

Geni Process

The GenI proces[1] ([dʒiːnaɪ] for Generic Intelligence) describes a time-discrete stochastic process $X : [0, 1] \times \mathbb{N}_0 \rightarrow 2^E$ in the state space of the finite subsets of a countable set "E", together with a mapping $2^E \rightarrow \mathbb{C}^n$ of the power set on E into an n-dimensional complex vector space. In principle, it can be classified as a Markov chain of first order, with variable transition probabilities $P(X_{n+1}|X_n)$. There are two expressions of the GenI process, for $n \geq 2$ and for $n = 2$.

GenI Process on Eigenvector Swarms

Background

The GenI random process determines changes in the complex vector space from the random behaviour of independent individuals within a swarm-like construct. The swarm has a superposition state

$\sum_{j=1}^n \beta_j e_j \in \mathbb{C}^n$, that controls the individual activities via a target function. The amplitudes $\beta_j e_j$ are also called ideas (cf "Generalized Quantum Modeling" [2]). The swarm takes one of the eigenstates γe_j after a finite number of steps, with the well-defined probability $\frac{|\beta_j|^2}{\sum_k |\beta_k|^2}$. The individuals follow defined rules

and are allowed to make mistakes, based on the processes in simulated shoals of fish[3]. The GenI algorithm starts a chaotic decision-making process as a competition of ideas, such as running in a team that has to choose from a limited number of solutions to a given task. In the course of the process, a selection mechanism leads to ideas becoming extinct one after the other until finally just one survives that represents the solution to the problem. The special properties of the GenI process also make it interesting for the interpretation of physical processes.

Definition

Terminology

Let E be a countable set and $\tilde{E} := \{S \subset E : |S| < \infty\} \subset 2^E$ the set of finite subsets of E . Next $B = (e_1, \dots, e_n)$ the canonical basis in \mathbb{C}^n and $\tilde{B} := \{i^k e_j : k = 0 \dots 3, j = 1 \dots n\}$, where i is the imaginary unit in \mathbb{C} . A given $\rho : E \rightarrow \tilde{B}$ maps each element of E into \tilde{B} , so that

$\forall e \in \tilde{B} : |\rho^{-1}(\{e\})| = \infty$. For a set $S \in \tilde{E}$ the complex vector $\rho(S) := \sum_{s \in S} \rho(s) = \sum_j \beta_j e_j$ denotes its state with complex

amplitudes β_j . Each such set S is called an Eigenvector-swarm or **E-swarm**. A pair $s, t \in E$ with $\rho(s) + \rho(t) = 0$ is a **null pair**. A tuple (s_0, s_1, s_2, s_3) is called a **null ring** generated by s_0 , if $\exists j : \rho(s_k) = i^k e_j$. A set $N \in \tilde{E}$ is called a **null set**, if $\rho(N) = 0$. A maximal null set $N \subset S$ is called the **entropy** of S and $S \setminus N$ the **entropy free residual swarm**. The term

$\epsilon_j(S) := 2 \frac{|\beta_j| \sqrt{\sum_{k \neq j} |\beta_k|^2}}{\sum_{k=1}^n |\beta_k|^2} \in [0; 1]$ denotes the **excitation** of the swarm in index j .

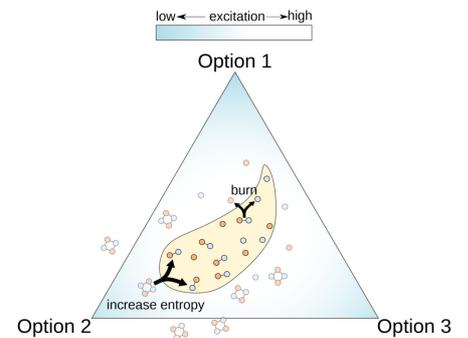


Figure 1: The GenI Process: A GenI swarm constantly sucks null rings from its environment and burns them randomly. The GenI process sets a gradient towards reducing the excitation. The swarm's state finally arrives at one of the given options.

Algorithm

Let $S^{(l)} = S_D^{(l)} + N_S^{(l)}$ be a series of swarms (as an instantiation of $X(\omega, l)$) with the respective separation into a maximal null swarm $N_S^{(l)}$ and the entropy free residual swarm $S_D^{(l)}$, $\rho(S^{(l)}) = \sum_{k=1}^n \beta_k^{(l)} e_k$ the respective state and $\epsilon_j^{(l)} := \epsilon_j(S^{(l)})$ the excitations.

