



# Adversarial Evolution and Deep Learning for Computational Creativity

Alan Blair, UNSW

(joint work with Jacob Soderlund and Darwin Vickers)

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## Outline

- ▶ Adversarial Coevolution in Nature
- ▶ Adversarial Coevolution in Computation
- ▶ Artist-Critic Co-Evolution
- ▶ Interactive Evolution (GP Artist; Human Critic)
- ▶ Generative Adversarial Networks (CNN Artist; CNN Critic)
- ▶ Evolutionary Art (GP Artist; GP, NN or CNN Critic)
- ▶ Artistic Techniques and Styles
- ▶ Discussion/Conclusion

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## Artwork of Hercule LeNet



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## Adversarial Coevolution in Nature

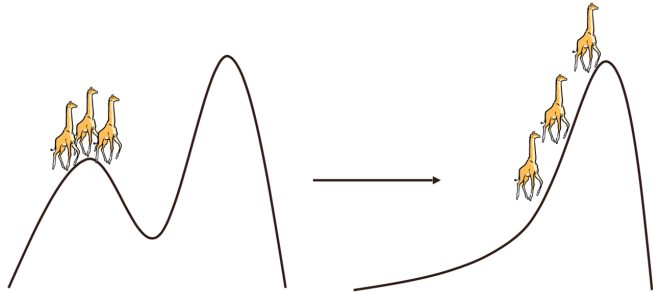


- ▶ Gazelle adapts to run faster and escape from the Leopard
- ▶ Leopard adapts to run faster and catch the Gazelle

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## Punctuated Equilibria (Eldredge & Gould, 1972)

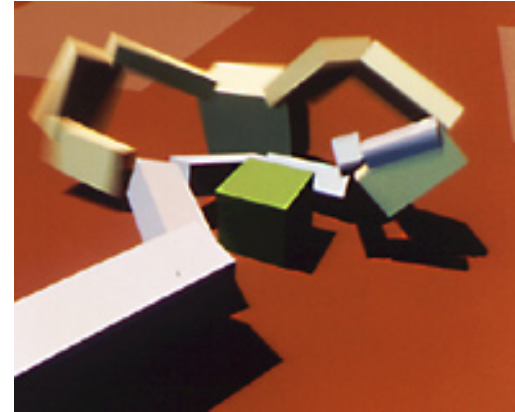
- Is Evolution gradual? Are there “Gaps” in the fossil record?



- species remain in a meta-stable “niche” for a long time
- external change can lead to rapid adaptation
- environmental change, or a new predator or competitor

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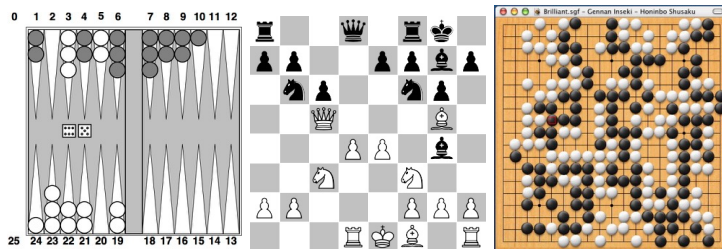
## Coevolving Virtual Creatures (Sims, 1994)



- both body and controller of creature evolve
- aim is to get the cube away from opponent

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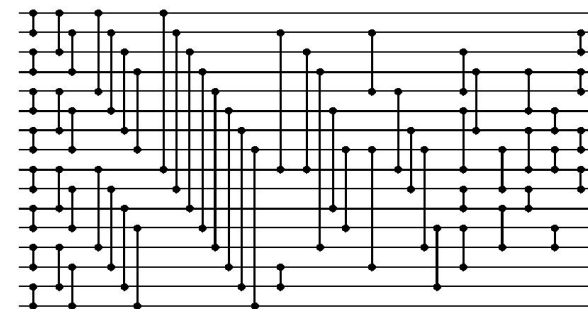
## Adversarial Game Learning



- learning strategic games by self-play
- coevolutionary dynamics
  - can help to improve robustness
  - could also get stuck in oscillation or mode collapse

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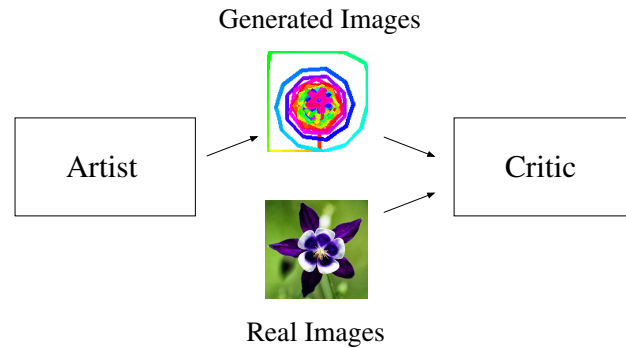
## Coevolution of Sorting Networks (Hillis, 1990)



- coevolution: sorting networks vs. strings to be sorted
- punctuated equilibria, can escape from local optima

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## Artist-Critic Co-Evolution



- ▶ Critic is rewarded for distinguishing real images from those generated by the Artist
- ▶ Artist is rewarded for fooling the Critic into thinking that generated images are real

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## The Creative Act (Marcel Duchamp, 1957)

"All in all the creative act is not performed by the artist alone; the **spectator** brings the work in contact with the external world by deciphering and interpreting its inner qualifications and thus adds his contribution to the creative act."



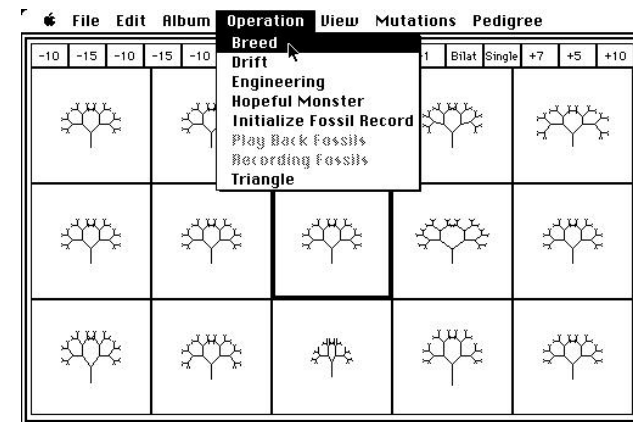
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## Artist-Critic Co-Evolution Paradigms

| Artist   | Critic | Method                      | Reference          |
|----------|--------|-----------------------------|--------------------|
| Biomorph | Human  | Blind Watchmaker            | (Dawkins, 1986)    |
| GP       | Human  | Blind Watchmaker            | (Sims, 1991)       |
| CPPN     | Human  | PicBreeder                  | (Secretan, 2011)   |
| CA       | Human  | EvoEco                      | (Kowaliw, 2012)    |
| GP       | SOM    | Artificial Creativity       | (Saunders, 2001)   |
| Photo    | NN     | Computational Aesthetics    | (Datta, 2006)      |
| GP       | NN     | Computational Aesthetics    | (Machado, 2008)    |
| Agents   | NN     | Evolutionary Art            | (Greenfield, 2009) |
| GP       | NN     | Aesthetic Learning          | (Li & Hu, 2010)    |
| HERCL    | HERCL  | Co-Evolving Line Drawings   | (Vickers, 2017)    |
| HERCL    | DCNN   | HERCL Function/CNN          | (Soderlund, 2018)  |
| DCNN     | DCNN   | Generative Adversarial Nets | (Goodfellow, 2014) |
| DCNN     | DCNN   | Plug & Play Generative Nets | (Nguyen, 2016)     |

