicated to the study of the problem of existence of chaotic attractors of discrete control systems and to description of its structure. We consider a family of continuous mappings of metric space \(W\) into itself, and \((W, f_i)_{i \in I}\) is the family of discrete dynamical systems. On the metric space \(W\) we consider a discrete inclusion

\[ u_{t+1} \in F(u_t) \]

associated by \(M := \{f_i : i \in I\}\), where \(F(u) = \{f(u) : f \in M\}\) for all \(u \in W\). If the family \(M\) consists of a finite number of maps, then the corresponding attractor is chaotic. We study this problem in the framework of non-autonomous dynamical systems (cocyles).

Keywords: Chaotic attractor; set-valued dynamical system; control system, collage, cocycle.

1 Introduction

The aim of this paper is the study of the problem of existence of compact global chaotic attractors of discrete control systems (see, for example, Cheban [4,5], Bobylev, Emel’yanov and Korovin [2] and the references therein). Let \(W\) be a metric space, \(M := \{f_i : i \in I\}\) be a family of continuous mappings of \(W\) into itself and \((W, f_i)_{i \in I}\) be the family of discrete dynamical systems, where \((W, f)\) is a discrete dynamical system generated by positive powers of continuous map \(f : W \to W\). On the space \(W\) we consider a discrete inclusion

\[ u_{t+1} \in F(u_t) \]

for some \(f_i, f_{j-1} \in M\) (trajectory of \(DI(M)\)).

We can consider that it is a discrete control problem, where at each moment of the time \(j\) we can apply a control from the set \(M\), and \(DI(M)\) is the set of possible trajectories of the system.
The problem of existence of compact global attractors for a discrete inclusion arise in a number of different areas of mathematics (see, for example, [5,6] and the references therein).

A sequence $\omega : Z_+ \mapsto \{1, 2, \ldots, m\}$ is called $p \in N := \{1, 2, \ldots\}$ periodic, if $\omega(n + p) = \omega(n)$ for all $n \in Z_+$. A point $x_0 \in W$ is said to be $p$-periodic for $DI(M)$ if there exists a $p$-periodic sequence $\omega : Z_+ \mapsto \{1, 2, \ldots, m\}$ such that solution $\{x(k)\}_{k \in Z_+}$ of equation (1) ($\omega(i) = i_j$ for all $i \in Z$ and $i_j \in \{1, 2, \ldots, m\}$) with initial data $x(0) = x_0$ is $mp$-periodic, i.e., $x(k + p) = x(k)$ for all $k \in Z_+$.

It is well known the following result.

**Theorem 1.** [1, Ch.II,IV] Let $M = \{f_1, f_2, \ldots, f_m\}$ be a finite family of continuous mappings from $W$ into itself. If there exists a number $q \in (0, 1)$ such that $\rho(f_1(x_1), f_1(x_2)) \leq q \rho(x_1, x_2)$ for all $x_1, x_2 \in W$ and $i \in \{1, 2, \ldots, m\}$, then the following statement hold:

1. $DI(M)$ admits a compact global attractor $I$, i.e.,
   
   (a) $I$ is a nonempty, compact and invariant set $F(I) = I$, where $F(x) := \{f_1(x), f_2(x), \ldots, f_m(x)\}$ for all $x \in W$ and $F(I) := \bigcup \{F(x) : x \in I\}$;
   
   (b) $\lim_{n \to \infty} \beta(F^n(x), I) = 0$ for all $x \in W$ uniformly with respect to $x$ on every compact subset $M$ from $W$, where $\beta(A, B) := \sup_{a \in B} \rho(a, B)$

   $\bigcup \{A, B \subseteq W\}$.

2. $I$ coincides with the closure of the all periodic points of $DI(M)$.

In the book [5] (Chapter VI) it was generalized this theorem for the finite family $M = \{f_1, f_2, \ldots, f_m\}$ when it is contracting in the generalized sense, i.e., there are two positive numbers $\mathcal{N}$ and $q \in (0, 1)$ such that

$$\rho(f_{i_n}f_{i_{n-1}} \cdots f_{i_1}(x_1), f_{i_n}f_{i_{n-1}} \cdots f_{i_1}(x_2)) \leq \mathcal{N}q^n \rho(x_1, x_2)$$

for all $x_1, x_2 \in W$ and $n \in N$, where $i_1, i_2, \ldots, i_n \in \{1, 2, \ldots, m\}$.

## 2 Some Notions and Facts from Dynamical and Control Systems

In this paper we will use some notions and facts from the theory of dynamical and control systems [5]. In this Section we collect some of them.

### 2.1 Set-valued dynamical systems and their compact global attractors

Let $(X, \rho)$ be a complete metric space, $S$ be a group of real ($R$) or integer ($Z$) numbers, $T$ ($S_+ \subseteq T$) be a semi-group of additive group $S$. By $C(X)$ we denote the family of all non-empty compact subsets of $X$. For every point $x \in X$ and number $t \in T$ we put in correspondence a closed compact subset $\pi(t, x) \in C(X)$. So, if $\pi(P, A) = \bigcup \{\pi(t, x) : t \in P, x \in A\}(P \subseteq T)$, then
1. \( \pi(0, x) = x \) for all \( x \in X \);  
2. \( \pi(t_2, \pi(t_1, x)) = \pi(t_1 + t_2, x) \) for all \( x \in X \);  
3. \( \lim_{x \to x_0, t \to t_0} \beta(\pi(t, x), \pi(t_0, x_0)) = 0 \) for all \( x_0 \in X \) and \( t_0 \in T \), where \( \beta(A, B) = \sup \{ \rho(a, B) : a \in A \} \) is a semi-deviation of the set \( A \subseteq X \) from the set \( B \subseteq X \).

In this case it is said \([11]\) that there is defined a set-valued semi-group dynamical system.

Let \( T \subset T' \subset S \). A continuous mapping \( \gamma_x : T' \to X \) is called a motion of the set-valued dynamical system \((X, T, \pi)\) issuing from the point \( x \in X \) at the initial moment \( t = 0 \) and defined on \( T' \), if  

a. \( \gamma_x(0) = x \);  
b. \( \gamma_x(t_2) \in \pi(t_2 - t_1, \gamma_x(t_1)) \) for all \( t_1, t_2 \in T' \) \((t_2 > t_1)\).

The set of all motions of \((X, T, \pi)\), passing through the point \( x \) at the initial moment \( t = 0 \) is denoted by \( \mathcal{F}_x(\pi) \) and \( \mathcal{F}(\pi) := \bigcup \{ \mathcal{F}_x(\pi) \mid x \in X \} \) (or simply \( \mathcal{F} \)).

The trajectory \( \gamma \in \mathcal{F}(\pi) \) defined on \( S \) is called a full (entire) trajectory of the dynamical system \((X, T, \pi)\).

Denote by \( \Phi(\pi) \) the set of all full trajectories of the dynamical system \((X, T, \pi)\) and \( \Phi_x(\pi) := \mathcal{F}_x(\pi) \bigcap \Phi(\pi) \).

A system \((X, T, \pi)\) is called \([3],[5]\) compactly dissipative, if there exists a nonempty compact \( K \subseteq X \) such that

\[
\lim_{t \to +\infty} \beta(\pi^t M, K) = 0;
\]

for all \( M \in C(X) \).

Let \((X, T, \pi)\) be compactly dissipative and \( K \) be a compact set attracting every compact subset of \( X \). Let us set

\[
J := \omega(K) := \bigcap_{t \geq 0} \bigcup_{\tau \geq t} \pi^\tau K. \tag{3}
\]

It can be shown \([3],[5]\) that the set \( J \) defined by equality (3) does not depend on the choice of the attractor \( K \), but it is characterized only by the properties of the dynamical system \((X, T, \pi)\) itself. The set \( J \) is called Levinson center of the compact dissipative dynamical system \((X, T, \pi)\).

### 2.2 Discrete inclusions, ensemble of dynamical systems (collages) and cocycles

Let \( W \) be a topological space. Denote by \( C(W, W) \) the space of all continuous operators \( f : W \to W \) equipped with the compact-open topology. Consider a set of operators \( \mathcal{M} \subseteq C(W, W) \) and, respectively, an ensemble (collage) of
discrete dynamical systems \((W, f)\) (\((W, f)\) is a discrete dynamical system generated by positive powers of map \(f\)).

A discrete inclusion \(DI(M)\) is called (see, for example, [2,7]) a set of all sequences \(\{x_j\} \subset W\) \((j \in \mathbb{Z}_+)\) such that

\[
x_j = f_i x_{j-1}
\]  

(4)

for some \(f_i \in M\) (trajectory of \(DI(M)\)).

A bilateral sequence \(\{x_j\} \subset W\) \((j \in \mathbb{Z})\) is called a full trajectory of \(DI(M)\) (entire trajectory or trajectory on \(Z\)), if \(x_{n+j} = f_i x_{n+j-1}\) for all \(n \in \mathbb{Z}\).

Let us consider the set-valued function \(F : W \to C(W)\) defined by the equality \(F(x) := \{f(x) | f \in M\}\). Then the discrete inclusion \(DI(M)\) is equivalent to the difference inclusion

\[
x_j \in F(x_{j-1}).
\]  

(5)

Denote by \(F_{x_0}\) the set of all trajectories of discrete inclusion \((5)\) or \((DI(M))\) issuing from the point \(x_0 \in W\) and \(\mathcal{F} := \bigcup \{F_{x_0} \mid x_0 \in W\}\).

Below we will give a new approach concerning the study of discrete inclusions \(DI(M)\) (or difference inclusion \((5)\)). Denote by \(C(Z_+, W)\) the space of all continuous mappings \(f : Z_+ \to W\) equipped with the compact-open topology. Denote by \((C(Z_+, X), Z_+, \sigma)\) a dynamical system of translations (shifts dynamical system or dynamical system of Bebutov [9,10]) on \(C(Z_+, W)\), i.e. \(\sigma(k, f) := f_k\) and \(f_k\) is a \(k \in Z_+\) shift of \(f\) (i.e., \(f_k(n) := f(n + k)\) for all \(n \in Z_+)\).

We may now rewrite equation \((4)\) in the following way:

\[
x_{j+1} = \omega(j)x_j, \ (\omega \in \Omega := C(Z_+, M))
\]  

(6)

where \(\omega \in \Omega\) is the operator-function defined by the equality \(\omega(j) := f_{j+1}\) for all \(j \in Z_+\). We denote by \(\varphi(n, x_0, \omega)\) the solution of equation \((6)\) issuing from the point \(x_0 \in E\) at the initial moment \(n = 0\). Note that \(F_{x_0} = \{\varphi(\cdot, x_0, \omega) | \omega \in \Omega\}\) and \(\mathcal{F} = \{\varphi(\cdot, x_0, \omega) | x_0 \in W, \omega \in \Omega\}\), i.e., \(DI(M)\) (or inclusion \((5)\)) is equivalent to the family of non-autonomous equations \((6)\) \((\omega \in \Omega)\).

From the general properties of difference equations it follows that the mapping \(\varphi : Z_+ \times W \times \Omega \to W\) satisfies the following conditions:
1. \(\varphi(0, x_0, \omega) = x_0\) for all \((x_0, \omega) \in W \times \Omega\);
2. \(\varphi(n + \tau, x_0, \omega) = \varphi(n, \varphi(\tau, x_0, \omega), \sigma(\tau, \omega))\) for all \(n, \tau \in Z_+\) and \((x_0, \omega) \in W \times \Omega\);
3. the mapping \(\varphi\) is continuous;
4. for any \(n, \tau \in Z_+\) and \(\omega_1, \omega_2 \in \Omega\) there exists \(\omega_3 \in \Omega\) such that

\[
U(n, \omega_2)U(\tau, \omega_1) = U(n + \tau, \omega_3),
\]  

(7)

where \(\omega \in \Omega, U(n, \omega) := \varphi(n, \cdot, \omega) = \prod_{k=0}^{n} \omega(k), \ \omega(k) := f_{i_k} \ (k = 0, 1, \ldots, n)\) and \(f_{i_0} := Id_W\).
Let \( T_1 \subseteq T_2 \) be two sub-semigroups of group \( S \), \( X, Y \) be two metric (or topological) spaces and \((X, T_1, \pi) \) (respectively \((Y, T_2, \sigma) \)) be a semigroup dynamical system on \( X \) (respectively on \( Y \)). A triplet \((X, T_1, \pi), (Y, T_2, \sigma), h)\) is called a non-autonomous dynamical system, where \( h : X \to Y \) is a homomorphism from \((X, T_1, \pi)\) onto \((Y, T_2, \sigma)\), i.e., \( h(\pi(t, x)) = \sigma(t, h(x))\) for all \( x \in X \) and \( t \in T_1 \).

Let \( W, \Omega \) be two topological spaces and \((\Omega, T_2, \sigma)\) be a semi-group dynamical system on \( \Omega \).

Recall [9] that a triplet \((W, \varphi, (\Omega, T_2, \sigma))\) (or briefly \( \varphi \)) is called a cocycle over \((\Omega, T_2, \sigma)\) with the fiber \( W \), if \( \varphi \) is a mapping from \( T_1 \times W \times \Omega \) to \( W \) satisfying the following conditions:

1. \( \varphi(0, x, \omega) = x \) for all \( (x, \omega) \in W \times \Omega \);
2. \( \varphi(t + \tau, x, \omega) = \varphi(t, \varphi(\tau, x, \omega), \sigma(\tau, \omega)) \) for all \( t, \tau \in T_1 \) and \( (x, \omega) \in W \times \Omega \);
3. the mapping \( \varphi \) is continuous.

Let \( X := W \times \Omega \), and define the mapping \( \pi : X \times T_1 \to X \) by the equality:
\[
\pi((u, \omega), t) := (\varphi(t, u, \omega), \sigma(t, \omega)) \quad \text{(i.e., } \pi = (\varphi, \sigma))\] Then it is easy to check that \((X, T_1, \pi)\) is a dynamical system on \( X \), which is called a skew-product dynamical system [9]; but \( h = \text{pr}_2 : X \to \Omega \) is a homomorphism of \((X, T_1, \pi)\) onto \((\Omega, T_2, \sigma)\) and hence \((X, T_1, \pi), (\Omega, T_2, \sigma), h)\) is a non-autonomous dynamical system.

From the presented above it follows that every \( DI(\mathcal{M}) \) (respectively, inclusion (5)) in a natural way generates a cocycle \((W, \varphi, (\Omega, Z_+, \sigma))\), where \( \Omega = C(Z_+, \mathcal{M}), (\Omega, Z_+, \sigma) \) is a dynamical system of shifts on \( \Omega \) and \( \varphi(n, x, \omega) \) is the solution of equation (6) issuing from the point \( x \in W \) at the initial moment \( n = 0 \). Thus, we can study inclusion (5) (respectively, \( DI(\mathcal{M}) \)) in the framework of the theory of cocycles with discrete time.

