

# Pfeiffer – A Distributed Open-ended Evolutionary System

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## Abstract

Pfeiffer contains a population of fractals which has been evolving continuously for more than three years. The animations are developed from embryos using a Lindenmayer grammar (L-System). These open generative representations potentially allow gene duplication and the evolution of higher order genetic operators and might be a step towards the emergence of social intelligence in swarms of artificial life (alife) agents. The fitness function is simply do the snowflake patterns appeal to the users: interactive evolution (IEC). To this end, images are placed in animated snow globes (computerised snowstorms) by world wide web (www) browsers (Netscape, Mozilla, Internet Explorer, Firefox, etc.) anywhere on the planet. More than 600 people have used <http://www.cs.ucl.ac.uk/staff/W.Langdon/pfeiffer.html>.

## 1 Introduction

For more than three years we have been running an experiment in distributed open-ended interactive evolution in which small local populations within each user's web browser communicate via Javascript with a central server holding a variable sized global population (see Figure 2). (Initial results were reported in Langdon (2004a).) Pfeiffer is intended to show the feasibility of evolving agents on many small computers running across the Internet under the user's actions as a fitness measure. The agents are intended to be attractive and therefore they are animated in a snowstorm. Their form is given by a D0L deterministic context free L-system Prusinkiewicz and Lindenmayer (1990) (see Figure 1), whose initial seed is a Koch fractal snowflake.

L-systems have the advantage over traditional programming in that they are inherently parallel. This is analogous to growing plant tissue (for which they first used to model) where each cell grows and divides in parallel with its neighbours and like DNA strands where, in principle, all genes can be expressed simultaneously. Karl Sims was perhaps the first person to combine L-systems with interactive evolution, e.g. Sims (1991).

The next section describe the evolutionary L-system. Section 3 summarises its usage (more details are given in Langdon (2004a) and Langdon (2004b)) while section 4 considers what lessons can be drawn. The penultimate section (5) discusses where evolutionary agents might lead us. We conclude, in Section 6.

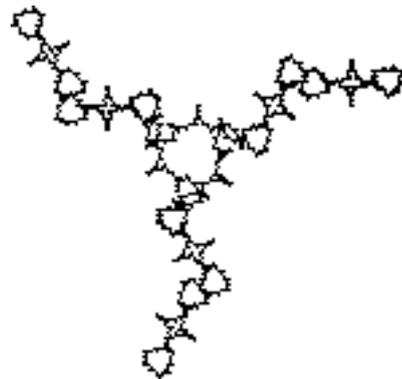


Figure 1: Example L-system fractal pattern. The picture is the phenotype developed from the 435<sup>th</sup> genotype (seed) saved by users in the global population. The seed defines the L-system's initial grammar symbol as  $F++F++F++F++F$  and the replacement rule as  $F \Rightarrow FF+FF--F$ . It also specifies that start symbol be expanded four times.

## 2 How Pfeiffer Works

Pfeiffer (cf. Figures 2 and 3) evolves agents and displays them moving in two dimensions across the screen of a world wide web (www) browser. The visual phenotype of each agent is given by a Lindenmayer (L-system) grammar. As the agents are moved or tumble across the screen they are subject to random encounters and changes which may effect their grammar. Each time the grammar is changed the agent's new shape is drawn on the screen. The user can save pretty shapes and delete ugly ones.

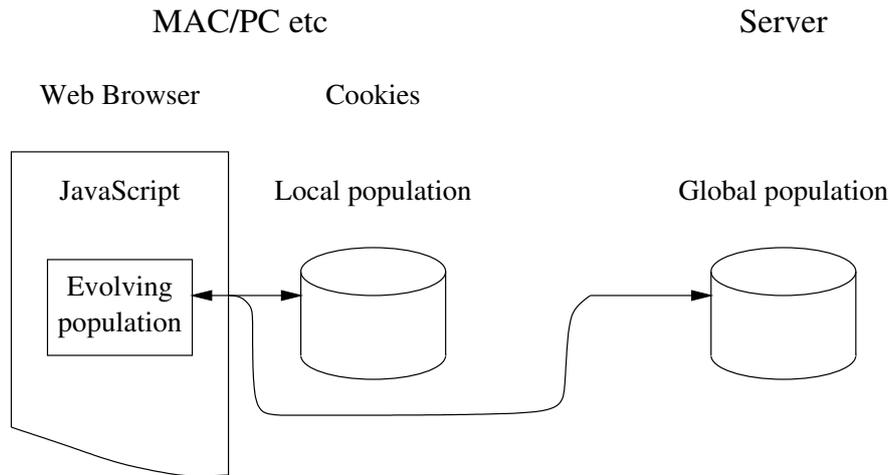


Figure 2: Overview of Pfeiffer. The user interacts via HTML and Javascript run in their web browser (left hand side). Initial seeds (chromosomes) are either retrieved from earlier runs via cookies or down loaded across the Internet via cgi programs from the global population. After evolution in the user's browser, the user may save new seeds both locally (as cookies) and in the global population.

## 2.1 Global and Local Populations

The active local population is stored in Javascript objects within the user's web browser. However these are lost when a new HTML page is selected. Therefore, "cookies" (if they are enabled) are used to provide off line storage of the local population.