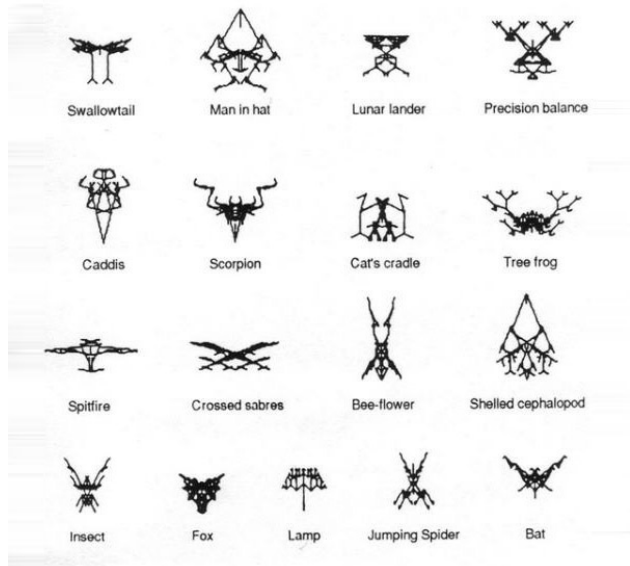
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## Blind Watchmaker (Dawkins, 1986)



- ▶ the (Human) user is presented with 15 images
- ▶ chosen image(s) are used to breed the next generation

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- ▶ Artist = Genetic Program (GP)
  - ▶ used as function to compute R,G,B values for each x,y pixel
- ▶ Critic = Human

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- ▶ Artist = Compositional Pattern Producing Network (CPPN)
  - ▶ Critic = Human
  - ▶ interactive Web site ([picbreeder.org](http://picbreeder.org)) where you can choose an existing individual and use it for further breeding
- 
- ▶ Interactive Evolution paradigm is cool, but it may require a lot of work from the Human
  - ▶ Can the Human be replaced by an automated Critic?

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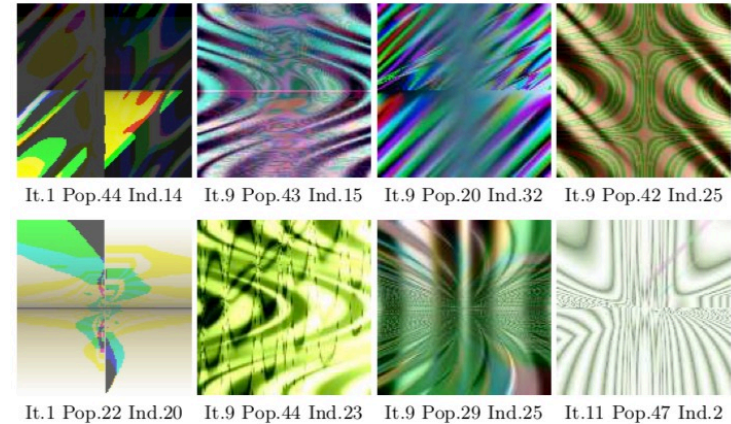


## Evolutionary Art (Fully Autonomous)

- ▶ Artist = Genetic Program (GP or HERCL)
  - ▶ artist used as a function to compute R,G,B values for each pixel location  $x, y$
  - ▶ alternatively, artist issues a series of drawing instructions
- ▶ Critic = GP (evolution) or Neural Network (backpropagation)
- ▶ Critic is presented with “real” images from a training set, and “fake” images generated by the Artist
- ▶ Critic is trained to produce output close to 1 for real images and close to 0 for generated images (or vice-versa)
- ▶ inputs to Critic
  - ▶ small number of statistical features extracted from the image
  - ▶ more recently, raw image, fed to DCNN

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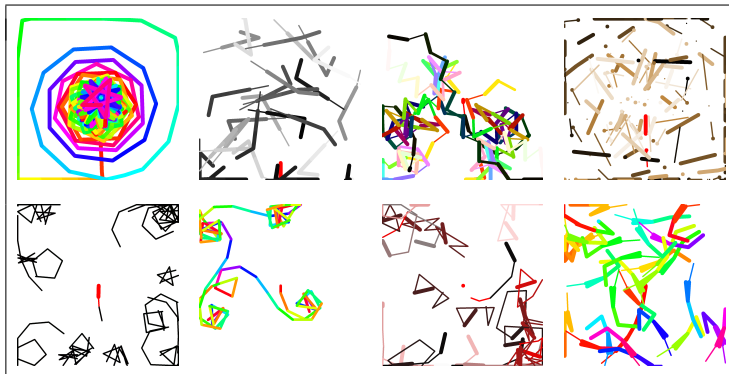
## Computational Aesthetics (Machado, 2008)



- ▶ Generator = Genetic Program
- ▶ Critic = 2-layer NN, using statistical features of image

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## Co-Evolving Line Drawings (Vickers, 2017)



- ▶ Generator = Genetic Program (HERCL)
- ▶ Critic = GP (HERCL), using statistical features of image

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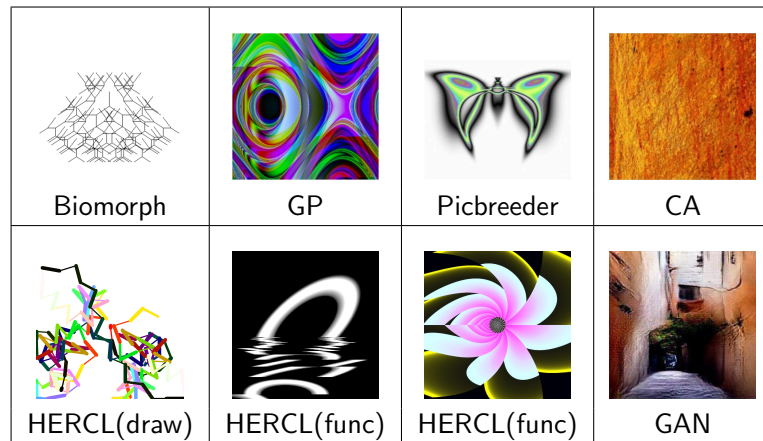
## Line Drawing Commands

|   |        |         |   |
|---|--------|---------|---|
| 0 | TOGGLE |         | lift pen on/off page                                  |
| 1 | MOVE   | $x$     | move pen forward by $x$ pixels ( $0 \leq x \leq 15$ ) |
| 2 | TURN   | $x$     | turn $x$ degrees clockwise                            |
| 3 | SIZE   | $p$     | set pen radius to $p$ pixels ( $1 \leq p \leq 4$ )    |
| 4 | COLOUR | $v$     | set greyscale value [greyscale mode]                  |
|   | COLOUR | $l h s$ | set colour in HSV colour space [colour mode]          |

- ▶ the output from the HERCL program is interpreted as a series of line drawing commands
- ▶ Critic is also a HERCL program, based on 20 statistical features extracted from the image

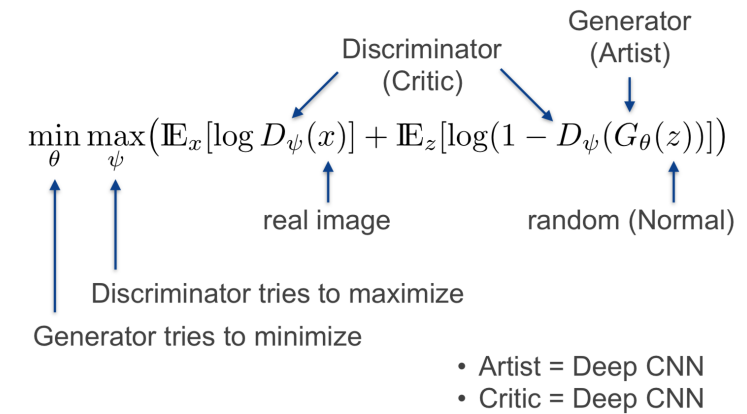
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## Image Generating Paradigms