Below we need the following result.

**Theorem 2.** [6] Let \( \mathcal{M} \) be a compact subset of \( C(W, W) \) and \((W, \varphi, (\Omega, Z_+, \sigma))\) be a cocycle generated by \( DI(\mathcal{M}) \). Then

1. \( \Omega = \text{Per}(\sigma), \) where \( \text{Per}(\sigma) \) is the set of all periodic points of \((\Omega, Z_+, \sigma)\) (i.e., \( \omega \in \text{Per}(\sigma) \) if there exists \( \tau \in N \) such that \( \sigma(\tau, \omega) = \omega \));
2. the set \( \Omega \) is compact;
3. \( \Omega \) is invariant, i.e., \( \sigma^t \Omega = \Omega \) for all \( t \in Z_+ \);
4. \( \varphi \) satisfies the condition (7).

## 3 Chaotic attractors of discrete control systems

In Section 3 we give the conditions of existence of chaotic attractor for discrete control systems.

Denote by \( \mathcal{A} \) the set of all mapping \( \psi : Z_+ \times R_+ \to R_+ \) possessing the following properties:
(G1) \( \psi \) is continuous;
(G2) there exists a positive number \( t_0 \) such that:
   (a) \( \psi(t_0, r) < r \) for all \( r > 0 \);
   (b) the mapping \( \psi(t_0, \cdot) : R_+ \rightarrow R_+ \) is monotone increasing.
(G3) \( \psi(t + \tau, r) \leq \psi(t, \psi(\tau, r)) \) for all \( t, \tau \in Z_+ \) and \( r \in R_+ \).

Remark 1. 1. Note that the functions \( \psi(t, r) = Nq^t r \) (\( N > 0 \) and \( q \in (0, 1) \)) and \( \psi(t, r) = \frac{r}{1 + rt} \) belong to \( A \), where \( (t, r) \in Z_+ \times R_+ \).

2. Let \( f : R_+ \) be a continuous function satisfying the conditions:
   1. \( f(r) < r \) for all \( r > 0 \);
   2. \( f \) is monotone increasing.
Then the mapping \( \psi : Z_+ \times R_+ \rightarrow R_+ \) defined by equality
\[
\psi(t, r) = x(t)
\]
for all \( (t, r) \in Z_+ \times R_+ \), where \( x(t) \) is a unique solution of difference equation
\[
x_{t+1} = f(x_t)
\]
with initial data \( x_0 = r \), belongs to \( A \).

Let \( \psi \in A \). A set \( M \) of operators from \( C(W, W) \) is said to be \( \psi \)-contracting, if
\[
\rho(f_{i_1} f_{i_{t-1}} \cdots f_{i_1} x_1, f_{i_1} f_{i_{t-1}} \cdots f_{i_1} x_2) \leq \psi(t, \rho(x_1, x_2))
\]
for all \( x_1, x_2 \in W \) and \( t \in N \), where \( f_{i_1}, f_{i_2}, \ldots, f_{i_t} \in C(W) \) and \( i_1, i_2, \ldots, i_t \in N \).

The set \( S \subset W \) is
1. nowhere dense, provided the interior of the closure of \( S \) is empty set, \( int(cl(S)) = \emptyset \);
2. totally disconnected, provided the connected components are single points;
3. perfect, provided it is closed and every point \( p \in S \) is the limit of points \( q_n \in S \) with \( q_n \neq p \).

The set \( S \subset W \) is called a Cantor set, provided it is totally disconnected, perfect and compact.

The subset \( M \) of \( (X, T, \pi) \) is called (see, for example, [8]) chaotic, if the following conditions hold:
1. the set \( M \) is transitive, i.e. there exists a point \( x_0 \in X \) such that \( M = H(x_0) := \{ \pi(t, x_0) : t \in T \} \);
2. \( M = \text{Per}(\pi) \), where \( \text{Per}(\pi) \) is the set of all periodic points of \( (X, T, \pi) \).

Recall that a point \( x \in X \) of the dynamical system \( (X, T, \pi) \) is called Poisson’s stable, if \( x \) belongs to its \( \omega \)-limit set \( \omega_x := \bigcap_{t \geq 0} \bigcup_{\tau \geq t} \pi(\tau, x) \).

Theorem 3. Suppose that the following conditions are fulfilled:
a. $\mathcal{M}$ is a finite subset of $C(W, W)$, i.e., $\mathcal{M} := \{f_1, f_2, \ldots, f_m\}$ and $m \geq 2$;
b. $\mathcal{M}$ is $\psi$-contracting for some $\psi \in \mathcal{A}$.

Then the following statement hold:

1. the cocycle $(W, \varphi, (\Omega, Z_+, \sigma))$ ($\Omega := C(Z_+, \mathcal{M})$) generated by $DI(\mathcal{M})$ is compactly dissipative;
2. the skew-product dynamical system $(X, Z_+, \pi)$ generated by $DI(\mathcal{M})$ is compactly dissipative;
3. $I = \text{Per} (\varphi)$, where $\text{Per} (\varphi) := \{u \in W \colon \exists \tau \in \mathbb{N} \text{ and } \omega \in \Omega \text{ such that }\ 
\sigma (\tau, \omega) = \omega \text{ and } \varphi (\tau, u, \omega) = u\}$;
4. if every map $f \in \mathcal{M}$ is invertible, then
   1. Levinson’s center $J$ of the skew-product dynamical system $(X, Z_+, \pi)$ is a chaotic Cantor set;
   2. there exists a residual subset $J_0 \subseteq J$ (large in the sense of Baire category), consisting from Poisson’s stable points, such that the positive semi-trajectory of every point $x_0 \in J_0$ is dense on $J$;
   3. $I = \text{pr}_1 (J)$ (pr$_1 : X \to \Omega$, where $I$ is the Levinson’s center of cocycle $\varphi$ and $X := W \times \Omega$), i.e., $I$ is a continuous image of the Cantor set $J$.

Proof. Let $Y = \Omega := C(Z_+, \mathcal{Q})$ and $(Y, Z_+, \sigma)$ be a semi-group dynamical system of shifts on $Y$. Then $Y$ is compact. By Theorem 2, $\overline{\text{Per} (\sigma)} = \Omega$ and $\Omega$ is compact and invariant.

Let $(W, \varphi, (\Omega, Z_+, \sigma))$ be a cocycle generated by $DI(\mathcal{M})$ (i.e., $\varphi (n, u, \omega) := U(n, \omega)u$, where $U(n, \omega) = \prod_{k=0}^{n} \omega (k)$ ($\omega \in \Omega$)), $(X, Z_+, \pi)$ be a skew-product system associated by the cocycle $\varphi$ (i.e., $X := W \times \Omega$ and $\pi := (\varphi, \sigma)$) and $(\langle X, Z_+, \pi \rangle, (Y, Z_+, \sigma), h)$ (h := pr$_2 : X \to Y$) be a non-autonomous dynamical system generated by cocycle $\varphi$. Under the conditions of Theorem 2 we have

$$\rho (\varphi (n, u_1, \omega), \varphi (n, u_2, \omega)) \leq \psi (n, \rho (u_1, u_2)) \quad (10)$$

for all $n \in Z_+$, $u_1, u_2 \in W$ and $\omega \in \Omega$. Now to finish the proof of the theorem it is sufficient to apply Theorem 6.1.3 [5, Ch.VI] and Theorem 2 to the non-autonomous dynamical system $(\langle X, Z_+, \pi \rangle, (\Omega, Z_+, \sigma), h)$ and denote by $J_\omega := \text{pr}_1 (J_\omega)$, where $J$ is Levinson center of the dynamical system $(X, Z_+, \pi)$, $J_\omega := J \cap h^{-1} (\omega)$ and $h := \text{pr}_2$.

Let $\mathcal{M} \subset C(W)$, $(W, \varphi, (\Omega, Z_+, \pi))$ (respectively $(X, Z_+, \pi)$ ) be a cocycle (a skew-product dynamical system) generated by $DI(\mathcal{M})$ and let $I (J)$ be Levinson center of the cocycle $\varphi$ (respectively, skew-product dynamical system $(X, Z_+, \pi)$).

The set $I$ is said to be a chaotic attractor of $DI(\mathcal{M})$, if

1. the set $J$ is chaotic, i.e. $J$ is transitive and $J = \overline{\text{Per} (\sigma)}$, where $J$ is the Levinson center of the skew-product dynamical system $(X, Z_+, \pi)$ generated by $DI(\mathcal{M})$.
2. $I = pr_1(J)$.

Remark 2. 1. Theorem 3 it was proved in [6] for the special case, when $\psi(t,r) = Nq^t - (t,r) \in \mathbb{Z}_+ \times \mathbb{R}_+^2$, $N > 0$ and $q \in (0, 1)$.

2. The problem of the existence of compact global attractors for $DI(M)$ with finite $M$ (collage or iterated function system (IFS)) was studied before in works [1,2] (see also the bibliography therein). In [1,2] the statement close to Theorem 3 was proved. Namely:

1. in [1] it was announced the first and proved the second statement of Theorem 3, if $\psi(t,r) = q^tr$ ($t \in \mathbb{Z}_+^1$ and $q \in (0, 1)$);

2. in [2] they considered the case when $W$ is a compact metric space and every map $f \in M = \{f_1, f_2, \ldots, f_m\}$ ($i = 1, \ldots, m$) is contracting (not obligatory invertible). For this type of $DI(M)$ it was proved the existence of a compact global attractor $A$ such that for all $u \in A$ and almost all $\omega \in \Omega$ (with respect to certain measure on $\Omega$) the trajectory $\varphi(n, u, \omega) = U(n, u, \omega)$ was dense in $A$.

The problem of description of the structure of the attractor $I$ of $DI(M)$ in general case (when the maps $f \in M$ are not invertible) is more complicated. We plan to study this problem in one of our next publication.

References


Implementation of CNN–Based Viterbi Algorithm Using FPGA

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Abstract. Viterbi decoder is commonly utilized in digital communication because of its effectiveness in error correction. In this paper, we implement Viterbi algorithm based on Cellular Neural Network (CNN) in which cells play roles as nodes in the trellis diagram. The simulation and realization of CNN–based Viterbi decoder show the feasibility and the accuracy of our approach. Moreover, our results are compared with ones of current FPGA–based Viterbi decoders.

Keywords: Viterbi decoder, Cellular Neural Network, Field Programmable Gate Array.

1 Introduction

Digital communication systems have been widely used because of their better features comparing to analog ones Viterbi and Omura[1]. Generally, quality of digital system is estimated by the bit error rate (BER). Due to the channel impairment as noise and fading, channel coding is required to improve performance of digital signal. There are two common types of channel coding: block coding and convolutional coding. Different from block coding, convolutional coding can be utilized for both block of data and data stream. Hence, convolutional coding has been applied to amount of practical applications, especially in wireless communication.

An effective decoding algorithm for convolutional coding was proposed by Viterbi Viterbi[2]. Viterbi algorithm reduced the computational load by taking advantage of the special structure of the trellis, was therefore realized by an application–specific integrated circuit Dawid et al.[3]. Applying the Cellular Neural Network (CNN) to the Viterbi decoder is another approach for achieving full parallelism Kim et al.[4]. Kim et al. proposed a decoding structure with a circularly connected CNN cell array for a fully parallel Viterbi decoder. This structure, corresponding to CNN analog processing array, had specific features as lower power consumption, no path memory requirement and no trace–back requirement.
With the developing of electronic design automatic tools, Viterbi decoder is possible to be synthesized on an Field Programmable Gate Array (FPGA) chip Yeh et al.[5], Hema et al.[6]. Viterbi decoders have designed by different hardware description languages such as VHDL Zengliang et al.[7] or Verilog Kavinilavu et al.[8]. In addition, the principal parameters of decoders can vary for diverse applications, for example general wireless communication Ali et al.[9], Wong et al.[10]. However, these FPGA–based Viterbi decoders are always required memories to store the metric and previous state information. Therefore, parallel computing ability of decoder is reduced significantly. In this paper, different from conventional Viterbi decoders, we introduce the CNN–based Viterbi decoding which is expected to implement effectively on FPGA platform.

The paper is structured as follows. Viterbi decoding algorithm is reviewed in the next Section while our CNN–based architecture for Viterbi decoder is introduced in Section 3. Section 4 presents obtained results. Finally some conclusions are given in the last Section.

2 Viterbi decoding algorithm

In order to understand the Viterbi algorithm, we consider firstly the operation of a convolutional encoder. A convolutional code is described by two parameters $R$ and $K$, the code rate and the constraint length, respectively. Fig. 1 illustrates a convolutional encoder with $R = 1/2$ and $K = 3$. The encoder consists of 2–stages shift register and three modulo–2 adders. Two bit stream outputs are created when one bit stream input is applied.

Another way to represent a convolutional encoder is with a state diagram as shown in Fig. 2 where the states of register are noted as 00, 01, 10, and 11. There are two transitions from each state, corresponding to zero input bit (dashed path) and one input bit (solid path). Notice that the output word is written next to each path between two states.

Viterbi algorithm bases on trellis diagram for decoding as shown in Fig. 3. The algorithm calculates Hamming distances of paths entering each state node and chooses the path with smallest distance. On the other word, Viterbi algorithm performs minimum distance metric Kim et al.[4] as follows

$$D_{i,j} = \min \{D_{k,l} + d_{ij,kl}, (k, l) \in S\},$$

where $D_{i,j}$, $D_{k,l}$ and $d_{ij,kl}$ are the accumulated minimum distance metric at node $(i,j)$, the minimum total distance metric from the start node to the node $(k,l)$ and the distance metric for a data bit assigned on the path between the two nodes $(i,j)$ and $(k,l)$, respectively. $S$ is the set of the nodes in the neighbour of the node $(i,j)$.

A hardware realization of Viterbi decoder [11] as presented in Fig. 4 is often includes: branch metric unit (BMU), add compare and select unit (ACS), trace back unit (TBU), bit error rate monitor (BER), memory (MEM)
and memory management unit (MMU). MEM stores the accumulated metric and the previous state information while MMU generates addresses and read/write signals for MEM during different phases of operation. It is significant that the store requirement of Viterbi decoder increases exponentially with parameter $K$ Sklar\cite{12} as given by

$$u = h2^{K-1},$$

where $u$ is the amount of path storage and $h$ is the length of the information bit path history per stage. A novel structure of Viterbi encoder based on CNN is expected to solve this problem.