Each time the Pfeiffer web page is down loaded, the initial value of each agent's chromosome is read from the corresponding cookie. However, if there is no cookie, the initial chromosome is down loaded from the global population across the network.

## 2.2 User Interaction

The primary goal of the user intervention is to use the user to provide selection pressure to drive the evolution of the agents. Passing the mouse over an agent causes its menu to be displayed. A text field allows the user to name the agent. While the pull down menu (see Figure 4) confirms the agent's identity and allows the user to: save the agent, make a copy of it (both automatically give it high fitness), delete it and close the menu. Naming an agent makes it easier for the user to track the agent he has evolved using "top ten" and "Hall of Fame" web pages. An agent "saved" by the user is stored in its cookie and appended to the global population. Once in the global population, the agent can be down loaded by other users and so distributed across the world. Cloning an agent causes an additional copy of the agent to be stored in the local population. This will often require the deletion of an-

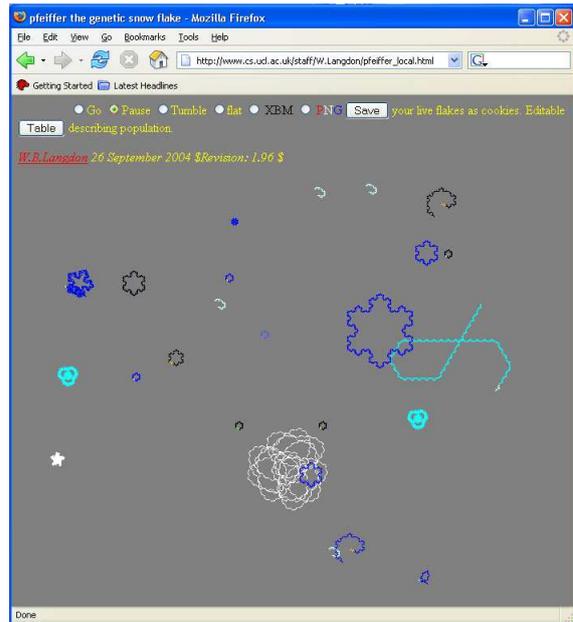


Figure 3: View of evolutionary arena as seen by user

other, low fitness, agent. These user initiated actions exert selection pressure on the local and global populations.

In addition to deciding life and death, the user can influence which agents mate. Using the mouse, an agent can be picked up and moved into the path of another agent. As with saving and cloning, moving an agent implies the user prefers it and it is given high

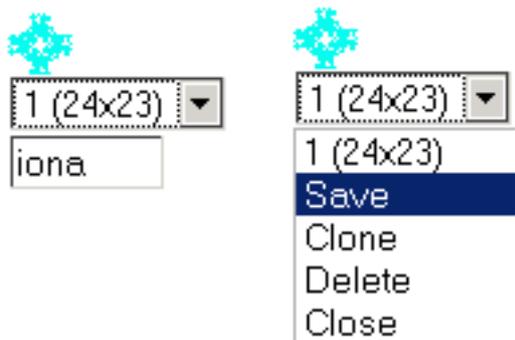


Figure 4: Example menu. Left hand, allows user to change agent's name. Right, pull down menu, allows user to save, copy or delete agent.

fitness, making it very likely to mate with the next mature agent it meets.

### 2.3 Generating the Phenotype

The system is able to display the results of arbitrary L-systems. In the original system (and even today in some browser) this is beyond Javascript. Therefore it was necessary to generate the graphics on a server and download them into the user's browser (see Figure 5). In this mode of operation, each new seed is passed to the server. It is interpreted as a Lindenmayer grammar. This generates a series of drawing instructions, which are immediately obeyed. The resulting picture is compressed and converted to .GIF format and passed back to the user's browser for display. Because of the data compression, this takes only a few seconds. However the delay could cause problems due to the agent's genotype and phenotype becoming out of step Langdon (2004a). Therefore the new version of Pfeiffer processes L-systems and graphics generation in the user's browser. However both systems are active (for compatibility with less able browsers).

### 2.4 Genetic Representation

Each agent seed is a variable length linear text string. The default seed grows into the Koch snowflake . The default seed is the 56 character string `v=60&str=F++F++F & it=2 & sc=5 & rules=('F', 'F-F++F-F')` (this can be replaced by the user).

The string is split by & characters into parameters. They are processed left to right. Thus if any parameter is repeated, the second "gene" is "dominant".

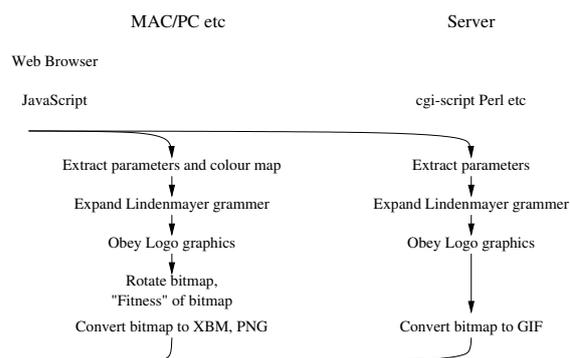


Figure 5: Mapping genotype to phenotype. In this development process the genotype (the L-system plus associated parameters) is converted to a graphic. The chromosome may either be passed to our server, interpreted and a .GIF file returned (right hand side) or interpreted locally (left). The local version avoids Internet delays, allows colour, and 3-D effects but is less portable.