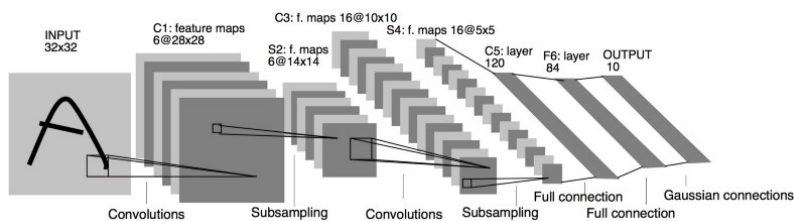
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## Generative Adversarial Networks (Goodfellow, 2014)



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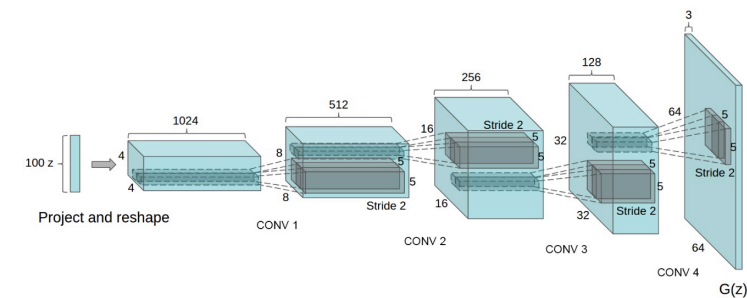
## LeNet CNN Discriminator (LeCun, 1998)



- convolutional layers
- max pooling
- fully connected layers
- for Discriminator, only two outputs

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## GAN Generator Architecture (Radford, 2016)



- differentials are backpropagated from Discriminator, through image and into Generator

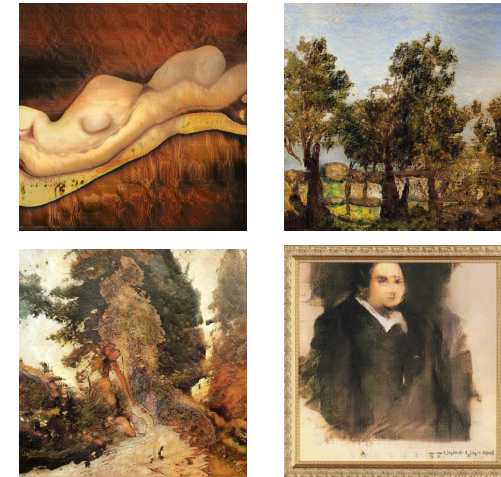
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## GAN Generated Images (Radford, 2015)



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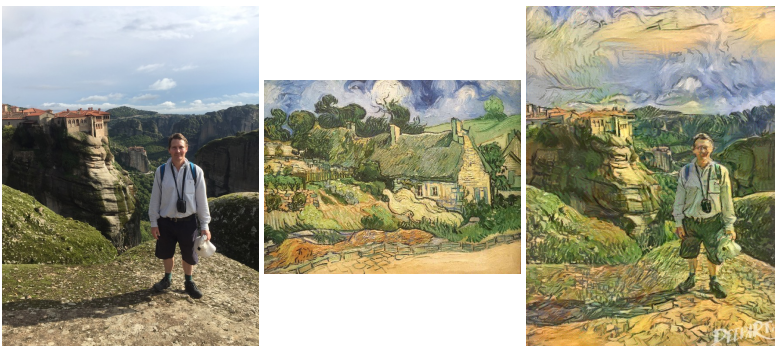
## GAN Generated Art (Robbie Barrat; Obvious)



GAN is shown paintings by humans and asked to mimic the style

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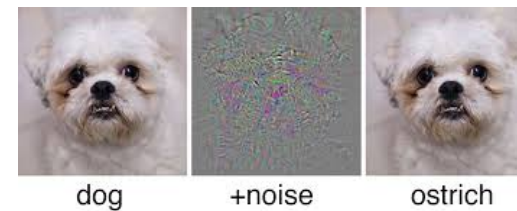
## Neural Style Transfer (Gatys, 2015)



Another method for producing art in the style of a human artist

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## GANs exhibit Coevolutionary Dynamics

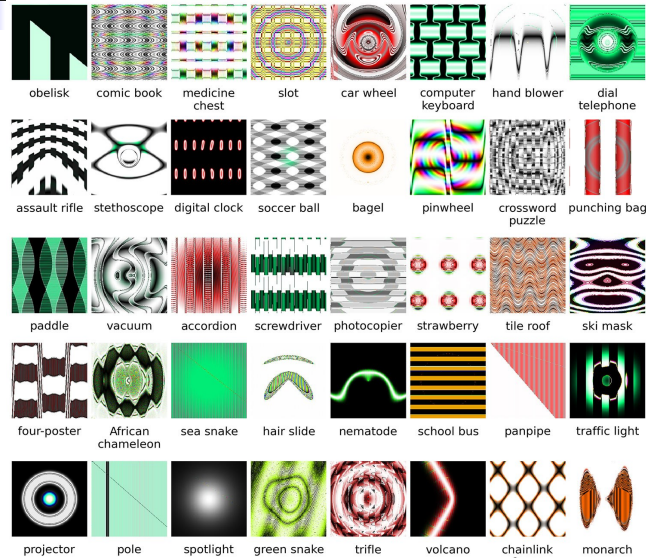


- ▶ Deep Networks are easily fooled
- ▶ Adversarial training improves quality of images, but can lead to well known problems of coevolutionary dynamics
  - ▶ oscillation
  - ▶ mode collapse

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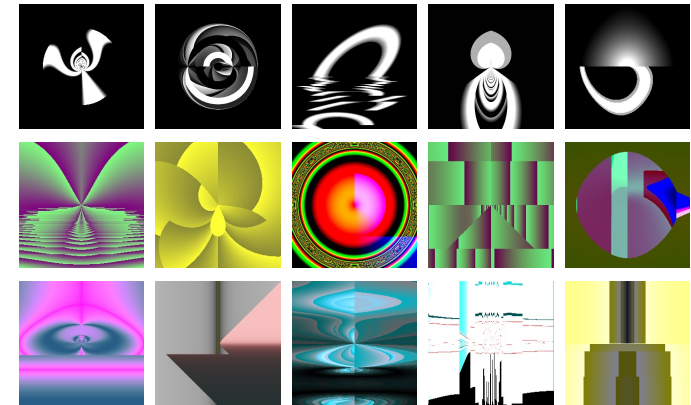


## CPPN vs. Pre-Trained ImageNet (Nguyen, 2015)



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## Adversarial Evolution and Deep Learning (Soderlund, 2018)



- ▶ Artist = HERCL program as a function from  $x, y$  to  $R, G, B$
- ▶ Critic = Deep Convolutional Neural Network (LeNet)

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## Adversarial Evolution and Deep Learning

- ▶ Generator = HERCL program, as function from  $x, y$  to  $R, G, B$
- ▶ Critic = LeNet CNN
- ▶ in each round, a new Critic is trained to distinguish real images from those previously produced by the Generator
- ▶ HERCL Generator is then evolved to produce an image for which the current Critic will assign the best possible score
- ▶ each round adds one new image to the gallery
- ▶ Generator can re-use code from previous images in gallery
- ▶ at the end of the process, Human chooses from the 600-1000 images generated

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## Hierarchical Evolutionary Re-Combination Language (HERCL)

```

INPUT:    ickey
OUTPUT:
MEMORY:   Minnie.....
REGISTERS:  ....[6]..[1].  [7]
STACK:    MM
CODE:     0[is|. <sy^5>};i|8{^s-~:~+7=;wo8|-wo]

```

- ▶ combines elements from Linear GP and Stack-based GP.
- ▶ programs have access to a stack, registers and memory.
- ▶ each instruction is a single character, possibly preceded by a numerical (or dot) argument.