### 3 CNN–based architecture for Viterbi decoding algorithm

CNN has introduced by Chua and Yang\cite{13}, \cite{14} in which cells are connected directly to nearest neighbours. In addition, because each cell is consisted of a capacitor, resistors and voltage–controlled current, CNN has been implemented successfully through VLSI process Roska and Chua\cite{15}. Applications of CNN have been reported in a considerable amount of fields such as image...
processing, solving partial differential equation Chua and Yang[13], modelling complex system Arena et al.[16] or realizing FFT/IFFT algorithm Le et al.[17].

CNN is suitable for real–time parallel signal processing, thus CNN is possible in convolutional decoding. By employing CNN cells at nodes on the trellis diagram, a circularly connected two–dimensional analog CNN enabled very high performance Viterbi decoder Kim et al.[4]. In this part, we present a novel CNN structure for Viterbi decoding algorithm as in Fig. 5. Each cell communicates with two cells at the left–hand side and two cells at the right–hand side. This structure is similar to the trellis diagram where there are two paths entering each state. Cell \((i,j)\) (see Fig. 6) calculates the accumulate minimum distance metric \(P_{m_{i,j}}\) of distance metrics received from left–hand side cells \(P_{m_{k,l}}\) and the distance metric at the cell \(D_{m_{i,j}}\)

\[
P_{m_{i,j}} = \min\{P_{m_{k,l}} + D_{m_{i,j}}\}. \quad (3)
\]

Thus, from Eq. (1) and Eq. (3) it can be proved that the new CNN structure can implement Viterbi algorithm.
4 Implementation results

Firstly basic cell is designed in Matlab/Simulink as illustrated in the Fig. 7. Here accumulation distance metric is calculated and is transmitted to the next right-hand side cells. Moreover the cell also corrects output bits. After constructing basic cells, they are connected together to create a complete Viterbi decoder. The simulation results show the correction of Viterbi decoder.

Using the hardware description language VHDL, CNN–based Viterbi decoder is implemented on EP2C35F672C6 FPGA chip. Detail of basic cell
Table 1. Utilized FPGA resources for realization of CNN–based Viterbi decoder

<table>
<thead>
<tr>
<th>FPGA Components</th>
<th>Utilized</th>
</tr>
</thead>
<tbody>
<tr>
<td>Total logic elements</td>
<td>2%</td>
</tr>
<tr>
<td>Total registers</td>
<td>23</td>
</tr>
<tr>
<td>Total memory bits</td>
<td>0%</td>
</tr>
</tbody>
</table>

Table 2. Comparison with different architectures

<table>
<thead>
<tr>
<th></th>
<th>R</th>
<th>K</th>
<th>Memory</th>
</tr>
</thead>
<tbody>
<tr>
<td>This work</td>
<td>1/2</td>
<td>3</td>
<td>no</td>
</tr>
<tr>
<td>Zengliang et al. [7]</td>
<td>1/2</td>
<td>5</td>
<td>yes</td>
</tr>
<tr>
<td>Kavinilavu et al. [8]</td>
<td>1/2</td>
<td>7</td>
<td>yes</td>
</tr>
<tr>
<td>Wong et al. [10]</td>
<td>1/2</td>
<td>3</td>
<td>yes</td>
</tr>
</tbody>
</table>

Fig. 7. Detail of a basic cell constructed in Matlab/Simulink.

is displayed in Fig. 8 by Register–Transfer Level (RTL) Viewer tool. Table 1 presents the resource utilization of our design. Note that no memory requirement is obtained in this case. From the simulation of the decoder by Quartus II software as in Fig. 9, it is obvious that it takes two clock durations to complete the calculation.

The comparison of our work and other implementations is summarized in Table 2. Obviously, our architecture has the advanced features comparing to conventional others because of non–memory necessity.

5 Conclusion

In this paper, we propose a new architecture of Viterbi decoder based on CNN. Unlike current Viterbi decoders, our architecture does not require memories, the vital factor reduces the calculating rate. As the result our CNN–based Viterbi decoder could be implemented easily on programmable
Fig. 8. RTL representation of a basic cell synthesised on FPGA.

Fig. 9. Simulation of the CNN–based Viterbi decoder which is implemented by VHDL.

devices. The new Viterbi decoder has been shown to be able to implement in FPGA chip of Altera. Although the parameters of the decoder are chosen relatively small $R = 1/2$ and $K = 3$, they could be modified to support diverse applications in communication.

Acknowledgement

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References


The Effects of Diffusion of Individuals between Two Single-Species Populations

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Abstract: In the paper we focused on a general model for the growth of a single-species population with non-overlapping generations. The data we have used correspond to Nicholson’s blow-flies population and lie in the chaotic regime. The population was divided in two groups. If these groups evolve in distinct locations, their behavior is chaotic and, after a few generations, the initial small difference in number of individuals becomes big enough and behaves randomly. The question I want to answer in the paper is: What happens with the two populations if the individuals can migrate in both directions within the time intervals between their reproduction and death? The effect of coupling the two groups consisted in a rich dynamic behavior depending on the coupling strength. It was found that there is a consistent region where the coupling brings out the full synchronization of the two chaotic systems, two transition regions where an intermittent behavior was observed and two peripheral regions where control of chaos is shown to coexist with quasi-periodic and chaotic regimes.

Keywords: Single-species populations, Synchronization, Intermittent chaos, Control of chaos.

1. Introduction

According to May [1], models for population growth in a limited environment are based on two fundamental premises: a) the populations have the potential to increase exponentially; b) there is a density-dependent feedback that progressively reduces the actual rate of increase. By using a variety of data from field and laboratory populations, some researchers have proposed continuous or discrete models of population growth. The most known of these models is the logistic equation (Verhulst, 1838). Other simple models were introduced by May (1974), Li & Yorke (1975), May & Oster (1976), and Hassel et al (1976). Their models, which refer to single-species population with discrete, non-overlapping generation, predict that most of the populations show monotonic damping back to an equilibrium following a disturbance, with some exceptions of oscillatory damping or some sort of low-order limit cycles. They concluded that high-order limit cycles and chaos appear to be relatively rare phenomena in naturally occurring single-species populations. Guckenheimer et al (1987) have found that more realistic models of population growth, such as these that include overlapping generations, are more likely to exhibit complex behaviors. If data from laboratory population are used, even for these simple models, it was found that some populations will not exhibit stable equilibrium points but stable cycles or chaotic behavior [2]. That is because the laboratory situation (homogeneous environment, constant food supply, no competitors, no predators) make possible
an exaggerated non-linear behavior. In this paper we focused on a general model for growth of a single-species population with non-overlapping generations, namely

\[ N_{t+1} = f(N_t) = \lambda N_t (1 + a N_t)^{-b} \]  

where \( N_t \) and \( N_{t+1} \) are the populations in successive generations, \( \lambda \) is the finite rate of increase and \( a, b \) are constant defining the density-dependent feedback term. The values for parameters correspond to Nicholson’s blowflies and lie in the chaotic regime [3].

The population of blowflies was divided in two groups. If these groups evolve in distinct locations their behavior is chaotic and, after a few generations, the initial small difference in number of individuals becomes big enough and behaves randomly. The question I want to answer in the paper is: \textit{What happens with the two populations if the individuals can migrate in both directions within the time intervals between their reproduction and death?}

![Divergence of two nearby orbits](image)

**Fig. 1. Divergence of the two isolated populations versus time**

### 2. The Model of Two-coupled Single-species Populations

To answer the above question let us hereafter turn our attention towards the following system of two-coupled single-species populations:

\[ N_{t+1} = f(N_t) + c \left[ f(M_t) - f(N_t) \right], \quad M_{t+1} = f(M_t) + c \left[ f(N_t) - f(M_t) \right] \]  

where the coupling parameter \( c \) can be thought as the fraction of the two populations which migrate to the neighboring location. Throughout the paper I used the fixed parameter values \( \lambda = 60, a = 0.003 \) and \( b = 6 \). The total population \( \overline{N_t} = 3950 \) was divided in two unequal groups, \( N_t = 1950 \) and \( M_t = 1950 \).
If no change between the groups was permitted, the initial small difference in number of individuals, $ΔN_i = 50$, increased quickly and behaved chaotically (see Figure 1). The effect of coupling consisted in a rich dynamic behavior having the main features as follows.

2.1. Complete synchronization

If two or more chaotic systems are couple, it is possible that the attractive effect of a suitable coupling to counterbalance the trend of the trajectories to separate due to chaotic dynamics. Synchronization of chaotic systems can be explained by the suppression of expanding dynamics in the state space transversal to the synchronization manifold (here $M_i = N_i$). It is natural then to ask for which values of coupling strength $c$ the two systems will oscillate in a coherent and synchronized way.

Laureano et al [4] have demonstrated that, for this kind of coupling, the range of synchronization (in the linear approximation) is given by

$$0.5(1 - \exp(-\lambda_u)) < c < 0.5(1 - \exp(-\lambda_u))$$

where $\lambda_u$ is the Lyapunov exponent for the uncoupled map $f$. For our data it was found that $\lambda_u \approx 0.35$, so $c \in (0.15; 0.85)$ is possible. As an example, let consider $c = 0.16$. The synchronization takes place after 200 generations (see figure 2).

![Fig. 2. Evolution to synchronous state for $c = 0.16$](image)

Each of the systems shows chaos and their states are identical at each moment in time (full synchronization). To verify that the synchronous state is chaotic, a Lyapunov exponent versus coupling strength diagram was considered (see Figure 3).
If $c$ is chosen deep inside the interval $S$, the synchronous state is reached after only few steps (see Figure 4). Otherwise, if $c$ is chosen near the borders of $S$ the synchronization is hard to obtain, a lot of steps being necessary (e.g. 2000 steps for $c = 0.15$).

Fig. 3. Lyapunov exponent versus coupling strength

Fig. 4. Evolution to synchronous state for $c = 0.25$
2.2. Intermittent chaos

If the coupling strength \( c \) falls short of the critical value \( c_{\text{crt}} = 0.15 \) the synchronized state \( M_{\text{T}} = N_{\text{T}} \) becomes unstable and an intermittent dynamics is observed. Figure 5 shows the time evolution of the transverse coordinate \( DN_{\text{T}} = N_{\text{T}} - M_{\text{T}} \), for \( c = 0.1467 \). The time periods of synchronicity are interrupted by aperiodic chaotic bursts.

![Fig. 5. Time periods of synchronicity interrupted by aperiodic chaotic bursts](image)

![Fig. 6. A completely erratic state for \( c = 0.14 \)](image)
The basic intermittency mechanism comes from the competition between the trajectory instability of chaotic elements and the synchronization tendency due to the diffusion-type coupling [8]. For $c = 0.14$ the chaotic bursts were already merged so the synchronization started to dissolve into a completely erratic state (see Figure 6).

2.3. Stabilization to an ordered state

Outside the interval of synchronization the dynamics is quite complicated. For very small values of $c$ (weak coupling) the system behaves chaotically, the $N_f$ values being distributed over an entire interval. By increasing $c$ the chaotic distribution of $N_f$ comes undone in strips, thinner and thinner (see Figure 7).

At $c \cong 0.007$ the system entered a periodic regime, and was subjected to a sequence of changes from a $2^n$ - period cycle to a $2^{n-1}$ - period cycle. A 8-period cycle was obtained for $c \in (0.0072; 0.0080)$ (see Figure 8).

Then, a quasi-periodic regime with two strips appeared (Figure 9) which, in its turn, was changed by a 2-period cycle for $c \in (0.013; 0.11)$. This periodic regime is interrupted by windows corresponding to a 4-period cycle or even to thin windows of chaotic regime.

Beginning with $c \cong 0.1$ the number of steps required for stabilization to the 2-period cycle became bigger and bigger so, finally, the chaotic regime was reached. An analogous discussion can be done for $c \in (0.85; 1)$. 

Fig. 7. A part of the bifurcation diagram $N_f = N_f(c)$
Fig. 8. Time evolution of $N_t, M_t$ for $c = 0.0079$ (8-period cycle)

Fig. 9. Time evolution of $N_t$ for $c = 0.99$ (quasi-period regime)
3. Conclusions

The dynamics for many biological populations, which breed seasonally and have non-overlapping generations, are described by a density-dependent relation of the form \( N_{t+1} = f(N_t) = \lambda N_t \left( 1 + a N_t \right)^b \). If data from laboratory tests are used it was found that populations can exhibit even a chaotic behavior. Two almost identical populations, living in distinct locations, evolved so that the initial small difference in number of individuals became big enough and behaved randomly. If the individuals representing the two populations could migrate in both directions within the time intervals between their reproduction and death then a rich dynamic behavior depending on the coupling strength was observed. It was found that there is a consistent region where the coupling brings out the full synchronization of the two chaotic systems, two transition regions where an intermittent behavior was observed and two peripheral regions where control of chaos is shown to coexist with quasi-periodic and chaotic regimes.

References

Thermal Emission, Density and Light Scattering in water

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Abstract:
At temperatures above absolute zero, particles can emit as well as absorb and scatter electromagnetic radiation. Emission does not strictly fall within the bounds. It is more akin to such phenomena as luminescence than to elastic scattering. Consider an enclosure of dimensions large compared with any wavelengths under consideration, which is opaque but otherwise arbitrary in shape and composition. If the enclosure is maintained at a constant absolute temperature $T$, the equilibrium radiation field will be isotropic, homogenous, and unpolarized. At any point the amount of radiant energy per unit frequency interval, confined to a unit solid angle about any direction, which crosses a unit area normal to this direction in unit time is given by the Plank function $P_e = \frac{h\omega^3}{4\pi^3c^2} \cdot \frac{1}{\exp(h\omega/k_B T)-1}$. Optical properties of liquids are similar in many ways to those of solids. Electrically, there are metallic liquids such as mercury and molten iron, but the majority of common liquids are nonmetallic. As an illustration of a liquid we have chosen H$_2$O, a ubiquitous substance on our planet; water dominates not atmospheric processes but the chemistry of life. The optical properties of water had been studied for centuries; the modern results are scattered throughout the literature of many scientific fields. Dielectric functions calculated from refractive indices tabulated by Hale and Querry. Optical constants at wavelengths shorter than 0.2 µm were taken from the work of Kerr et al, where electronic absorption by H$_2$O begins. Dipole relaxation in H$_2$O maintains a high level of absorption from the microwave well in to the infrared; this also is the cause of the major differences between the optical properties of liquid and solid water.