1. **Step:** Set $l \leftarrow 0$ and start with a given swarm $S^{(0)}$.
2. **Step:** If $\forall j : \epsilon_j^{(l)} = 0$, then finish the process.
3. **Step:** Each element $s \in S_D^{(l)}$ generates an additional null ring within the swarm.
4. **Step:** Each null pair $r, t \in N_S^{(l)}$ (including the newly generated) with $\rho(r) = i^k e_j$, $\rho(t) = -i^k e_j$ gets selected with probability $p = \epsilon_j^{(l)2}$ (and "burned" in the next step).
5. **Step:** For each selected null pair $r, t \in N_S^{(l)}$, t leaves the swarm with probability $\frac{\epsilon_j(S \setminus \{r\})}{\epsilon_j(S \setminus \{r\}) + \epsilon_j(S \setminus \{t\})}$. Otherwise t stays and r leaves the swarm.
6. **Step:** The resulting swarm is $S^{(l+1)}$.
7. **Step:** Set $l \leftarrow l + 1$ and start over with step 2.

Interpretation

As soon as the excitation disappears in each index, the process naturally comes to rest in step 4 (except for the hard abort condition in step 2), since no null pair is "burned" and the state of the swarm no longer changes. The role of the excitation here reminds of the dynamics of a grain of sand in the formation of the Chladnic sound figures. On the other hand, excitation as a target in step 5 leads to a systematic distortion from 50% likelihood for an individual to remain. This leads here to an improved tendency to reduce the excitation. The following interpretation is obvious based on biological swarm behaviour: Each individual tends to follow the rule "reduce the excitation". It remains free in its decision to do nothing (step 4), to follow the rule, or to disregard it (step 5).

Simulation

The reference implementation under JAVA[4] shows an extremely good convergence of the process. The table shows an example of the result of 1000 simulation runs (each simulation run aborts after more than 500 iterations or for swarm sizes > 10 million for performance reasons):

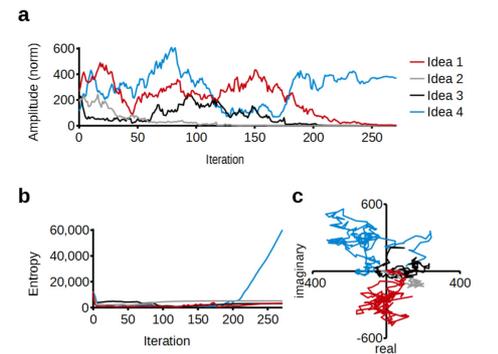


Figure 2: Competition of ideas within an GenI swarm: Graph a shows the evolution of absolute amplitudes during a GenI process operating in a four options environment. Due to its intrinsically chaotic behaviour, it is impossible to predict the GenI process evolution at any point. Interestingly, here the option with the lowest initial chance finally wins. Chart b shows the according evolution of entropy rising dramatically at the end for the winning idea. Figure c displays the native paths of each idea in the complex plane.

shoals of fish. The GenI algorithm starts a chaotic decision-making process as a competition of ideas, such as running in a team that has two possible solutions to a given task. In the course of the process, a selection mechanism leads to the survival of only one of the two ideas that represents the solution to the problem. The model in principal allows a moving environment. The special properties of the P-process also allow interpretations of physical processes (see in particular Carl Friedrich von Weizsäcker's ur-alternatives (archetypal objects)[5], which he outlined for the reconstruction of quantum mechanics).

Definition

Terminology

Let E be a countable set and $\tilde{E} := \{S \subset E : |S| < \infty\} \subset 2^E$ the set of finite subsets of E . Next $P = (p_0, p_1, p_2, p_3) \subset \mathbb{C}^{2 \times 2}$ the Pauli matrices and $\tilde{P} := \{i^k p_j : k = 0 \dots 3, j = 1 \dots n\}$ the image of the **Pauli group** in the complex matrix algebra as its irreducible representation. Such a subset $S \in \tilde{E}$ is called a **Pauli-Swarm** or a **P-Swarm**.

A given $\rho : E \rightarrow \tilde{P}$ maps each element of E onto one element of the Pauli group, so that $\forall e \in \tilde{P} : |\rho^{-1}(\{e\})| = \infty$.

A basis $a_1, a_2 \in \mathbb{C}^2$ is called an **environment**, a non zero vector $v \in \mathbb{C}^2$ a **perspective**. For a swarm $S \in \tilde{E}$ denotes $\rho(S) := \sum_{s \in S} \rho(s)$ its matrix image, $\tilde{S} = \rho(S)v := \beta_1 a_1 + \beta_2 a_2$ its **state** with complex

amplitudes β_j at the given environment and perspective. A pair $s, t \in E$ with $\rho(s) + \rho(t) = 0$ is a **null pair**. A tuple (s_0, s_1, s_2, s_3) is called a **null ring** generated by s_0 , if $\exists j : \rho(s_k) = i^k p_j$. A set $N \in \tilde{E}$ is called **null set**, if $\rho(N) = 0$. A maximal null set $N \subset S$ ist called the **entropy** of S and $S \setminus N$ its **entropy freed residual swarm**. The term $\epsilon(S) := 2 \frac{|\beta_1| |\beta_2|}{|\beta_1|^2 + |\beta_2|^2} \in [0; 1]$ denotes the **excitation** of the swarm.