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**Input and Output**

- i fetch INPUT to input buffer
- s SCAN item from input buffer to stack
- w WRITE item from stack to output buffer
- o flush OUTPUT buffer

**Stack Manipulation and Arithmetic**

- # PUSH new item to stack .....  $\mapsto$  ..... x
- ! POP top item from stack .....  $x \mapsto$  .....
- c COPY top item on stack .....  $x \mapsto$  ..... x, x
- x SWAP top two items ... y, x  $\mapsto$  ... x, y
- y ROTATE top three items z, y, x  $\mapsto$  x, z, y
- NEGATE top item .....  $x \mapsto$  ..... (-x)
- + ADD top two items ... y, x  $\mapsto$  ... (y+x)
- \* MULTIPLY top two items ... y, x  $\mapsto$  ... (y \* x)

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**Mathematical Functions**

- r RECIPROCAL ..  $x \mapsto$  ..  $1/x$
- q SQUARE ROOT ..  $x \mapsto$  ..  $\sqrt{x}$
- e EXPONENTIAL ..  $x \mapsto$  ..  $e^x$
- n (natural) LOGARITHM ..  $x \mapsto$  ..  $\log_e(x)$
- a ARCSINE ..  $x \mapsto$  ..  $\sin^{-1}(x)$
- h TANH ..  $x \mapsto$  ..  $\tanh(x)$
- z ROUND to nearest integer
- ? push RANDOM value to stack

**Double-Item Functions**

- % DIVIDE/MODULO .. y, x  $\mapsto$  .. (y/x), (y mod x)
- t TRIG functions ..  $\theta, r \mapsto$  ..  $r \sin \theta, r \cos \theta$
- p POLAR coords .. y, x  $\mapsto$  ..  $\text{atan2}(y, x), \sqrt{x^2 + y^2}$

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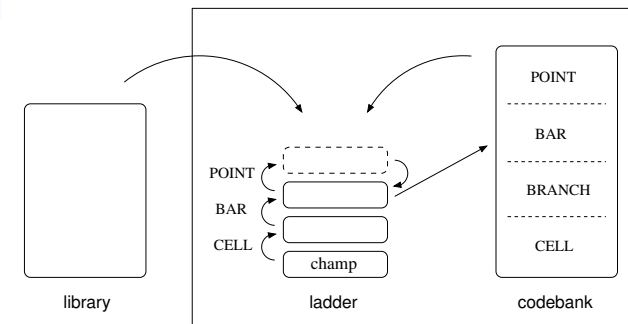
**Registers and Memory**

- < GET value from register
- > PUT value into register
- ^ INCREMENT register
- v DECREMENT register
- { LOAD from memory location
- } STORE to memory location

**Jump, Test, Branch and Logic**

- j JUMP to specified cell (subroutine)
- | BAR line (RETURN on . | HALT on 8 |)
- = register is EQUAL to top of stack
- g register is GREATER than top of stack
- : if TRUE, branch FORWARD
- ; if TRUE, branch BACK
- & logical AND / logical OR ~ logical NOT

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- ▶ large crossover/mutation can be followed up by smaller ones.
- ▶ if top agent becomes fitter, it moves down to replace the one below it (which is moved to the codebank).
- ▶ if top agent exceeds max number of offspring, it is removed.
- ▶ good for co-evolution because it keeps the number of competing agents small while preserving diversity.

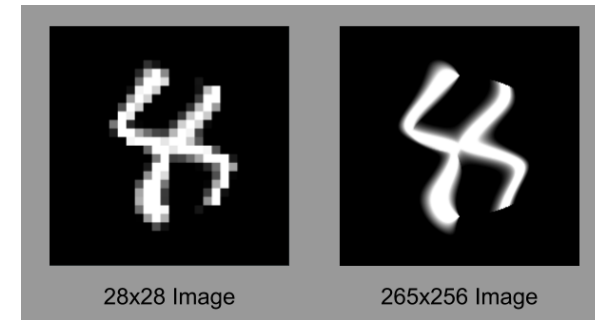
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## Previous HERCL Tasks

- ▶ Classification Tasks
  - ▶ sonar
  - ▶ ionosphere
  - ▶ promoters
  - ▶ Australian credit card fraud
  - ▶ Pima Diabetes prediction
- ▶ Control Tasks
  - ▶ double pole balancing
- ▶ String Processing Tasks
  - ▶ strcat
  - ▶ strlen
  - ▶ strchr
  - ▶ strcmp
- ▶ Other Tasks
  - ▶ Caesar & Vigenere Cipher
  - ▶ Postfix Calculator

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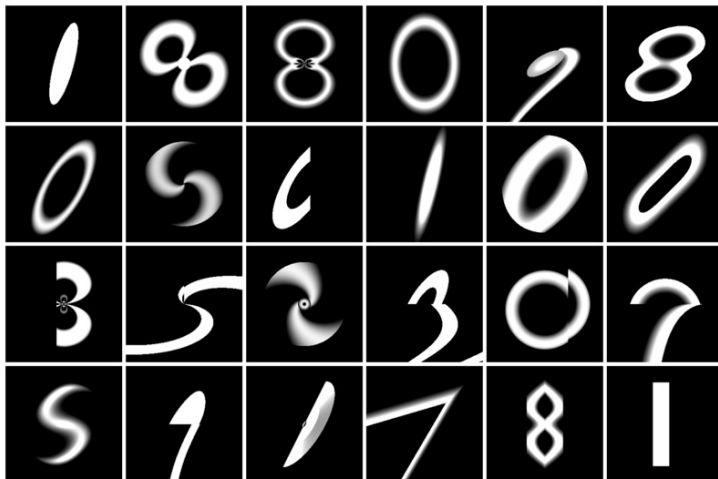
## Up-Scaling of Images



- ▶ images are fed to the critic at low resolution
- ▶ afterwards, images can be re-generated at high resolution

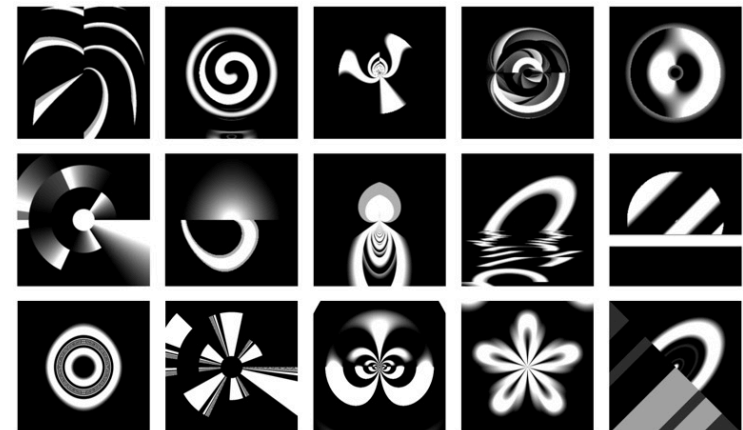
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## Images trained with MNIST digits



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## Images trained with MNIST digits