Introduction
Let us contemplate the sea water near the costs. This water includes some particles of sodium chloride, the quantity of them is increasing or decreasing according to meteorological conditions and to electromagnetical moon’s functions. So the thermal emission and light scattering is up to particles assemble, phenomenon that affects the life of the aquatic organisms. Scattering or extinction measurements allow to determine size and distribution of sizes, shape and orientation, and chemical composition of the particles. If we consider that the salt (sodium chloride) consists of small scattering and absorbing particles, we write the complex refractive index of the salt as $m = \tilde{n} = i\tilde{\kappa}$. The light scattered from a fluid containing optically isotropic molecules. Many new features arise in the spectrum of light scattered from fluids containing optically anisotropic molecules, of course. In the dense water of these arises as
to how the collective motions of the fluid can be described by the rotation and how the thermal emission is changed. Let us note that the depolarization spectrum $I_{\nu}(\omega)$ consists of Lorentzian bands, all centered at zero frequency. If we have symmetric top rotors, the spectrum consists of a single band with a width $[\Delta D + 6\Theta]$ which depends only on the translational self-diffusion $D$ and on the rotational diffusion coefficient $\Theta$. Then, the spectrum appears to be split. So, it is presented a spectrum theory from a molecular point of view.

**OPTICAL PROPERTIES OF WATER:**

Optical properties of water are similar in many ways to that of solid. Electrically, there are metallic liquids such as mercury and molten iron, but the majority of common liquids are non-metallic. As liquid we have chosen the water because it is much important for the chemistry of life. During the centuries, the optical properties of water have been studied by the scientists. Irvine and Pollack (1968), Hale and Querry (1973) have calculated $n$ and $k$ for small particle calculations. Dielectric functions calculated by Hale and Querry. Instant to other scientists who use electron volts, we use a wavelength scale. As was done for ClNa, $\epsilon'$ is plotted linearly and $\epsilon''$ is plotted logarithmically. We also include on the $\epsilon''$ plot some results for solid H2O from Irvine and Pollack. The comparison of ice and water gives the field of the similarities and differences between the solid and the liquid phase. The insulating liquid H2O is not unlike the insulating solid ClNa. The electrons and vibrational excitation regions for both materials are well separated by a highly transparent region. Electronic absorption by H2O begins at about 0.2 µm. The major difference between clear water and salt water occurs longward of the vibrational absorption region. Dipole relaxation of H2O maintains a high level of absorption from the microwave well well into the infrared. This is the cause that makes the differences between the optical properties of liquid and solid water.
THERMAL EMISSION

In our case, let us have a spherical particle that is placed in the enclosure, then in equilibrium the distribution of radiation is uncharged. A spherical surface of radius R is surrounding the particle that has radius a, each element has entropy $dS$ of which is the source of a nearly plane wave that illuminates the
particle with irradiance \( P_e \cdot \frac{dS}{R^2} \). The amount of energy absorbed per unit time by the particle is:
\[
\int_0^\infty \int_S P_e C_{abs} \cdot \frac{dS}{d\omega} = 4\pi \int_0^\infty P_e C_{abc} \cdot d\omega.
\]

**CALCULATION OF THE ENTROPY PRODUCTION**

The sunrays are scattering on the water particles and this scattering is related to the water density, because of the salt (ClNa). By Gibbs equation we have:

\[
T dS = dE + P dV - \sum_{i=1}^n \mu_i d n_i
\]

where \( S, E, P, V, \mu_i, d_i \) are the total entropy, internal energy, pressure, volume, chemical potential of species \( i \) and the number of moles of salt respectively.

The integrated form of this equation:

\[
T S = E + PV - \sum_{i=1}^n \mu_i n_i.
\]

Dividing this later equation by \( V \), we get an expression for the local densities \( s, e, c_i \), corresponding to \( S, E, n_i \). So we have:

\[
T s = e + P - \sum_{i=1}^n \mu_i c_i.
\]
Then, substituting $sV$ to $S$, $eV$ to $E$, $ciV$ to $ni$ we have:

$$Td(Vs)=d(Ve)+PdV-\sum_{i} \mu id(Vci)$$

So; $V(TdS-de+\sum \mu idci)=dV(-Ts+e+p-\sum \mu ici)$.

The right part of this equation is zero and we obtain:

$$\frac{n}{Tds}=de-\sum_{i=1}^{\infty} \mu idci$$

This equation is a view of entropy density, internal energy density, and the molecules concentration. Alternatively, this equation can be expressed:

$$TdS=dq-\sum_{i=1}^{\infty} \mu idci$$

$dq$ is the increment of pure heat per unit volume.

CONCLUSION

These equations make evidence how much serious and important for the seawater life is the influence of the light scattering, the density of the water, and the thermal emission. The density of the water can be changes because of the pouring rain, the snow’s dissolving, the atmosphere thermal changes. If the seawater has high density of sodium chloride, then, it has the possibility to gather much more energy by the sunshine. It creates the presuppositions for growth of phytoplankton, zooplankton and fish.

References:
1. Asano S., 1979. Light Scattering properties of spheroidal particles
Bifurcation Analysis of Nociceptive Neurons*

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Abstract: As known, modification of specified slow sodium channels (Na\textsubscript{1,8}) in the membrane of nociceptive neurons is the basis of the pain perception by the human brain. The work is devoted to determination of parameters of the channels most sensitive to perceiving the painful signals. Using the bifurcation analysis of the model system describing the impulse activity of the membrane of mammalian nociceptive neuron we partition the parameter planes into the regions corresponding to stable and unstable periodic solutions. The left boundary of the region corresponds to subcritical Hopf bifurcation and emergence of the rough excitation in the form of large amplitude oscillations. The right boundary relates to supercritical Hopf bifurcation and appearance of the smooth excitation in the form of small large amplitude. Integrating inside the region of stable solutions we obtain the relationship between the parameter and frequency values. Bifurcation parameters such as the effective charge transfer of the activation gating system of the sodium channels and the maximal conductance of the channels play the main role in increasing the frequency and, hence, in transformation of the unpainful stimulus into the painful one. The results explain ionic mechanisms of action of analgesic drugs having high selectivity to Na\textsubscript{1,8} channels independently of the primary target of action.

Keywords Hopf bifurcation, Membrane model, Sodium channels, Nociceptive neuron.

1. Introduction

It is known that in response to injury of nervous system nociceptive neurons can become hyper-excitabile and generate spontaneous impulse activity of unusual frequency [1]. Perception of painful feeling is connected with activation of peripheral nociceptors recording painful signals and transmitting them by afferent nerve fibers to nociceptive neurons soma of which are in spinal ganglia. Low frequency of nerve impulses carries information about adequate tactile action and rise of the frequency for amplification of signal testifies about possible injury [2]. Slow sodium Na\textsubscript{1,8} channels are considered significant in generation of painful feeling since the enhancement of synthesis and functional activity of these channels is related to hyper-excitability of nociceptive neurons and high frequency neurophatic pain [3, 4]. The failure in the synthesis of the channels causes the reduce of neurophatic pain [5]. Modulation of activity of the channels by mediators of inflammation can lead to pathological state such as hyperalgesia (an increase of painful sensitivity) [6]. Hyperalgesia is removed by agents descending impulse activity of Na\textsubscript{1,8} channels [7]. That is why these

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agents are believed as the analgesic highly selective drugs [2]. The aim of the work is to answer the question: what parameters of the slow sodium Na$_1$V channels do maximal influence on pain signaling transduction? To answer the question it is necessary 1) to study relations between these parameters, an applied external stimulus and a type of stable solution of the model system describing the impulse activity of the nociceptive neuron; 2) to clarify what parameters do determine the possibility of the nociceptive neuron to generate spontaneously a signal of a painful range frequency?

2. The model

We have used the space-clamped Hodgkin-Huxley type model:

\[
\begin{align*}
\frac{dE}{dt} &= c_m \left( I - g_{Na} m^3 h (E - E_{Na}) - g_K n^4 (E - E_K) -
\right.

g_L (E - E_L) - g_{NaS} m_S^3 h_S (E - E_{Na}) \\
\frac{dm}{dt} &= \alpha_m (E) (1 - m) - \beta_m (E) m, \\
\frac{dh}{dt} &= \alpha_h (E) (1 - h) - \beta_h (E) h, \\
\frac{dn}{dt} &= \alpha_n (E) (1 - n) - \beta_n (E) n, \\
\frac{dm_s}{dt} &= \alpha_{m_s} (E) (1 - m_s) - \beta_{m_s} (E) m_s, \\
\frac{dh_s}{dt} &= \alpha_{h_s} (E) (1 - h_s) - \beta_{h_s} (E) h_s,
\end{align*}
\]

where $E$ is the membrane potential, the variables $m, h, n, m_s, h_s$ represent the probabilities of activation and inactivation of fast sodium, potassium and slow sodium channels, respectively.

The constants $c_m = 20$ pF, $g_{Na} = 40$ nS, $g_K = 20$ nS, $g_L = 5$ nS, $E_{Na} = 55$ mV, $E_K = -85$ mV, $E_L = -70$ mV are the membrane capacitance, the maximal conductance of the fast sodium, potassium and leakage ions channels and the reversal potentials for Na$^+$, K$^+$ and leakage ions.

The voltage-dependent expressions

\[
\begin{align*}
\alpha_m (E) &= \frac{0.115(1 + e^{(E + 70)/10})}{1 + e^{(E + 40)/42}}, \\
\beta_m (E) &= 0.015(1 + e^{(E + 25)/8}), \\
\alpha_h (E) &= 0.012(1 + e^{-(E + 43)/10}), \\
\beta_h (E) &= \frac{1.32}{1 + 0.2e^{(E + 10)/7}}, \\
\alpha_n (E) &= \frac{0.006(E + 45)}{1 - e^{-(E + 45)/12}}, \\
\beta_n (E) &= 0.13e^{-(E + 45)/30},
\end{align*}
\]
\[
\alpha_{m_s}(E) = e^{k_1(E + G) + d_1}, \quad \beta_{m_s}(E) = e^{k_2(E + G) + d_2}, \\
\alpha_{h_s}(E) = 0.0015e^{-(E + 4)/30}, \quad \beta_{h_s}(E) = \frac{0.01}{1 + 0.2e^{-(E + 10)/7}}
\]

describe rates of transfer of the activation and inactivation gating structures of ionic channels between the closed and open states.

According to the Boltzmann’s principle for the channel with the two-state open-closed structure the ratio of the number of open channels \(N_0\) to the number of closed channels \(N_C\) is determined by

\[
\frac{N_0}{N_C} = \frac{m_S}{1 - m_S} = \frac{Z_{\text{eff}}(E - \overline{E})/kT}{e^{m_S} - e^{m_S}}
\]

where \(Z_{\text{eff}}\) is the effective charge of the activation gating structure (in electron units) coupled with conformational change of the gating structure during the ion transfer through the membrane, \(k\) is the Boltzmann’s constant, \(T\) is the absolute temperature, \(\overline{E}\) is the electron charge, \(\overline{E}\) is the membrane potential such that \(N_0 = N_C\).

Then at \(E = \overline{E}\) for the activation gating structure of the slow sodium channels one can write \(\alpha_{m_S} = \beta_{m_S}\), whence it follows that the effective charge value of the activation gating structure can be gained as

\[
Z_{\text{eff}} = \frac{kT}{\overline{E}}(k_1 + k_2)
\]

3. Partition of the model parameter space into regions of qualitatively different solutions

To obtain relationship between the type of stable solution of the system, its parameters and an applied external stimulus it is sufficient to find points belonging to the boundary partitioning the parameter space of the model system into the regions of the qualitatively different types of stable solutions (steady states and stable periodic oscillations). For constructing the boundary the method of bifurcation analysis is applied.

On the \(I\) axis there are at least 3 bifurcation points \((I_0 < I_1 < I_2)\) [8]. For \(I < I_0\) and \(I > I_1\) there is a one-to-one correspondence between the type of steady state (unstable or stable) and the presence or absence of a stable periodic solution. For \(I \equiv I_0\) and \(I \equiv I_2\) the steady state is stable and a limit cycle does not exist. While the bifurcation parameter \(I\) increases in interval \((I_0 < I \equiv I_1)\) the steady state is stable and a stable and unstable periodic solutions coexist appearing via fold limit cycle bifurcation. The unstable periodic solution shrinks down to the rest state and makes it lose stability via subcritical Andronov-Hopf bifurcation.
Therefore, for \( I > I_1 \) the stable periodic oscillations of large amplitude exist both with decreasing and increasing \( I \) value. But for \( I_0 < I \leq I_1 \), the stable limit cycle of large amplitude is exhibited only with decreasing \( I \). Since for the Hodgkin-Huxley type system \( I_0 = I_1 \) for all the physiologically possible parameter values [9], the value of \( I_1 \) can be used as an approximate value of \( I_0 \). That is why the task of finding the boundary of qualitatively different types of stable solutions can be reduced to the more simple numerical task of constructing the boundary of various steady states (stable and unstable).

We write system (1) in the form

\[
\frac{dx}{dt} = F(x, p, I), \quad (2)
\]

where \( x = (E, m, h, n, m_S, h_S) \) is a vector of the phase coordinates, \( p = (g_{NaS}, k_1, k_2) \) is a vector of parameters which can be considered as bifurcation ones.

The method for determining the boundary points of the region of stable periodic solutions is reduced to the sequence of operations:

1) finding the equilibrium state of system (2) as a unique solution \( x_0(p, I) \) of the equation

\[
F(x, p, I) = 0,
\]

2) calculating the eigenvalues \( \{\lambda_i(p, I)\}_{i=1}^{6} \) of the Jacobian matrix

\[
J(p, I) = \left[ \frac{\partial F_i}{\partial x_j} \right]_{x=x_0(p, I)}, \quad i, j = 1, ..., 6,
\]

3) finding the parameter values satisfying the Hopf bifurcation, namely, arising of a pair of purely imaginary eigenvalues

\[
\lambda_1 = i\omega, \quad \lambda_2 = -i\omega, \quad \lambda_3 < 0, \quad \lambda_4 < 0, \quad \lambda_5 < 0, \quad \lambda_6 < 0.
\]

To determine values \( I_1(p) \) and \( I_2(p) \) at which maximal real part \( \lambda_m(p, I) \) becomes equal to zero the following algorithm is used.

1) The interval \([I_0, I_1]\) of possible values is discretized with \( k \) consecutive subintervals of length \( \Delta \).

In the case of existence even though one solution \( I_1(p) \) of equation

\[
\lambda_m(p, I) = 0,
\]
the $i$-subinterval involving the solution, is determined by the consecutive search beginning from the left side of the interval. The value of this solution is determined by linear interpolation.

2) The value of $I_2(p)$ is calculated in the interval $[i \Delta, I_0]$ by the method of bisection followed by linear interpolation.

The numerical solution of system (1) inside the obtained region of stable periodic solutions is found by a fourth-order Runge-Kutta method with a modified variable step size and Gear algorithm. The frequency of the periodic solution is calculated by the time values corresponding to local maxima.