Algorithm

Let $S^{(l)} = S_D^{(l)} + N_S^{(l)}$ be a series of swarms (as an instance von $X(\omega, l)$) with the respective separation in a maximal null swarm $N_S^{(l)}$ and the entropy freed residual swarm $S_D^{(l)}$, $\tilde{S}^{(l)} = \beta_1^{(l)} a_1 + \beta_2^{(l)} a_2$ the respective states and $\epsilon^{(l)} := \epsilon(S^{(l)})$ the excitations.

1. **Step:** Set $l \leftarrow 0$ and start with a given swarm $S^{(0)}$.
2. **Step:** If $\epsilon^{(l)} = 0$, then finish the process.
3. **Step:** Each element $s \in S_D^{(l)}$ generates an additional null ring within the swarm.
4. **Step:** Each null pair $r, t \in N_S^{(l)}$ (including the newly generated) gets selected with probability $p = \epsilon^{(l)2}$ (and gets "burned" in next step).
5. **Step:** For each selected null pair $r, t \in N_S^{(l)}$, t leaves the swarm with probability $\frac{\epsilon(S \setminus \{r\})}{\epsilon(S \setminus \{r\}) + \epsilon(S \setminus \{t\})}$. Otherwise t stays and r leaves the swarm.
6. **Step:** The resulting swarm is $S^{(l+1)}$.

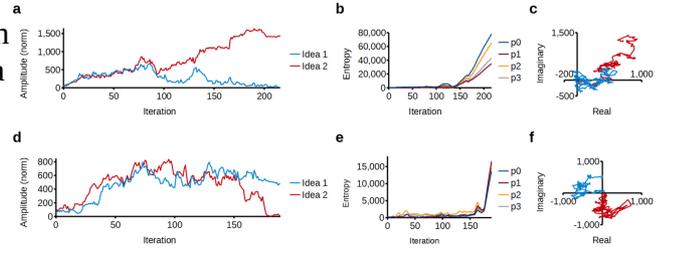


Figure 3: Competition within a GenI-swarm: The diagrams a-c demonstrate the Pauli process (as a special expression of the GenI process) evolution using a fixed perspective $p = (1; 1)$ under an external environment defined by the observable p_3 . Diagrams d-f show another test case under the internal environment defined by the swarm itself. Graphs a/d show the evolution of absolute amplitudes. Charts b/e show the according evolution of entropy for each type according to the swarm member images in $\{p_0, \dots, p_3\}$. Null pairs for P-swarm do not relate to environment options as is true for E-swarm. Figures c/f display the native paths of each idea in the complex plane.

7. **Step:** Set $l \leftarrow l + 1$ and start over with step 2.

Simulation

The reference implementation under JAVA[4] shows an excellent convergence of the process at any fixed environment and perspective according to the chart right hand.

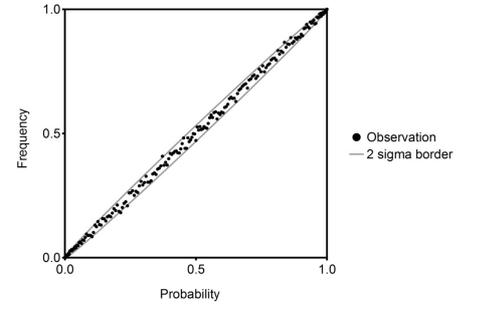


Figure 4: GenI swarms observations along target values from 0 to 1: The sample comprises 1000 measurements each on 201 test points on P(auli)-swarms (a special expression of a Geni swarm). More than 97% of observed frequencies are in the 2 sigma interval around the target values expected from quantum measurements. The chi square test value 92.6 is much lower than the critical level 168 at 95% confidence and 200 degrees of freedom.

The results support the **statement of convergence (hypothesis):**

Let $S^{(0)} = S$ be a given P-swarm with $\tilde{S} = \beta_1 a_1 + \beta_2 a_2$, $b_j = |\beta_j|$, $|S| = \sqrt{b_1^2 + b_2^2}$, at any fixed environment (a_1, a_2) and perspective v .

Let $S : \mathbb{N}_0 \times [0; 1] \rightarrow \tilde{E}$ be a P-process with $\tilde{S}^{(m)}(\omega) = \beta_1^{(m)}(\omega) a_1 + \beta_2^{(m)}(\omega) a_2$.

Then $P\left(\tilde{S}^{(m)} \xrightarrow{m \rightarrow \infty} \gamma a_j\right) = P\left(\sum_{k \neq j} b_k^{(m)2} \xrightarrow{m \rightarrow \infty} 0\right) = \frac{b_j^2}{|S|^2}$, where

$j, k \in \{1, 2\}$

References

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- [3] I. D. COUZIN, J. KRAUSE, R. JAMES, G. D. RUXTON, and N. R. FRANKS, 'Collective Memory and Spatial Sorting in Animal Groups', *J. Theor. Biol.*, vol. 218, no. 1, pp. 1–11, Sep. 2002.
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