These ones don't look like digits, but may have some artistic merit

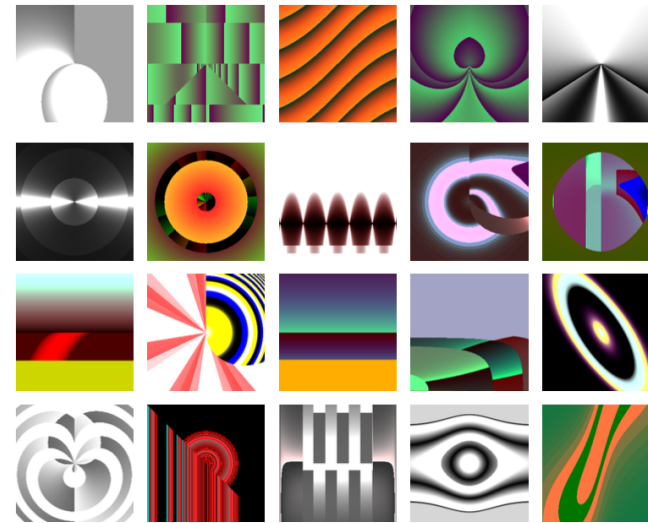
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Images trained with CIFAR-10 photographs



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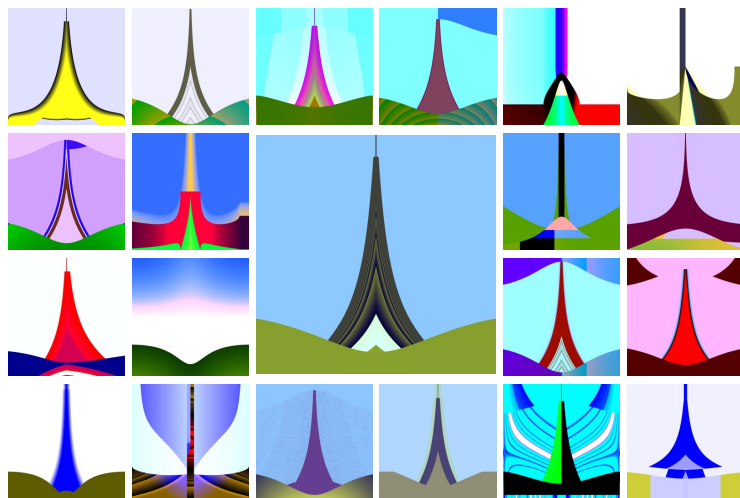
Images trained with CIFAR-10 photographs



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Eiffel-ution

Eiffel-ution



Pic&ArtSo.com

Hercule LeNet

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Mona Lisa



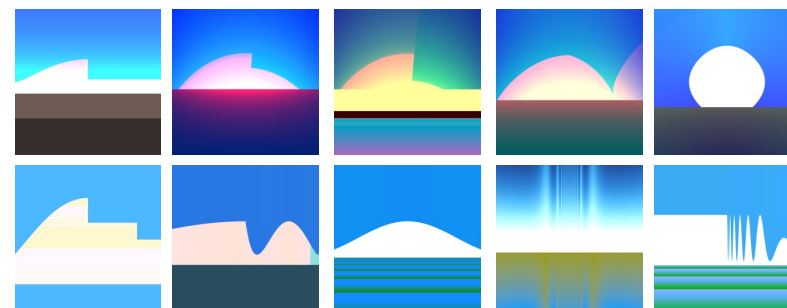
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## Experiments with Landmarks

- ▶ collected photographs of 10 famous landmarks
- ▶ HERCL artist acting as function from  $x, y$  to R,G,B
- ▶ LeNet CNN critic with 16 filters in conv1, 24 in conv2
- ▶ data augmentation, by cropping
- ▶ two different resolutions ( $48 \times 48$  and  $64 \times 64$ )
- ▶ selected 5 best images from each run

## Sydney Opera House

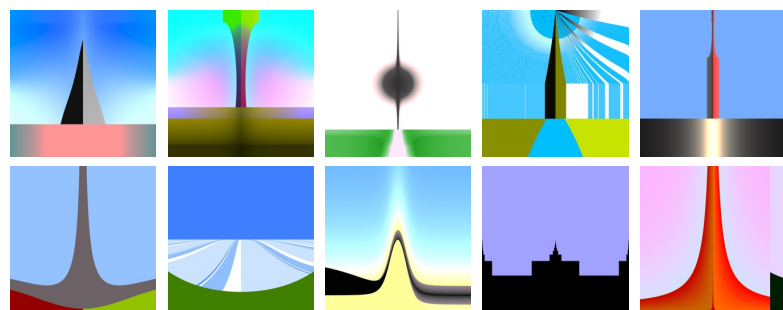
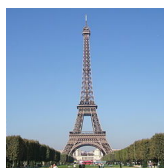
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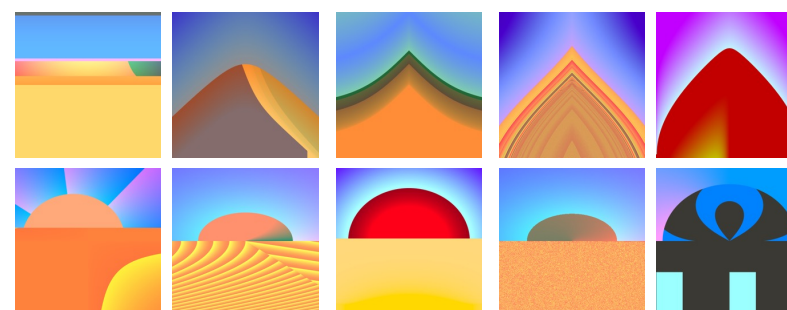
## Eiffel Tower



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## Pyramids

© Ricardo Liberato

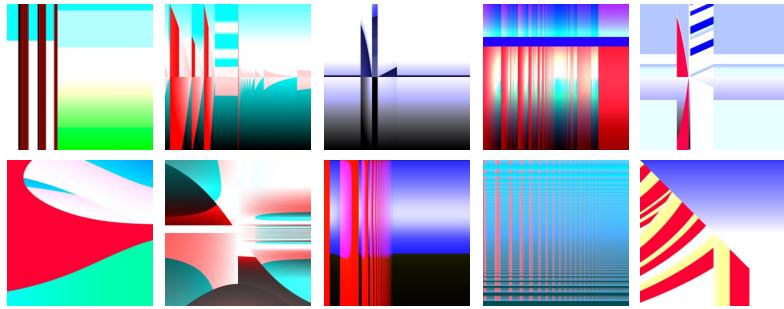


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# Golden Gate Bridge

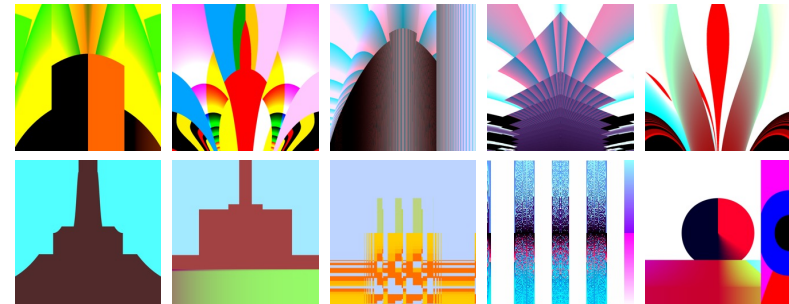
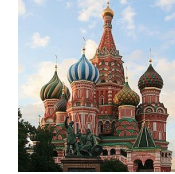
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# Saint Basil's Cathedral

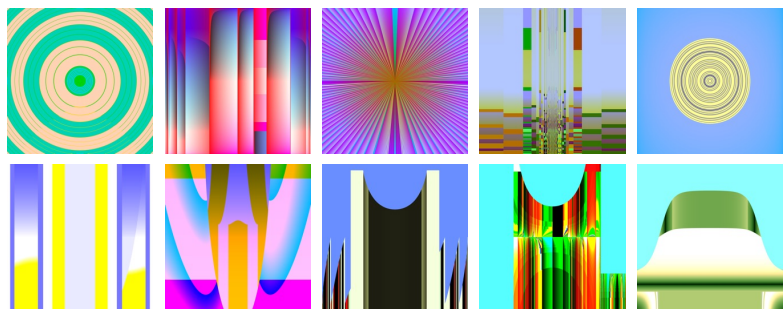
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# Notre Dame de Paris