### 4. Results and discussion

To elucidate the role of slow sodium channels in generation of the painful stimulus the maximal conductance of the slow sodium channels ($g_{NaS}$), the effective charge transfer of the activation gating system of the channels ($Z_{eff}$) and the shift ($G$) of the activation curve along the membrane potential axis have been used as variable parameters.

The family of the plane sections of the boundary partitioning the parameter space ($g_{NaS}, Z_{eff}, I$) into the regions of stable and unstable steady states are given in Fig.1a, b.

Inside the each found region the steady state is unstable and there is a stable limit cycle corresponding to stable periodic solution.

Stain-stepping effect of the left boundary of the region is related to features of arising limit cycles on the left and right sides of the boundary. The left boundary of the region corresponds to subcritical Hopf bifurcation and emergence of the rough excitation in the form of large amplitude oscillations. The right boundary relates to supercritical Hopf bifurcation and appearance of the smooth excitation in the form of small large amplitude.

As is seen, if $g_{NaS}$ = 0, periodic oscillations are absent at any stimulus value.

The minimal value of $g_{NaS}$ such that the oscillations emerge is equal to 14.9 nS at the stimulus – 142.5 pA and the value grows when $Z_{eff}$ increases.

When the effective charge is less than $e$ the periodic oscillations arise only by hyperpolarizing stimulus ($I < 0$ pA).

With increasing $Z_{eff}$ the steady periodic solutions region extends significantly and shifts in direction of depolarizing stimulus ($I > 0$ pA).

Integrating inside the constructed regions we obtain the relationship between the model parameter and frequency values. The examples of steady periodic solutions are represented in Fig.2 and Fig.3.

The periodic oscillations emerging on the left boundary of the region have large amplitude and small frequency. When moving inside the region from left to right an amplification of the external stimulus tends to change in amplitude and frequency of the nociceptive neuron. In other words, for the constant maximal conductance of the slow sodium channels and effective charge transfer of the activation gating system an enhancement of the external stimulus leads to the increase of the frequency of periodic oscillations and then their disruption.
Fig. 1. The examples of the plane sections of the boundary partitioning the parameter space \((g_{NaS}, Z_{eff}, I)\) into the regions of stable and unstable steady states. Each section is constructed with 800 points on the \((g_{NaS}, I)\) plane corresponding to 700 net values of the parameter \(g_{NaS}\). Values \(G=10\, mV\), \(Z_{eff} = \{3.5\overline{7}, 4\overline{7}, 4.5\overline{7}, 5\overline{7}, 5.5\overline{7}, 6\overline{7}, 6.5\overline{7}, 7\overline{7}, 8\overline{7}, 9\overline{7}, 10\overline{7}\}\).

The periodic oscillations emerging on the left boundary of the region have large amplitude and small frequency. When moving inside the region from left to right an amplification of the external stimulus tends to change in amplitude and frequency of the nociceptive neuron. In other words, for the constant maximal conductance of the slow sodium channels and effective charge transfer of the activation gating system an enhancement of the external stimulus leads to the increase of the frequency of periodic oscillations and then their disruption.
Fig. 2. The examples of steady solutions for $Z_{\text{eff}} = 6.5\bar{E}$, $g_{\text{NaS}} = 100\text{nS}$ and various values of stimulus.

Fig. 3. The examples of steady solutions $Z_{\text{eff}} = 6.5\bar{E}$, $I = 0$ and different values of $g_{\text{NaS}}$.

The periodic oscillations emerging on the left boundary of the region have large amplitude and small frequency. When moving inside the region from left to right an amplification of the external stimulus tends to change in amplitude and frequency of the nociceptive neuron. In other words, for the constant maximal conductance of the slow sodium channels and effective charge transfer of the activation gating system an enhancement of the external stimulus leads to the increase of the frequency of periodic oscillations and then their disruption.
When moving inside the region from bottom to top a growth of the maximal conductance of the slow sodium channels for the constant stimulus value results in increase of the frequency, period doubling and also failure of periodic oscillations (Fig.3). Thus, both factors, namely, decrease of effective charge transfer of the activation gating system of the slow sodium channels for the constant maximal conductance of the channels and decrease of the maximal conductance of the slow sodium channels for the constant stimulus decline the frequency of impulse activity. Since an increase in the frequency of impulse activity of nociceptive neurons is related to the emergence of neuropathic pain, our findings indicate the direction of looking for chemical agents possessing analgesic properties.

3. Conclusion
The form of the constructed regions demonstrates that ability of each parameter to be bifurcation one significantly depends on the other parameter values. Thus, the conclusions about bifurcation properties of the system parameters are determined by the investigated point in the parameter space.
The character of changes in the system solutions and in the frequency of periodic solutions can be used in searching of chemical agents aimed for selective removal of neuropathic pain.

References

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Modeling of a Scenario of Transition to Chaos in Plasma through Sub-harmonic Bifurcation

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Abstract: We report on theoretical modeling of a scenario of transition to chaos in plasma through sub-harmonic bifurcations, based on scale relativity theory. Experimental results show that a discharge plasma which self-structures as concentrically multiple double layers is able to evolve towards chaos through a cascade of spatio-temporal sub-harmonic bifurcations when the applied constrain is gradually increased. By considering that the plasma particles (electrons, ions, neutrals) move on continuous but non-differentiable curves (because of the collisions), i.e., on fractal curves, we develop a model based on scale relativity theory capable to explain the self-structuring of the electrically conductive plasma as multiple double layers, as well as the transition to chaos through sub-harmonic bifurcations of these structures when the applied constrain increases. Thus, by considering that the complexity of the physical processes in plasma is replaced by fractality, it is no longer necessary to use the whole classical “arsenal” of quantities from the standard physics (differentiable physics). The plasma system will behave as a special interaction-less “fluid” by means of geodesics in a fractal space. In such context, by using the fractional revival formalism, a Reynold’s fractional criterion of evolution to chaos through cascade of spatio-temporal sub-harmonic bifurcations was stated. A good agreement between the experimental results and those provided by the theoretical model was obtained.

Keywords: Chaotic modeling, Fractal, Scale relativity theory, Pattern formation, Sub-harmonic bifurcations.

1. Introduction

Concentrically multiple double layers are complex nonlinear potential structures in plasmas, consisting of two or more double layers attached to the anode of a dc glow discharge [1] or to a positively biased electrode immersed into plasma [2]. They appear as several bright and concentric shells attached to the electrode. The number of layers depends on the background gas, on its pressure, on the electrode voltage and on the discharge current. The axial profile of the plasma potential has a stair step shape, with potential jumps close to the ionization potential of the working gas [2]. At high values of the voltage applied to the electrode, the multiple double layer structure passes into a dynamic state which consists of periodic disruptions and re-aggregations of the constituent double layers [2].

In many systems where deterministic chaos arises, spatial and temporal structures were also experimentally observed. For time scales large with respect
to the inverse of maximum Lyapunov exponent, deterministic trajectories can be replaced by families of potential trajectories and the concept of definite positions by that of probability density. This allows the description of chaos in a stochastic way by a diffusion process as described in [3]. By considering that the particles movement takes place on continuous but non-differentiable curves, \textit{i.e.} on fractal curves, the scale relativity theory approaches the chaotic effects in the same way as in [3], except diffusion, which becomes a spatio-temporal scale dependent process [4].

The theory of scale relativity (SR) applies the principle of relativity to scale transformations. In such conjecture, the principle of SR requires that the fundamental laws of nature apply whatever the state of scale of the coordinate system. In order to include the non-differentiable quantum motion, the quantum space-time must be considered relative and fractal, as described in [5]. In this theoretical framework, it is not necessary to endow a point particle with mass, energy, momentum or velocity. The particle may be reduced to and identified with its own trajectory [4].

Here we report on experimental results revealing how the electrically conductive plasma can self-structure in the form of concentrically multiple double layers. We show that these new structures evolve towards chaos through a cascade of spatio-temporal sub-harmonic bifurcations when the applied constraint in form of the voltage applied on a supplementary electrode increases. A theoretical model based on the scale relativity theory is developed, and its results are found to be in good agreement with the experimental data.

2. Experimental Results

The experiments were performed in a plasma diode, schematically presented in Fig. 1. Plasma is created by an electrical discharge between the hot filament (marked by F in Fig. 1) as cathode and the grounded metallic tube as anode. The plasma was pulled away from equilibrium by gradually increasing the voltage applied to a tantalum disk electrode (marked by E in Fig. 1) with 1 cm diameter, under the following experimental conditions: argon pressure $p = 10^{-2}$ mbar, plasma density $n_{pl} = 10^9$ cm$^{-3}$ and electron temperature $kT_e \cong 2$ eV. When the voltage on the electrode reaches $V_E \cong 55$ V, a double layer structure appears in front of the electrode (see photo in Fig. 2(a)). The structure is in dynamic state, phenomenon extensively described in [6]. The FFT of the oscillations of the current collected by the electrode is shown in Fig. 3(a). By a further increase of the voltage on the electrode, new double layers develop in front of the electrode, giving rise to a dynamic multiple double layer structure (see photos in Fig. 2(b)-2(e)). Simultaneously with every new double layer formation, a new sub-harmonic appears in the FFT spectrum of the current oscillations (see Fig. 3(b)-3(f)). Thus, we recorded, in fact, spatio-temporal bifurcations in the plasma system (sudden changes in the spatial symmetry and in the temporal dynamics of the plasma system). At high values of the applied potential, the plasma system passes into a chaotic state, characterized by uncorrelated and intermittent oscillations (see FFT’s of these oscillations in Fig. 3(g)).
3. Theoretical Model
Assuming that the discharge plasma particles (electrons, ions, neutrals) motions take place on continuous but non-differentiable curves, \textit{i.e.} on fractal curves of fractal dimension $D_F$, the dynamics of such a system can be described by the complex operator (see [7]):

![Fig. 1. Schematic of the experimental setup (F – filament, A – anode, E – supplementary electrode, $U_1$ – power supply for heating the filament, $U_2$ – power supply for discharge, PS – power supply for supplementary electrode bias, $R_1$, $R_2$ – load resistors).](image1)

![Fig. 2. Photos of the multiple double layers in different stages (at different increasing value of the potential applied on the electrode) of its formation.](image2)
Fig. 3. FFT of the oscillations of the current collected by the supplementary electrode, at different increasing values of the voltage applied on it.

\[ \dot{\mathbf{V}} = \mathbf{V} - i\mathbf{U} \]  

(2)

where

\[ \dot{V} = \frac{\partial}{\partial t} V + \nabla \cdot V - i \nabla V \delta(-D_r)^{-1} \Delta \]  

(1)

The real part \( V \) of the complex field \( \hat{V} \) represents the standard classical velocity, which is differentiable and independent of resolution \( dt \), while the imaginary part \( U \) is a new quantity arising from fractality, which is non-differentiable and resolution-dependent. The quantity \( \delta \) is Nottale’s coefficient and corresponds to the fractal – non-fractal transition. The geodesics equation (a generalization of the first Newton’s principle) can be written in the form:

\[ \frac{d\hat{V}}{dt} = \frac{\partial}{\partial \Delta} \hat{V} + (\hat{V} \cdot \nabla) \hat{V} - \overrightarrow{\nabla} \Delta \hat{V} \]  

(3)
which is a Navier-Stokes type equation with the imaginary “viscosity coefficient”:
\[
\overline{\eta} = i \partial (dt)^{(4/D_r)-1}
\]  
(4)

If the motions of the fractal fluid are irrotational, i.e. \( \nabla \times \mathbf{V} = 0 \), we can choose \( \mathbf{V} \) on the form:
\[
\dot{\mathbf{V}} = -2i \partial (dt)^{(4/D_r)-1} \nabla \ln \psi
\]  
(5)

with \( \psi \) being the scalar potential of the complex velocity. Then, by substituting (5) in (3) and using the method described in [9] it results:
\[
\frac{d\dot{V}}{dt} = -2i \partial (dt)^{(4/D_r)-1} \left( \frac{\partial \ln \psi}{\partial t} - 2i \partial (dt)^{(4/D_r)-1} \frac{\nabla \psi}{\psi} \right) = 0
\]  
(6)

This equation can be integrated in a universal way and yields:
\[
\begin{align*}
\dot{L} \psi &= 0 \\
\dot{L} &= 4 \partial^2 (dt)^{(4/D_r)-2} \Delta + 2i \partial (dt)^{(4/D_r)-1} \frac{\partial}{\partial t}
\end{align*}
\]  
(7a,b)

up to an arbitrary phase factor which may be set to zero by a suitable choice of the phase of \( \psi \). Equation (7a) is of Schrödinger type. If the fractal fluid is placed into the external field \( W \), then (7a,b) become:
\[
\begin{align*}
\dot{L} \psi &= 0 \\
\dot{L} &= 4 \partial^2 (dt)^{(4/D_r)-2} \Delta + 2i \partial (dt)^{(4/D_r)-1} \frac{\partial}{\partial t} - \frac{W}{m_0}
\end{align*}
\]  
(7c,d)

where \( m_0 \) is the rest mass of the fractal fluid particle.

In our experiment the particle interacts with an external potential which results in a discontinuous change in momentum. This interaction can be theoretically modeled by means of a one-dimensional infinite square-well potential that confines a particle to a box of given width. Localized time-dependent solutions for this problem can be constructed in a very straightforward way from solutions of the free-particle problem [9].