Five parameters are recognised. They are `v` (angle), `str` (start string of grammar), `it` (depth of recursive expansion), `sc` (side, in units of 0.2 pixels) and `rules` (grammar replacement rules). Each substring formed by splitting the seed at the & is further split by =. If the first part of the substring exactly matches one of the parameter names then its value is set to the text between the first and second (if any) =. If a parameter is missing, its default is used. The defaults come from the Koch snowflake, they are `v=60`, `str=F++F++F`, `it=2`, `sc=5` and `rules=('F', 'F-F++F-F')`. When `rules` is parsed characters such as ( and ) are removed. In our Koch example this means the single substitution rule is:  $F \Rightarrow 'F-F++F-F'$ . The use of the defaults is effectively the same as if the default text were inserted at the start of every seed (albeit protected from genetic operators).

Once parameters have been decoded the L-system is interpreted. First the start string `str` is expanded `it` times. At each step every character which matches the left hand symbol of a rule is replaced by the corresponding right hand side. Note any letter can potentially match a rule, not just those used by the turtle graphics, allowing indirect rules. The expansion yields a potentially very long string. To avoid infinite or very long recursions, time outs are applied.

The string is interpreted as a series of "turtle" drawing instructions. Except for 3-D instructions, predefined graphics and increasing the line width, all of the turtle instructions given in Pruskinkiewicz and Lindenmayer (1990) are supported. The graphic is



First parent

```
v =1-72&strF+'+F+4+F+F&&st F+2& 'c s |5tetulF+ =-F Fe , - F&F(Fl+&=
```

Second parent

```
v =1=72&&strF ' +F+4+F+F&&st F+2& 'c s |5tetulF+ =-F Fe , - F&F(Fl++=
```

Offspring, replaces first parent

```
v =1-72&strF+'+F+4+F+F&&st F+2& 'c s |5tetulF+ =-F Fe , - F&F(Fl+&=
```

Figure 6: Example crossover. Length of first parent 67, first cut point at 37, remove 10 characters, insert 11 characters. 68 characters in offspring.



Figure 7: Usage of Pfeiffer up to April 2004. Red lines connect each user's country to the central server. Heaviest use has been from UK, USA and Canada, but users have also come from the far and middle east, India, Europe, Latin American and South Africa.

## 5 Future: Breeding “Intelligent” agents

Our agents are very limited. We feel they need to be able to evolve to react to their environment. They need to be able to evolve to predict their environment. Of course this makes requirements of both the agent and the environment. Also, perhaps crucially, each agent needs to be able to effect the environment, and predict what those effects will do for it (and for others). While L-systems have been mainly used (as we have done here) to create static structures, they can describe networks. Those networks could contain sensory, processing and active elements Hornby

and Pollack (2002) and/or use cultural evolution– simulation, imitation and knowledge-based operators–such as used by the vehicles of Gabora (1995). Gruau (1994) describes another indirect approach to evolving artificial neural networks (ANNs). While Stanley and Miikkulainen (2003) surveys developmental evolution in computer science.

There is a strand of thought in which intelligence came from a co-evolutionary struggle between members of the same species Ridley (1993). If true, can intelligence arise in isolated agents? Or are interacting/communicating agents needed?

A problem with simulated worlds has been hosting sufficient complexity so as to be challenging but still

allowing agents to be able make predictions about what will happen next and what will happen to me or to others if I do this. The Internet hosts tides of data. This data is not random. It ought to be possible to harness it to give a suitable virtual environment.

We have fallen for the usual trap of constructing a two dimensional world (on the computer screen). However is there any hope of evolving artificial life (and thereby artificial intelligence) in two dimensions? Obviously three dimensions are sufficient but computer simulations offer many dimensions ( $N \gg 3$ ).

## 6 Conclusions

Lindenmayer grammars can be used as the basis for a distributed interactive evolutionary system and produce interesting fractal like patterns. Many new patterns have been evolved Langdon (2004b), some exploiting the L-system to produce some regularities and re-use of motifs. It is feasible to represent individual agent's genetic material (seed/chromosome) with a variable length text string without defined fixed semantic fields and using crossover at the character level. The representation allows a huge degree of redundancy to evolve. The "fitness landscape" clearly contains a huge degree of "neutrality" and evolution is using it. This loose representation allows the location etc. (as well as the meaning) of the L-system to evolve. Gene duplication, translocation and other genetic operations could be supported by such a general representation.

In terms of harvesting spare CPU cycles, the project confirms it can be done using Javascript and user's web browser. The project does hint at some successes. World wide distributed evolution is clearly feasible. Perhaps more importantly one can recruit users (which are much more valuable than their CPUs) to assist in guided evolution. Finally animated tools are an attractive way to spread interest in artificial evolution, intelligence and life.

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