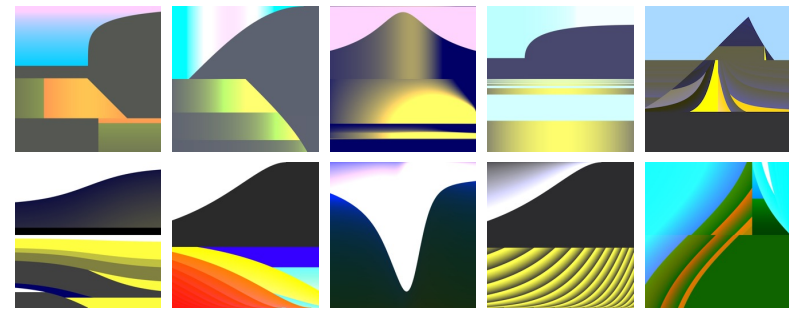
© Jérôme Blum



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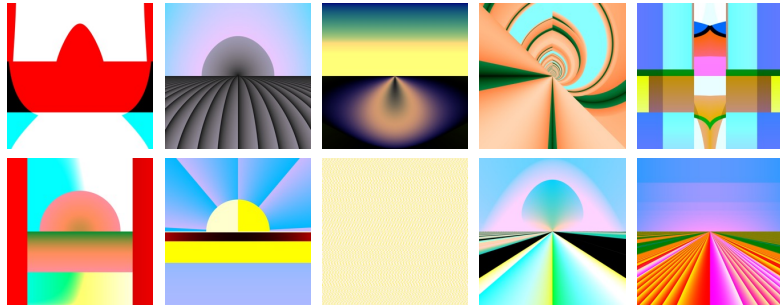
# Machu Picchu

© Allard Schmidt



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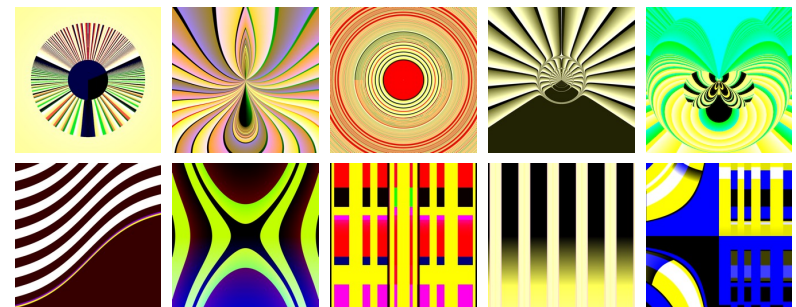
## Taj Mahal



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## Angel Oak Tree

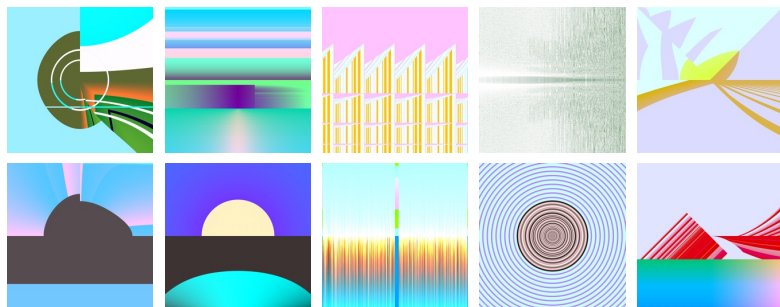
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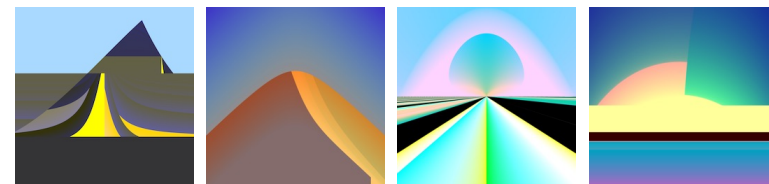
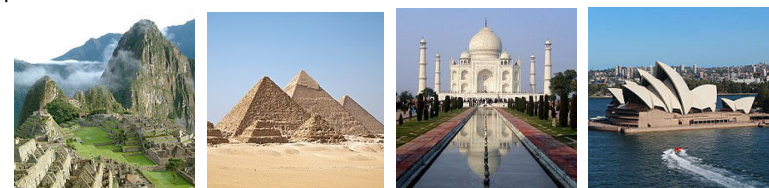
## Grand Canal in Venice

© Hans Peter Schaefer



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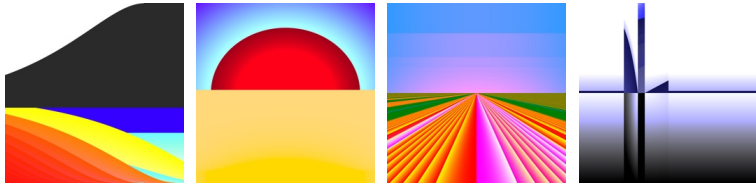
## Minimalism



► the object is suggested by simple lines and shapes

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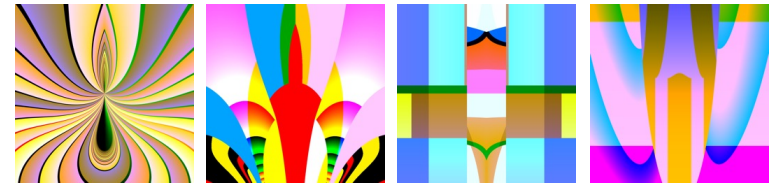
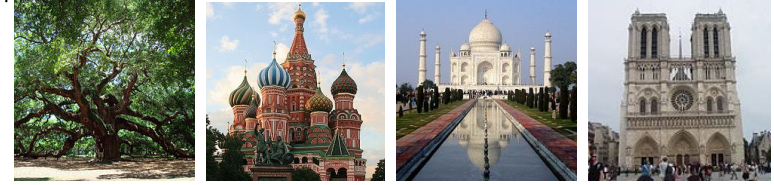
## Colors and Shading



- ▶ vibrant, fauvist colors; or sometimes near black-and-white
- ▶ enhanced contrast, halo effect

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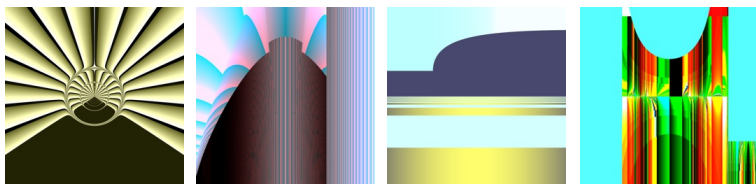
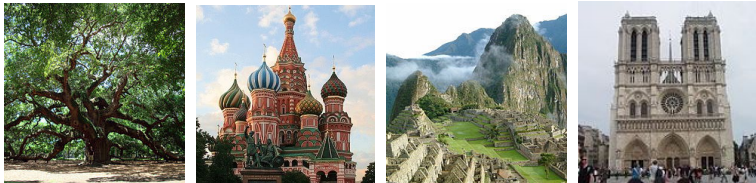
## Abstraction



- ▶ abstract, rather than figurative rendition of the subject
- ▶ colors and shapes recombined in different patterns

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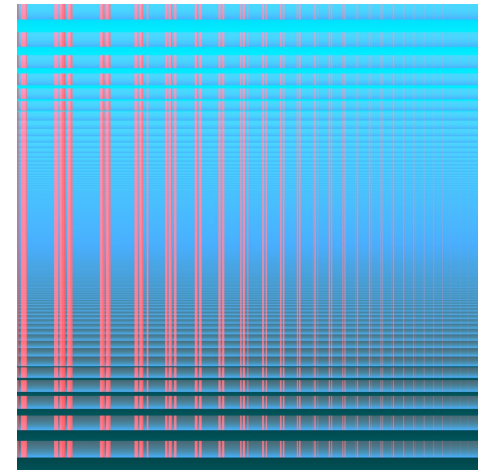
## Fractals



- ▶ low algorithmic complexity achieved through self-similarity
- ▶ fractal art, psychedelic art, stained glass or glass art

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## M.C. Escher?