The one-dimensional infinite square-well potential confines a particle to a box of width \( a \) and is described by
A particle of mass $m_0$ is placed in this potential. The energy eigenvalues and eigenstates are found by solving the time-independent Schrödinger-type equation (as (7c,d)). The discrete energy eigenvalues are

$$E_n = 2m_0 D \left( \frac{n \pi}{a} \right)^2, \quad D = \mathcal{D}(d) \left( \frac{2}{D_e} \right)^{-1} \quad (9a,b)$$

for positive integer $n$. The scalar potential eigenvalues corresponding to the energy eigenvalues (9a,b) are

$$\phi_n(x) = \begin{cases} \sqrt{\frac{n}{a}} \sin \left( \frac{n \pi x}{a} \right), & |x| \leq a/2, \quad n \text{ even} \\ \sqrt{\frac{n}{a}} \cos \left( \frac{n \pi x}{a} \right), & |x| \leq a/2, \quad n \text{ odd} \\ 0, & |x| > a/2 \end{cases} \quad (10)$$

Let’s consider the expression $k_n^2 = n^2 k_i^2$ induced by the generalized coherence (in the present context, the physical mean of the generalized coherence refers to the generation of multiple double layers) (see for details [10]). Then, the relation

$$E_n = E_i n^2, \quad E_i = 2m_0 D \frac{k_i^2}{a} = \frac{\pi}{a} \quad (11a-c)$$

expanded around $\bar{n}$ either in the form

$$E_n = E_i \bar{n}^2 + 2E_i (n - \bar{n}) + E_i (n - \bar{n})^2 \quad (12)$$

or in the form

$$E_n = E_i + 4\pi m_0 D \left[ \frac{(n - \bar{n})}{T_\phi} + \frac{(n - \bar{n})^2}{T_\phi^2} \right] \quad (13a,b)$$

induces the characteristic times:
Because $T_\beta$ is independent of $\pi$, expression (14b) defines a universal time scale. By means of relations (14b) and (11b), a characteristic frequency can be associated:

$$f_0 = \frac{1}{T_\beta} = \frac{k_1}{2\pi} \left( \frac{E_1}{2m_0} \right)^{1/2}$$  \hspace{1cm} (15)$$

Let us evaluate expression (15) with respect to the experimental results. Thus, by identifying $a$ with a characteristic length of the double layer, namely the width of the double layer, $a = (2\varepsilon_0 V / \varepsilon n_0)^{1/2}$, expression (15) takes the form

$$f_0 = \frac{1}{2} \left( \varepsilon n_0 E_1 \right)^{1/2} \left( m_0 \varepsilon_0 V \right)$$  \hspace{1cm} (16)$$

where $E_1$ is the ion energy, $m_0$ is the ion mass, $n_0$ is the ion density in the double layer and $V$ is the voltage on the electrode. For our experimental conditions $f_0 \cong 150$ kHz. This value is in a good agreement with those experimentally obtained by us (see Fig. 3).

Now, the fractional revival formalism may be applied. A fractional revival of a physical function occurs when a physical function evolves in time to a state that can be described as a collection of spatially distributed physical sub-functions that each closely reproduces the initial physical function shape – see for details [11]. In this context, we write the particle’s initial ($t = 0$) scalar potential velocity function in the square well as:

$$\psi(x, t = 0) = \psi_i(x)$$  \hspace{1cm} (17)$$

We expand this scalar potential velocity function by using the relation (10)

$$\psi_i(x) = \sum_{n=1}^{\infty} c_n \phi_n(x)$$  \hspace{1cm} (18)$$

with

$$c_n = \int_{-\infty}^{\infty} \phi_n(x) \psi_i(x) dx$$  \hspace{1cm} (19)$$
By using the time scale $T_\beta$, the time evolution of the system is found from the Schrödinger’s type equation (the charge transport takes place on fractal curves – see the operator $\hat{L}$ from (7c,d))

$$4\partial^2(dt)^{1/(2D_c)-1}\frac{\partial^2\psi}{\partial x^2} + 2i\partial(dt)^{1/(2D_c)-1}\frac{\partial\psi}{\partial t} - \frac{W}{m_0}\psi = 0 \quad (20)$$

to be

$$\psi(x,t) = \sum_n \exp\left[-i2\pi(n/T_\beta)^2\right] c_n \psi_n(x) \quad (21)$$

A function $F(n)$ whose domain is restricted to the integers ($n \in \mathbb{Z}$) can be written as a finite sum of exponentials if and only if it is $r$ periodic, that is, if there is an integer $r$ such that $F(n) = F(n + r)$ for all $n$. Such a finite sum is called the finite Fourier series. In our case, we identify

$$(n/T_\beta)^2 = \frac{p}{q} \quad (22)$$

for relatively prime integers $p$ and $q$ (that is, $p/q$ forms a simplified fraction). In terms of frequency, expression (48) takes the form:

$$f(p,q) = \frac{1}{T(p,q)} = \frac{q}{p} f_0, \quad f_0 = \frac{1}{T_\beta} \quad (23a,b)$$

The condition $p > q$ specifies the presence of sub-harmonics in the Fourier spectra. Now, through the fractal expressions

$$E_i = \frac{m_0V_i^2}{2} = 4\pi m_0 D \frac{p}{q} f(p,q) \quad (24)$$

we can introduce the Reynolds’s “fractional” criterion.
\[ \text{Re}(p,q) = \frac{V_c L}{\nu} = \frac{p}{q} \tag{25} \]

where we used the substitutions:

\[ V_c = V, \quad L_c = Vf^{-1}(p,q), \quad \nu = 8\pi D \langle dt \rangle^{2/D} \tag{26a-c} \]

From (25) and [12] it results a critical value for the Reynolds number, \( \text{Re}_c \). Up to this value the fractal fluid flow becomes turbulent. Because from (23a) with \( p > q \) results sub-harmonics for \( \text{Re}_c \), according to [13-15] a criterion of evolution to chaos through a cascade of spatio-temporal sub-harmonic bifurcations is stated. We call this criterion the “fractional” criterion of transition towards chaos.

3. Conclusions
A scenario of transition to chaos through cascade of sub-harmonic bifurcations was experimentally evidenced in connection to the generation and dynamics of concentric multiple double layers in plasma.

By considering that the particles movements in the gas discharge plasma take place on fractal curves, a mathematical model according with the scale relativity theory was developed in order to describe its dynamics. In this model, the self-structuring of a discharge plasma as concentric multiple double layers was analyzed, a good similarity between the experimental and theoretical dependences being obtained. By using the fractional revival formalism, a Reynold’s fractional criterion of evolution to chaos through cascade of spatio-temporal sub-harmonic bifurcations was stated.

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References
Height Field Representation and Compression Using Fractal Interpolation Surfaces

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Abstract. Height fields provide efficient means for representing surface elevation data which can be used for rendering 3D terrains or landscapes. In this paper, a novel method for representing height fields using fractal interpolation techniques is presented. The proposed methodology allows describing natural surfaces with an intrinsic fractal structure in a more convenient manner. Specifically, fractal interpolation surfaces constructed on rectangular domains have been used. Results indicate the feasibility and advantages of the proposed method in terms of quality of representation as well as compression ratios.

Keywords: Fractals, IFS, Interpolation, Surfaces.

1 Introduction

Height fields provide an efficient tool for representing surface elevation data and are often used, among other applications, in 3D computer graphics for rendering 3D terrains or landscapes. A height field is essentially a 2D array of height values and is usually stored as a raster image; the pixel intensity corresponds to the height at the location defined by the pixel coordinates. An example of landscape rendering based on a height field is shown in Figure 1.

\textbf{Fig. 1.} Rendering of a landscape defined by a height field.
Fractal interpolation as defined by Barnsley[1] and other researchers is based on the theory of iterated function systems. It provides an alternative to traditional interpolation techniques, aiming mainly at data which present detail at different scales or possess some degree of self-similarity. These properties define an irregular, non-smooth, structure which is inconvenient to be described by using elementary functions such as polynomials. A fractal interpolation function is a continuous function whose graph is the attractor of an appropriately chosen iterated function system. In case this graph, usually of non-integral dimension, belongs to the three-dimensional space and has Hausdorff - Besicovitch dimension between 2 and 3, the resulting attractor is called fractal interpolation surface.

In this paper, a novel methodology for representing height fields using fractal interpolation techniques is introduced. Our motivation stems from the fact that natural surfaces, such as earth terrains, present an intrinsic fractal structure, i.e. detail at multiple scales and some degree of self-similarity. The most important and non-trivial part for constructing fractal interpolation surfaces on rectangular domains involves ensuring their continuity. We also present the application of the proposed methodology to terrain data, indicating its feasibility and advantages in terms of quality of representation as well as compression ratios.

The paper is structured as follows. In Section 2 we briefly review the theory of iterated function systems. The necessary concepts of fractal interpolation surfaces, focusing on the proposed construction on rectangular domains, are introduced in Section 3. In Section 4, we present the proposed methodology for height field representation and compression using the surfaces of the previous section. Section 5 contains the result of the application of our method to terrain data, in terms of quality of representation as well as compression ratios. Finally, Section 6 summarizes our conclusions and indicates areas of further work.

2 Iterated function systems

Let $X, Y \subset \mathbb{R}^n$. A function $f: X \rightarrow Y$ is called a H"older function of exponent $a$ if

$$|f(x) - f(y)| \leq c|x - y|^a$$

for $x, y \in X$, $a \geq 0$ and for some constant $c$. Note that, if $a > 1$, the functions are constants. Obviously, $c \geq 0$. The function $f$ is called a Lipschitz function if $a$ may be taken to be equal to 1. A Lipschitz function is a contraction with contractivity factor $c$, if $c < 1$. An iterated function system, or IFS for short, is a collection of a complete metric space $(X, \rho)$ together with a finite set of continuous mappings $w_n: X \rightarrow X$, $n = 1, 2, \ldots, N$, where $\rho$ is a distance between elements of $X$. It is often convenient to write an IFS formally as \{X; w_1, w_2, \ldots, w_N\} or, somewhat more briefly, as \{X; w_{1\ldots N}\}. 
The associated map of subsets $W: \mathcal{H}(X) \to \mathcal{H}(X)$ is given by

$$W(E) = \bigcup_{n=1}^{N} w_n(E) \text{ for all } E \in \mathcal{H}(X),$$

where $\mathcal{H}(X)$ is the metric space of all non-empty, compact subsets of $X$ with respect to some metric, e.g. the Hausdorff metric. The map $W$ is called the *collage map* to alert us to the fact that $W(E)$ is formed as a union or ‘collage’ of sets. Sometimes $\mathcal{H}(X)$ is referred to as the “space of fractals in $X$” (but note that not all members of $\mathcal{H}(X)$ are fractals).

If $w_n$ are contractions with corresponding contractivity factors $s_n$ for $n = 1, 2, \ldots, N$, the IFS is termed *hyperbolic* and the map $W$ itself is then a contraction with contractivity factor $s = \max\{s_1, s_2, \ldots, s_N\}$ (Barnsley[1], Theorem 7.1, p. 81). In what follows we abbreviate by $f^k$ the $k$-fold composition $f \circ f \circ \cdots \circ f$. The graph of $f$ is the set of points $G(f) = \{(x, f(x)) : x \in X\}$.

The attractor of a hyperbolic IFS is the unique set $A_\infty$ for which $\lim_{k \to \infty} W^k(E_0) = A_\infty$ for every starting set $E_0$. The term attractor is chosen to suggest the movement of $E_0$ towards $A_\infty$ under successive applications of $W$. $A_\infty$ is also the unique set in $\mathcal{H}(X)$ which is not changed by $W$, so $W(A_\infty) = A_\infty$, and from this important perspective it is often called the *invariant set* of the IFS.

### 3 Rectangular subdomain fractal interpolation surfaces

Fractal interpolation surfaces constructed as attractors of iterated function systems were first proposed by Peter R. Massopust[4], where he considered the case of a triangular domain with coplanar boundary data. A slightly more general construction of such fractal surfaces was later presented by Jeffrey S. Geronimo and Douglas Hardin[3], where the domain used was a polygonal region with arbitrary interpolation points but same contractivity factors. Here, we focus on fractal interpolation surfaces constructed on rectangular domains with arbitrary boundary data and same contractivity factors.

Let $D$ be a closed nondegenerate rectangular region in $\mathbb{R}^2$ and let $S = \{q_0, q_1, \ldots, q_{m-1}\}$ be $m$, with $m > 4$, distinct points in $D$ such that $\{q_0, q_1, q_2, q_3\}$ are the vertices of $D$. Given real numbers $z_0, z_1, \ldots, z_{m-1}$ we wish to construct a function $f$ such that $f(q_j) = z_j$, $j = 0, 1, \ldots, m - 1$ and whose graph is self-similar. Notice that the constructed FIS in the present work is not self-affine since it is resulting from bivariate functions. Let us denote by $C(D)$ the linear space of all real-valued continuous functions defined on $D$, i.e. $C(D) = \{f: D \to \mathbb{R} \mid f \text{ continuous}\}$. The basic idea is to decompose $D$ into $N$ subrectangles $R_1, R_2, \ldots, R_N$ with vertices chosen from $S$ and define affine maps $L_i: D \to R_i$ and contractions $F_i: D \times \mathbb{R} \to \mathbb{R}$, $i = 1, 2, \ldots, N$ such that $\Phi$ defined by

$$\Phi f(x, y) = F_i(L_i^{-1}(x, y), f(L_i^{-1}(x, y)))$$

(1)
maps an appropriate subset of \( C(D) \) onto itself. If \( L_i \) is invertible, \( G(f) \) is mapped onto \( G(\Phi(f)|_{R_i}) \) by \( (L_i(x,y), F_i(x,y, f(x)) \). Henceforth we assume that \( \{ R_i \}_{i=1}^N \) consists of nondegenerate rectangles whose interiors are nonintersecting, \( L_i^{-1}(R_i) = D \) and that the set of vertices of \( \{ R_i \}_{i=1}^N \) equals \( S \). Let \( k: \{ 1, 2, \ldots, N \} \times \{ 0, 1, 2, 3 \} \rightarrow \{ 0, 1, \ldots, m - 1 \} \) be such that \( \{ q_{k(i,j)} \}_{j=0}^3 \) gives the vertices of \( \{ R_i \}_{i=1}^N \).

Let \( i \in \{ 1, 2, \ldots, N \} \). Since \( D \) and \( R_i \) are nondegenerate, there is a unique invertible affine map \( L_i: \mathbb{R}^2 \rightarrow \mathbb{R}^2 \) satisfying

\[
L_i(q_j) = q_{k(i,j)}, \quad j = 0, 1, 2, 3.
\]

Let \( s_i \) be given such that \( |s_i| < 1 \) and \( F_i: \mathbb{R}^3 \rightarrow \mathbb{R} \) be defined by

\[
F_i(x, y, z) = a_i x + b_i y + g_i xy + s_i z + c_i,
\]

where \( a_i, b_i \) and \( c_i \) are uniquely determined by

\[
F_i(q_j, z_j) = z_{k(i,j)}, \quad j = 0, 1, 2, 3.
\]

With these definitions for \( L_i \) and \( F_i \) we have \( \Phi(f)|_{R_i} \in C(R_i) \) and \( \Phi(f)(q_{k(i,j)}) = z_{k(i,j)}, \quad j = 0, 1, 2, 3, \) whenever \( f \in C(D) \) and \( f(q_j) = z_j, \quad j = 0, 1, 2, 3. \) If \( R_i \) and \( R_{i'} \) are adjacent rectangles with common edge \( \overline{q_j q_{j'}} \), it remains to be determined whether \( \Phi f \) is well-defined along \( \overline{q_j q_{j'}} \), i.e., whether \( \Phi f \) satisfies the “join-up” condition

\[
F_i(L_i^{-1}(x,y), f(L_i^{-1}(x,y))) = F_{i'}(L_{i'}^{-1}(x,y), f(L_{i'}^{-1}(x,y)));
\]

for all \( (x,y) \in \overline{q_j q_{j'}} \). We consider the case where the graph associated with the tessellation \( \{ R_i \}_{i=1}^N \) has chromatic number 4. The chromatic number of a graph is the least number of symbols required to label the vertices of the graph so that any two adjacent vertices (i.e., joined by an edge) have distinct labels. Since each edge is part of some \( R_i \) this implies the vertices \( \{ q_j \}_{j=0}^m \) can be labelled with \( l = l(j) \in \{ 0, 1, 2, 3 \} \) such that the vertices of each \( R_i \) have distinct labels. For \( i = 1, 2, \ldots, N \) and \( j = 0, 1, 2, 3 \) let \( k(i, j) \) be determined by the condition

\[
k(i, l(j')) = j' \quad \text{for all vertices } q_{j'} \text{ of } R_i.
\]

Then, Eqs. (2) and (4) become

\[
L_i(q_{l(j)}) = q_j, \quad F_i(q_{l(j)}, z_{l(j)}) = z_j
\]

for each of the vertices \( q_j \) of \( R_i \).