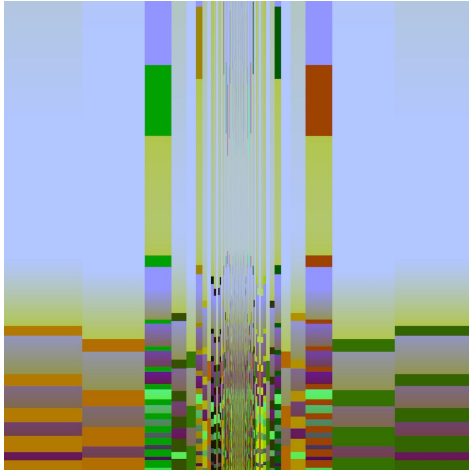


- ▶ where does the real beam end and the reflection begin?

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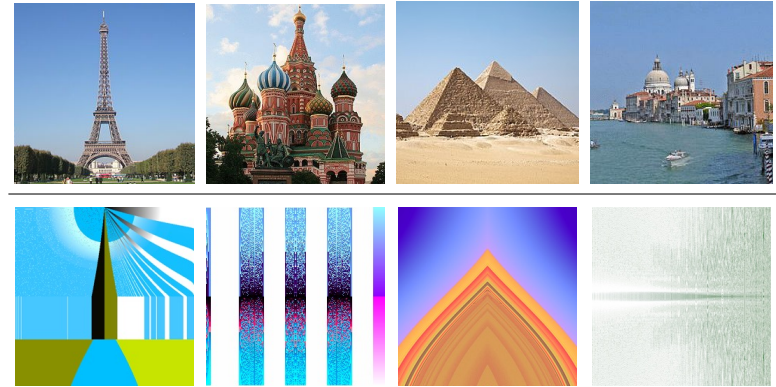
## Notre Dame



- ▶ this image has a distinctive rectangular fractal structure

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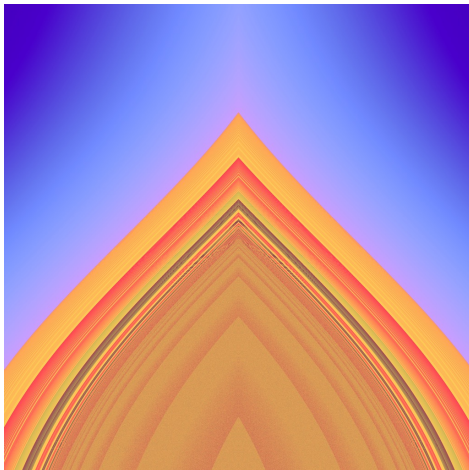
## Pointillism



- ▶ sensitive function from  $x, y$  to  $R, G, B$  creates a pattern of dots

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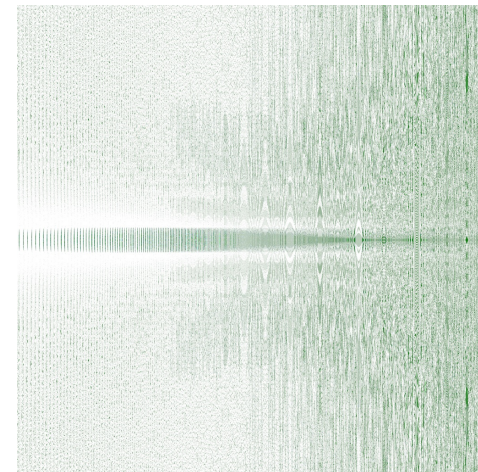
## Great Pyramid



- ▶ can we see the individual grains of sand in the desert?

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## Grand Canal

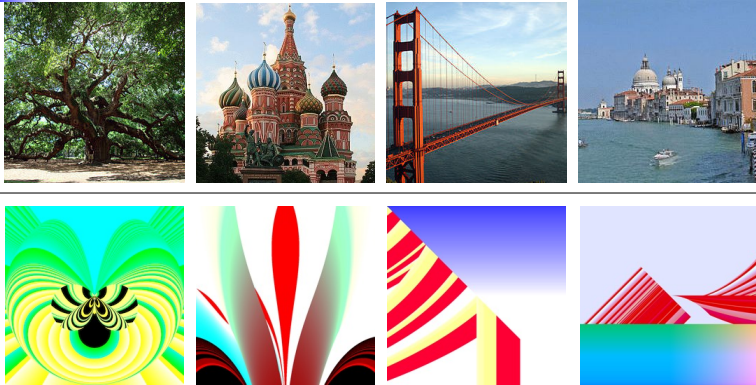


- ▶ is there some kind of structure, reflected in the water?

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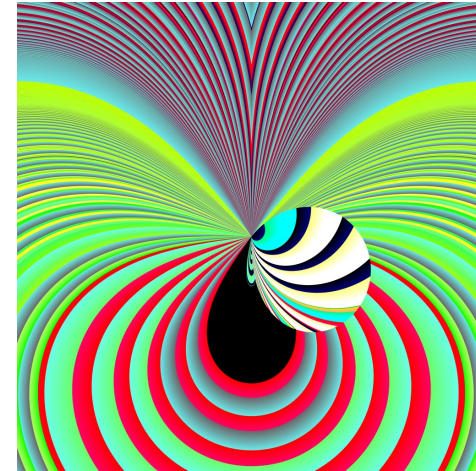
## Metaphor



- ▶ images evolved to resemble one thing may end up looking like something else

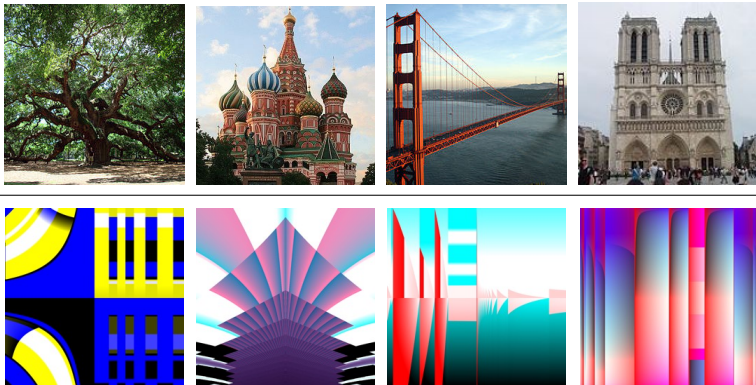
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## Tree or Insect?



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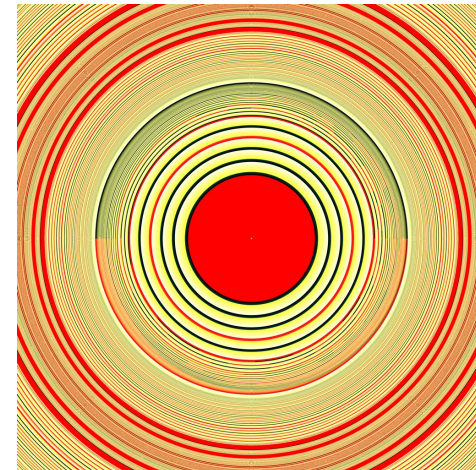
## Repeated Substructures, with Variations



- ▶ imperfectly repeated substructures give the impression of having arisen from some natural process

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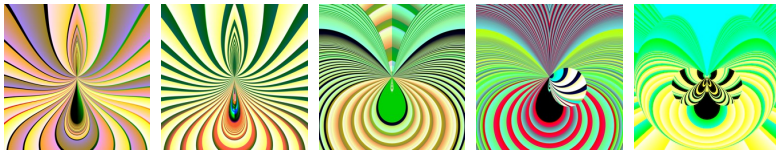
## Tree Rings



- ▶ the tree rings are so real, we can almost smell the sawdust!