Let \( C_B(D) \) denote the collection of continuous functions \( f \) such that \( f(q_j) = z_j, q_j \in \partial D \).

**Theorem 31** Suppose the graph associated with \( \{ R_i \}_{i=1}^N \) has chromatic number 4. Let \( L_i \) and \( F_i, i = 1, 2, \ldots, N, \) be determined by (3) and (5) with \( s_i = s(|s| < 1) \). Let \( \Phi \) be defined by (1). Then \( \Phi: C_B(D) \rightarrow C_B(D) \) is well-defined and contractive in the sup-norm with contractivity \( s \). Furthermore \( \Phi f(q_j) = z_j, j = 0, 1, \ldots, m - 1 \) and \( f \in C_B(D) \).
Proof. See Drakopoulos and Manousopoulos[2] □

Then the corresponding IFS is of the form $\{\mathbb{R}^3; w_{1-N}\}$, where

$$w_i(x, y, z) = (L_i(x, y), F_i(x, y, z)).$$

An illustration of this is shown in Fig. 2, where the left part indicates the vertices and connecting edges of $D$ and the middle and right parts of the figure indicate where these vertices are mapped by the domain contractions. For larger data sets, this pattern is repeated as necessary.

![Fig. 2. Rectangular domain contractions to satisfy join-up conditions.](image)

4 Height fields

A height field or relief map (see e.g. Theoharis et al.[5], p. 505) is defined as a 2D array of height values:

$$H = \{(x_i, y_j, z_{ij} : i = 0, 1, \ldots, M \text{ and } j = 0, 1, \ldots, N\},$$

where the $x$, $y$ coordinates define a rectangular grid on the plane and the $z$ coordinate defines the height. The underlying grid is usually regular, i.e.

$$x_i = x_0 + i\Delta_x, \quad y_j = y_0 + j\Delta_y,$$

for every $i = 0, 1, \ldots, M$ and $j = 0, 1, \ldots, N$, where $\Delta_x = (x_M - x_0)/M$ and $\Delta_y = (y_N - y_0)/N$. From the above definition, it is clear that a height field can be directly represented by a fractal interpolation surface of Section 3. The only issue to be determined is whether all of the height field data will be used in the construction of the surface or only a subset of them. This can be achieved by regularly sampling the height field along the $x$, $y$ dimensions. The sampling frequency defines a trade-off between quality of representation and compression ratio.
The proposed representation is expected to be especially fruitful for height fields defining natural surfaces, such as terrains. These often possess an intrinsic fractal structure which is conveniently described by fractal interpolation models. The idea of representing natural surfaces using fractal interpolation has also been suggested in Xie et al. [6], where the generation of rock fracture surfaces using fractal interpolation was examined.

5 Results

A height field of resolution $257 \times 257$ is presented in Figures 3(a) and 3(b). Specifically, the former figure depicts the height field as a 2D raster image where brighter areas indicate greater height; the latter figure contains its 3D depiction. This height field, which was created using Terragen Classic, contains a total of $257 \times 257 = 66049$ points.

Fig. 3. The original height field depicted (a) as a 2D image and (b) as a 3D surface.

Figures 4(a) and 4(b) show the 2D and 3D representation of this height field, respectively, using the proposed method. Specifically, it has been represented by a fractal interpolation surface constructed on a subset of the original data with $s = 0.02$; every 8th point along each dimension of the height field has been chosen as interpolation point. This results in a rectangular grid of resolution $33 \times 33$, containing 1089 points in total, i.e. about 1.65% of the original points. Despite the significant sparsity of the interpolation points, the quality of the reconstructed height field is satisfactory.

Another example is given in Figures 5(a) and 5(b), where the same height field has been represented by a fractal interpolation surface using even fewer interpolation points. Specifically, every 16th point along each dimension has been chosen as interpolation point. This results in a rectangular grid of resolution $17 \times 17$, containing 289 points in total, i.e. about 0.44% of the original points. Also in this case, the results are satisfactory despite the even smaller number of interpolation points. These results indicate that fractal interpolation surfaces are indeed capable of describing natural surfaces, such
Fig. 4. The reconstructed height field, using every 8th data point as interpolation point, depicted (a) as a 2D image and (b) as a 3D surface.

as terrains, with considerable quality even when high compression ratios are involved.

Fig. 5. The reconstructed height field, using every 16th data point as interpolation point, depicted (a) as a 2D image and (b) as a 3D surface.

Figures 6–7 depict an artistic rendering of the original height field as well as its two aforementioned reconstructions; these figures were created using Terragen™ Classic. As shown in the figures, the reconstructed height fields produce equivalent results compared to the original one, even though the significant sparsity of the interpolation points.

6 Conclusions and future work

We have presented a novel method for the representation and compression of height fields using fractal interpolation techniques. Specifically, we have represented a height field as a fractal interpolation surface constructed on the rectangular domain defined by a subset of the original data. The results indicate that the proposed methodology is feasible, while achieving satisfactory results in terms of quality of representation as well as compression ratios.
Further work will focus on the calculation of the optimal values of the vertical scaling factors in order to achieve increased localized accuracy, as well as on the exploration of alternative fractal interpolation surface models, affine or bivariate, including recurrent fractal interpolation surfaces.

References

The Baltic Sea coupled ice-ocean model

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Abstract: Community Earth System Model (CESM) is a coupled climate model that consists of five separate components with additional central coupler which controls time, exchanging data, domains, grids and other model data. Here CESM was adopted for the Baltic Sea and called 3D-CEMBS. This is not fully coupled configuration. In our case we have been taken into account ocean (POP model, version 2.1) and ice (CICE model, version 4.0) models which are forced by atmospheric data model (datm7). Other models are turned off in our configuration. However the main task of the datm7 is to interpolate data on model domain, we have interpolated data outside of the models. As an external forces daily-averaged forty-years reanalysis data derived from European Centre for Medium-range Weather Forecasts (ECMWF, ERA-40 reanalysis) has been used. Currently in pre-operational state 3D-CEMBS is using 48-hour atmospheric forcing data from ICM-UM model (University of Warsaw). 3D-CEMBS model has also ecosystem part (currently work in progress).

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Keywords: Baltic Sea, 3D model, hydrodynamic variables

1. Introduction

In 2011, the operational ecohydrodynamic model (3D CEMBSv2 - new version) was launched at the Institute of Oceanology PAS in the parallel version on the 2 km grid with rivers and the open boundary for the hydrodynamic module http://deep.iopan.gda.pl/CEMBaltic/new_lay/index.php. Therefore, the following works were completed:

- a new version of the model was launched (3D CEMBSv2), which includes the ocean-ice module POP-CICE (POP, version 2.1, the ice module CICE, version 4.0) and the ecosystem module (Fig.1):
  - compilation of the 2 km bathymetric grid,
  - adaptation of the model to meteorological data UM (ICM UW),
  - compilation of a data set with meteorological forecasts (UM) from 2010 and 2011 in the required format, preparation of the initial fields for the Baltic area in the 2 km grid,
- development of the operational system for data retrieval necessary for the model 3D CEMBSv2,
- development of the operational ecohydrodynamic model (3D CEMBSv2) in the parallel version on the 2 km grid with rivers and the open boundary:
  - compilation of climatic river discharge,
  - compilation of climatic conditions along the open boundary.

This paper presents an integrated, operational model of the Baltic ecosystem – the hydrodynamic part, the ocean-ice module POPCICE (version 2.1 and 4.0)
and the initial verification, the comparison of the obtained results with data from other models and measurements in situ.

Fig. 1. Schematic presentation of the Baltic model

2. 3D-CEMBS±2 MODEL

CCSM4.0/CESM1.0 (the Community Climate System Model/Community Earth System Model) consists of five separate elements with an additional coupler (CPL7), which monitors the time, exciting forces, domains, grids and exchange of information between models. The central part of the model is based on MCT (The Model Coupling Toolkit) – which is a parallel tool providing a number of services, such as a register of particular components of the model, descriptors of the domain distribution into processes, communication, redistribution of data and other very helpful tools. For our purposes, CESM was adapted for the Baltic Sea and was called 3D-CEMBS. However, it is not an entirely active configuration. The ocean model (POP, version 2.1) and the ice model (CICE, version 4.0) work in the active mode, and they are imposed by the model of atmospheric data (datm7). Other models are excluded from this configuration (the stub mode). The main task of datm7 is interpolation of atmospheric data into the domain of the model. External forces are daily-averaged for the period of forty years, which come from the ECMWF re-analysis (ERA-40). At present, in the operational mode, 48-hour atmospheric forecasts are used, which are supplied by the UM model of the Interdisciplinary Centre for Mathematical and Computational Modelling of the Warsaw University. 3D-CEMBS Model is also equipped with the ecosystem module, on which works are being currently carried out in order to incorporate it into the operational mode.

POP
The ocean model is based on the Parallel Ocean Program (POP, [7]) from the National Laboratory in Los Alamos (LANL), which is derived from the global ocean model ([6]) with additional conditions for free surface. This is a model of ‘z’ type (identical thickness of layers for every cell); the three-dimensional
equations describing the behaviour of the stratified ocean are solved by parametrization. Numerically the model defines spatial derivatives in the spherical coordinates using the method of finite elements. Physical quantities of the model are embedded in the spherical grid of Arakaw B ([1]). The barotropic equation is solved by ‘preconditioned conjugate gradient solver’ (PCG), and advection is determined by a centred differential. Parametrization of horizontal mixing is accomplished by a biharmonic operator, and vertical turbulence is determined by KPP parametrization. The equation of state, introduced by McDougall et al. ([5]), is also used.

POP is a three-dimensional hydrodynamic model derived from the ocean model created in the late 1960s by Kirk Bryan ([3]) and Michael Cox from the NOAA Geophysical Fluid Dynamics Laboratory in Princeton. The model was later modified and adapted by Semtner ([6]) for vector processors. The whole class of models from which POP is derived are of Brayan-Cox-Semtner type (B-C-S). The code of the model is adapted inter alia for supercomputers, but is well adaptable (porting) also for machines of different architecture, such as cluster types. The code of the POP model is characterised by good numerical performance and is well scalable on a large number of O(1000) processors. A special feature of the POP code (written in Fortran 90) is orientation to parallel machines and the use of MPI (Message Passing Interface) or SHMEM (Shared Memory) libraries to communicate between processors. And therefore, the most technologically advanced machines can be used to perform calculations and to solve major computational problems.

CICE
CICE (Community Ice CodE) is based on elastic-visco-plastic (EVP) rheology ([4]). It is designed to work in accordance with the POP ocean model using the parallel computing machines. It consists of several interactive elements: the thermodynamic model, which computes local growth rates of snow and ice owing to vertical conduction of energy and momentum fluxes. It also defines velocity of each ice cell based on wind and ocean velocity. It has a few vertical categories, so that the stress distribution is much closer to the real one. CICE was applied worldwide in different configurations. It was used to study the variability and the impact of ice on the processes occurring in the atmosphere and the ocean in time scales ranging from decades to hundreds of years ([2]). It is used in regional models and global applications. Incorporation of the CICE model (version 4.0) into the regional POP model (version 2.1), completed within the scope of the research grant (No NN305111636 – the Polish State Committee for Scientific Research), is the first application of such an advanced ice model for the Baltic Sea.

Ecosystem
The ecosystem model consists of 11 main components: zooplankton, small phytoplankton, diatoms, cyanobacteria, two types of detritus deposits (fraction of dissolved DOM and molecular POM detritus), dissolved oxygen, and components of nutrients, such as: nitrates, ammonia, phosphates and silicates. The class of small phytoplankton should reflect nano- and picophytoplankton, and can be limited by nitrates and phosphates. The class of larger phytoplankton
is represented by diatoms and can be limited by the aforementioned factors, as well as by silicates. The growth rate of cyanobacteria may depend on phosphates and the available light. Many biotic and detrital compartments contain multiple elemental pools as we track carbon, nitrogen, phosphorus and silicon through the ecosystem.

3D-CEMBS model is currently configured at two horizontal resolutions of 9 km and approximately 2 km (1/12° and 1/48° respectively). The model bathymetry is represented by 21 vertical levels and the thickness of the first four surface layers was five metres. The bottom topography is based on ETOPO1 1 arc-minute global relief model (http://www.ngdc.noaa.gov/mgg/global/global.html, National Geophysical Data Center). The bathymetric data were interpolated to the model grid using the kriging method. The initial state of the ocean model was prepared using temperature and salinity climatic data. The ocean surface level (5m deep) is restored based on the monthly timescale to the monthly average $T$ and $S$ climatology, as a correction term to the explicitly calculated fluxes and overlying atmosphere or sea ice. The restoring time was set to 30 days at the surface and 10 days at the domain boundary. 3D-CEMBS model domain is based on stereographic coordinates, but the equator of these coordinates is in the centre of the Baltic Sea (so we actually use rotated stereographic coordinates) and we can assume that shape of the cells is square and they are identical.

The current calculations are performed on supercomputers of cluster type, Galera, which is located at TASK (the Academic Computer Centre in Gdańsk). The time needed for computing 1 model year, for the ecohydrodynamic model, for resolution of 9 km, is 30 h on 16 processors, and for resolution 2 km – 120 h on 256 processors.

3. Results

In the first half of 2011, the new CESM model (Community Earth System Model - UCAR/NCAR USA) was adapted for the Baltic domain. Models have been adapted and work properly (see the website). The results of simulations for 48-hour forecast are presented for the model with resolution of 9 km and 2 km.

Select 2 km forecast for the area or the point (Fig. 2).
Fig. 2. Select 2km forecast for the area.

Simulations for 48-hour forecast for 2 km:
Forecast for the area:
• select the time of forecast (48-hour forecast) (Fig. 3a),
• select a required variable (temperature, salinity, currents, sea surface height, ice cover area) (Fig. 3b),
• select the depth, one of the ten layers for which you would like to see the model results (Fig. 3c),
• optionally you may provide boundaries to narrow the area of your query by selecting two random points (corners of a rectangle) on the map (Fig. 3d),
• to change coordinates x and y of selected points, press: Reset Coordinates (Fig. 3e),
• after all parameters are selected, press: Submit (Fig. 3f).
Fig. 3. Forecast for the area. Choose parameters of forecast for 05.03.2012.

The example is presented below for the following situations:
1. Forecast start: 2012.03.05, 18:00 UTC (Fig. 4)
Select forecast parameters: hour: +6 (forecast is for 2012.03.06, 00:00 UTC), variable: currents (cm s$^{-1}$); depth: 20-26 m (5th layer), coordinate X: 187 to 425, coordinate Y: 32 to 168. After selection of parameters, press: Submit and the screen will show a drawing with results of the model.
Fig. 4. Forecast for the area.

Forecast for the point:

- select the depth,
- select a point on the map to get 48 h time series,
- to change coordinates x and y of selected points, press: Reset Coordinates,
- after selection of all parameters, press: Submit

The example is presented for the following situation: forecast start: 2012.03.05, 12:00 UTC (Fig. 5).

4. Conclusion

The work presents the idea of a three-dimensional Coupled Ecosystem Model Baltic Sea version 2 (3D CEMBSv2) – a hydrodynamic part, which determines the main physical parameters of the environment: horizontal components of the velocity $u, v$, and the vertical component of velocity $w$, pressure $p$, density $\rho$, temperature $T$ and salinity $S$ of water. Variables presented on the website for 48-hour forecast are as follows: temperature, salinity, currents, sea surface height and ice area cover.

The 3D CEMBSv2 model (at present – the hydrodynamic module) is a suitable tool for studying the annual, seasonal, monthly and daily variability of environmental parameters in the southern Baltic Sea. It can therefore be applied in the forecasting of ecological changes in the Baltic.
Fig. 5. Forecast for the point.

References

Extraction of the Dominant Features of Complex Dynamics in Experimental Air-Water Two-Phase Flows

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Abstract: This paper aims at describing a novel approach to the analysis of the experimental void fraction time series detected from air-water upward two phase flows in a vertical channel. This system can express a great variety of different flow patterns, whose characterisation and classification strongly depends on the approach used for feature extraction. Phase space analysis in a traditional delayed embedding has allowed for the observation of the complex dynamics of the system. Nonetheless, the attractors obtained in a delayed embedding, though characterised by a regular complex structure, appear partly folded and are affected by noisy hydrodynamic high order dynamics. The present paper proposes the application of the SVD approach with the aim of assessing a more appropriate embedding into the phase space spanned by the principal vectors. In this way the dominant features of the system dynamics can be separated from noise-like dynamics, i.e. unfolded and noise-free attractors can be obtained.

Keywords: Feature extraction, Two-phase flows, Experimental nonlinear dynamics, SVD analysis.

1. Introduction

Several basic industrial processes, ranging from power generation, chemical and processing plants to oil pipelines, present heat and mass transfer applications of two phase flows. When two phase flows occur, very different flow patterns can be observed as well as transitions from a flow pattern to another. Indeed, the dynamical behaviours associated to the various types of flow pattern established in the system represent critical factors for the performances of such industrial systems. This explains the great efforts that have been and are still devoted to flow patterns identification, which represents a fundamental basis for appropriate characterisation of two phase flow systems.

The dynamics of two phase flows are typically of highly complex pulsating nature, under the effect of several nonlinearities deriving from the strong coupling of different mechanisms and of the dependence on various factors. Among the others, the most important factors are the differential action of gravity on the two phases and the effect of shear and surface tension forces at their interface. As a consequence, several different flow patterns can be identified, each of which can be characterised in terms of the dynamical behaviour of the void fraction time series.
Among the other, two phase flows of air-water mixtures are often theoretical and experimental analysed with the aim of achieving a reference perspective on the general dynamical behaviours, often valid also for more complex flows, such as those arising in presence of phase changes. In particular, the present study aims at analysing the behaviour of ascending air-water two phase flows in vertical pipes. For this kind of flows heat transfer phenomena connected to phase change are not involved, so that the flow pattern established in the system mainly depends on the mass flow rates of the two phases. By varying the mass flow rate of the two phases, in fact, bubbly, slug, churn and annular flows can be identified as the main flow patterns typical of several classifications [1-3].

The bubbly flow exists for low values of the gas mass flow rate and consists in the motion of dispersed and small gas bubbles in the liquid phase. Coalescence phenomena are at the basis of the transition from bubbly to slug flow, which can be observed by increasing the gas mass flow rate.Slug flow is characterised by gas bubbles, namely Taylor bubbles, enveloped by a liquid film separating them from the pipe walls, alternated to liquid slugs. In the class of slug flow, it is possible to distinguish between: cap flow, with short air bubbles (with the head approximately connected to the tail) separated by long liquid slugs; plug flow, with gas bubbles and liquid slugs of comparable length; proper slug flow, characterised by elongated gas bubbles separated from relatively short liquid slugs, often aerated for the presence of small dispersed air bubbles.

For growing gas mass flow rate, bubble coalescence and increasing aeration of the liquid slug leads to a highly unstable flow pattern addressed as churn flow, characterised by waves propagating through the liquid film enveloping the bubbles and occasionally falling within the tube, so to form a short, unstable and highly aerated liquid slug. Finally, the annular flow consists of a thin annular liquid film at the tube wall on which small ripples, interspersed occasionally with large disturbance waves, flow in a regular manner up the tube.

It is usual practice to perform flow pattern identification on the basis of the differences of the dynamical behaviour of the time series of the local void fraction. Therefore, the reliability of the identification approach is highly dependent on the accuracy of the technique adopted to measure the void fraction. Several techniques have been proposed [4-9] and impedance measurements seem to be recognized as the most reliable [6]. At the same time, the performances of flow pattern identification approaches depend also on the techniques adopted for time series analysis and feature extraction. Statistical [1, 2, 6] or spectral [9-12] techniques indeed represent the typical approach for flow patterns identification on the basis of the analysis of the experimental void fraction time series. Nonlinear techniques have been also adopted, among the others see [10, 13-16], but a main drawback has been represented by the relatively poor spatial and temporal resolution of the experimental time series.

In order to address this problem, the experimental time series considered in the present study have been detected by means of a resistive probe characterised by high temporal and spatial resolution, which has been appositely set-up as described in [17]. The preliminary analysis in a delayed embedding of the void
fraction time series detected by means of this sensor has shown the existence of strange attractors of interesting morphology for the various flow patterns [18]. Nonetheless, attractors obtained in this way are somewhat noisy as a consequence of the superposition of high order dynamics to the dominant dynamics characterizing the flow pattern. Among the others, the most important high order “noisy” dynamics are those of hydrodynamic nature associated to small diameter bubbles dispersed in the liquid slugs and to disturbances on the liquid film enveloping the Taylor bubbles.

Therefore, the present study aims at extracting the dominant features of the flow dynamics under various flow pattern conditions so to separate the dominant features of the system dynamics from noise-like dynamics. The proposed approach is analogous to that proposed in [19] and is based on the calculation of the singular vectors of a n-dimensional delayed embedding, through the application of the technique known as Singular Value Decomposition (SVD) [20], and in the analysis of the restricted portion of the dynamics that is obtained by projecting the attractor onto the phase space spanned by the singular vectors corresponding to the three highest singular values.

Reported results, show that the attractors described in the new embedding present a well defined and regular structure, indicating the existence of a low order source of the system dynamics, which will be analysed in future studies.

![Fig.2 Experimental apparatus](image)

**2. Experimental Apparatus**

The experimental apparatus reported in Fig. 1 has been built and tested in order to study the dynamics of two-phase flow in vertical pipes. The test section is a vertical pipe of diameter 0.026 m diameter and length 3 m. The apparatus is equipped by an electromagnetic flowmeter and three air flow metres,
respectively used for the measure of the water velocity and mass flow rate and for the regulation of the air flow rate in the range between 10 and 210 l/min. The air is supplied to the mixing section by a pressurised tank fed by a compressor, whereas the water flow rate can be varied in the range 0-150 l/min by means of a series of valves and bypasses placed at the pump outlet.

A resistive probe for the measure of the void fraction is placed at a distance of over 100 times the diameter of the pipe from the mixing section, i.e. over the required entry region for two phase flows, in order to ensure a well established flow regime. In particular, the void fraction probe has been designed and realised for the experimental campaign and operates in the resistive range (carrier frequency of 20 kHz). The sampling frequency was set at 1 kHz with a cut-off frequency of 200 Hz. A detailed description of the experimental probe and on the wide set of experimental tests performed is reported in [17].

3. Dynamical Feature Extraction

The results of preliminary linear analyses of the experimental time series have been shown to be unable to deal with the intrinsic complexity of two phase flows dynamics. Hence, in [18] a morphological analysis of the three-dimensional attractors has been proposed in a classical Takens’ delayed embedding of the experimental void fraction time series [21]. In particular, it has been observed that the attractors obtained for some of the flow patterns are characterised by a regular fractal structure, which is indeed one of the most important evidences of deterministic chaotic behaviour.

In the present study, the aim is to improve the dynamical representation by adopting a new embedding, derived through the application of Singular Value Decomposition technique, SVD [20], to the classical delayed embedding based on Takens’ theorem, similarly to the approach proposed in [19]. The new representation is characterised by a drastic reduction of noisy dynamics and, above all, a sensitive improvement of the attractor unfolding, so that the dominant morphological characteristic can be fully exploited.

As a first step, the phase space reconstruction consists in the creation of a \( n \times w \) matrix, \( S \), where \( n \) is the length of a window moving through the data and \( w \) are the independent variables defining the phase space, i.e. delayed version of the experimental void fraction time series \( s(t) = (s_0, s_1, s_2, \ldots, s_i, \ldots) \), with each column delayed \( \tau \) time steps from the previous. The condition \( w > 2d + 1 \) for an appropriate embedding is implicitly respected if \( w \) is set much greater than the unknown fractal dimension \( d \) on the basis of a preliminary estimation.

The second step consists in the application of the SVD approach to matrix \( S \). This is done through the calculation of a new diagonal matrix, equivalent to the original one, i.e. with identical singular values but in decreasing order. In particular, \( S \) is factorized into its singular values according to equation:

\[
A = M^T S C
\]  

(1)

In (1) \( A \) is the diagonal matrix containing the singular values of \( S \) in decreasing order and \( M \) and \( C \) are the matrices of the singular vectors associated with \( A \).
Details on the factorization can be found in [20]; what is interesting for the scopes of the present study is that the high level singular values in $\Lambda$ are associated to the dominant singular vectors, i.e. those representing the dominant features of the system dynamics, whereas the low level ones correspond to local behaviors or noise-like components. Therefore, the system can be virtually partitioned into two subsystem: the first deriving from noise free data (i.e. the main features and the relevant details) and the second from noisy dynamical behaviours, which can be considered superimposed and then eliminated.

4. Results and Discussion

The described approach has been used in the present study in order to obtain a denoised and unfolded representation of the experimental dynamics. The SVD technique has been applied to the delayed embedding $S$ of the experimental void fraction time series, created considering $\tau=1$ and $w=40$ in order to ensure that $w$ is sufficiently greater than $m$, i.e. greater than the (unknown) system dimension. The length $n$ of the observation window has been set at 10000 data samples in order to be wide enough to obtain a well defined attractor in phase space, i.e. an attractor whose morphology does not change if further data samples are added. The claimed advantages of the proposed methodology can be observed in the results reported in the Fig. 3 to 8, which report the attractors of the same operating condition in two different embeddings. In particular, the phase space adopted for the plots on the left hand side of each figure is the basic three dimensional Takens’ delayed embedding, whereas the projections on the pseudo-phase space spanned by the three dominant principal vectors of the improved embedding obtained through application of SVD are those reported on the right hand side of each figure. By comparing the two methods of representation it is possible to observe that the attractors in the delayed phase space are in all cases sensibly affected by a higher noise level and are not sufficiently unfolded with respect to the corresponding attractors in the principal component embeddings, the last being characterised by a very low level of noise and a satisfactory unfolding. It is worth to remind that, even if the two attractors of each flow pattern appear different, they are, nonetheless, expressions of the same dynamical behaviour. In fact, they are morphologically equivalent and, therefore, characterised by the same invariants of the dynamics, such as fractal dimension and Lyapunov exponents [22-25].

The successful unfolding contributes to the achievement of a clear and well defined morphology of the attractors. This is a main advantage for the distinction of different flow patterns through a comparison of the representation of their dynamics in the phase space spanned by the principal components. Moreover, in some cases the proposed embedding amplifies important characteristics of the system dynamics. For example, the right hand cap flow attractor in Fig. 4 shows a clear distribution of the trajectory in alternated bands, which is a hint of the fractal (i.e. chaotic) nature of the system dynamics. Finally, the representation in the principal component phase space is very effective in underlining the differences between the various flow patterns.
Each type of flow pattern is, in fact, characterised by a specific morphology, sufficiently different from that of the other flow patterns. In particular:
- each flow pattern attractor occupies a different phase space region;
- each attractor differently “fills” its own region of phase space; for example, the cap flow attractor (properly 3-D) has a higher filling rate than that of the plug flow (which moves around a sort of 2-D limit cycle);
- the attracting region is progressively shifted, with a continuous trend from bubbly to annular flow, with respect to the first principal component.

These differences are very important as the morphological considerations drawn insofar are related to the fractal nature and to the stretch and folding behaviour of the attractors [24], which can be considered as the topological expressions of the mentioned invariants of the dynamics, whose calculation is behind the scope of the present study and will be the object of future studies.

5. Conclusions
This study proposes a phase space approach for the description of typical complex dynamics of two-phase flow. At first the singular vectors of the classical delayed embedding are calculated and the attractors of the system dynamics are projected on the state space spanned by these eigenvectors. In this way the dominant feature of the dynamics, corresponding to a subset of the highest singular values, are separated from noisy dynamics in the time series, corresponding to the remaining lower singular values. The morphology of the attractors in the obtained unfolded and noise-free representation are analysed. Reported results demonstrate that the proposed approach represent a powerful tool for the identification of two-phase flow patterns.

Fig.6 Attractors in the delayed and principal component embeddings for the slug flow; air flow rate 40 lit/min - water flow rate 16.80 lit/min.

Fig.7 Attractors in the delayed and principal component embeddings for the churn flow; air flow rate 80 lit/min - water flow rate 9.01 lit/min.
Fig. 8: Attractors in the delayed and principal component embeddings for the annular flow; air flow rate 80 l/min - water flow rate 5.58 l/min.

References