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## Re-Combination, Variations on a Theme



- ▶ code for previous images in the gallery is made available for genetic re-combination, allowing the artist to revisit and further develop earlier themes

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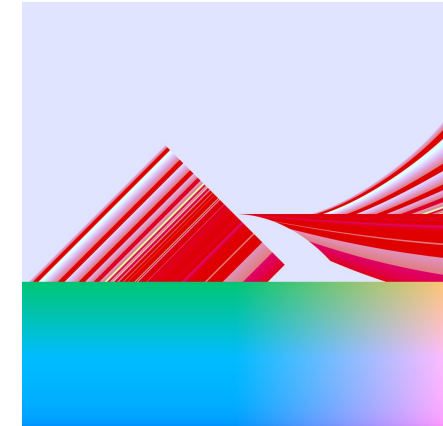
## Pseudocode

```

scan (x, y)           // -1 ≤ x ≤ 1, (upper) -1 ≤ y ≤ 1 (lower)
if y ≥ 0.32           // water
    return (√(y² + (sin⁻¹y)²), atan2(y, sin⁻¹y), x)
else
    if y > x           // obstacle
        u = sin⁻¹(x + y)
    else               // ship
        r = √(y² + tanh(x)²), θ = atan2(y, tanh(x))
        u = sin⁻¹(⌊θ/r⌋ + (θ mod r))
    end
    φ = π/4(-1 + 2 sgn(u)), ρ = sin⁻¹(√2u)
    z = atan2((φ mod ρ), 0.4338), s = √((φ mod ρ)² + 0.4338²)
    if s ≤ z           // sails
        return (z, s, 0.88)
    else               // hull
        v = sin⁻¹(√z) cos(⌊φ/ρ⌋)
        return (v, ⌊v⌋, 0.88) // (blue, green, red)
    end
end
    
```

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## Genotype to Phenotype Mapping



HERCL code:

```

0[!qatcz]
1[capwwwo.]
2[%]
3[is.32#>sg:1j|c>xg:hp2j|+a{>cpa%.4338#p>g~<:0j|xww.88#wo]
    
```

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## Discussion

- ▶ interplay between evolution and deep learning
- ▶ enhanced contrast, variations in color
- ▶ minimalism, abstraction, fauvism, fractals, pointillism, metaphor
- ▶ ability to surprise
- ▶ recombination of previous elements
- ▶ low algorithmic complexity, but realistic enough to fool the critic

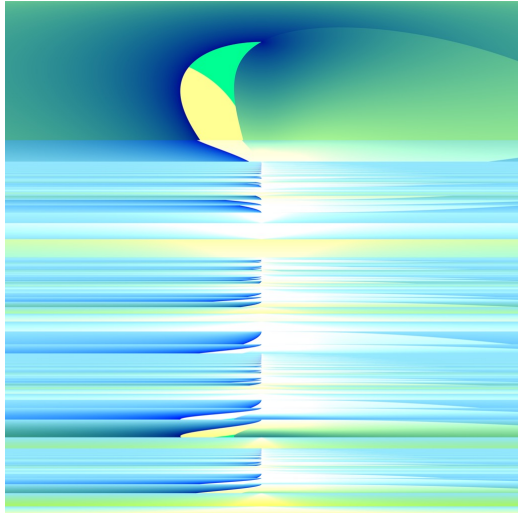
"Imagination is a good servant, and a bad master.  
The simplest explanation is always the most likely."

- Hercule Poirot

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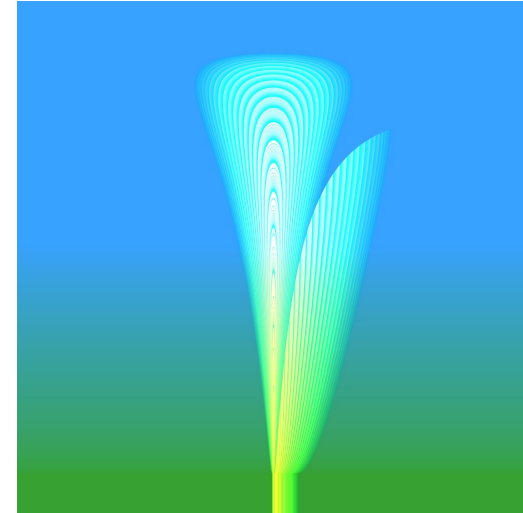
*Burj al Arab*



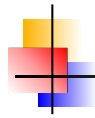
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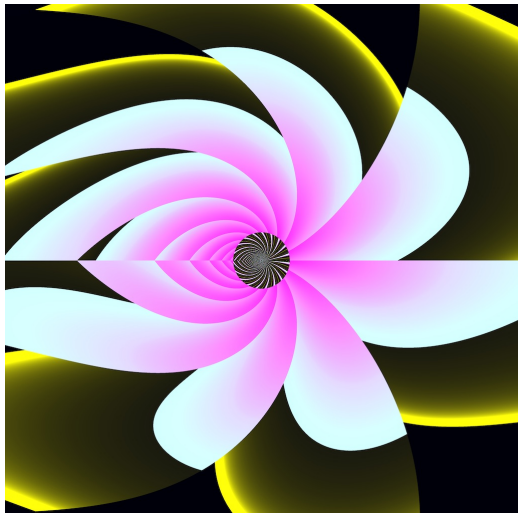
*Cosmic Flower*



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*Water Lily Flowers*



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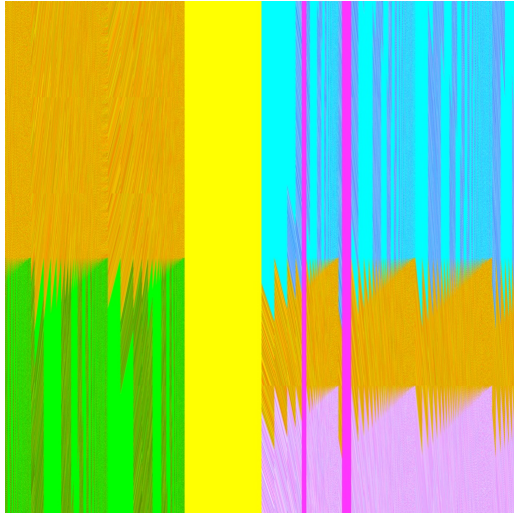
*Water Lily Flowers*



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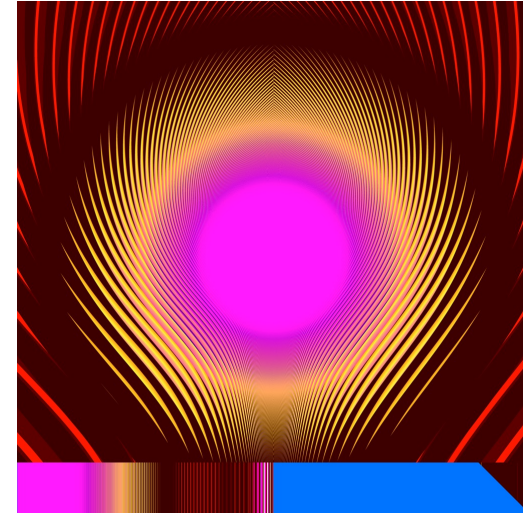


## Abstraction / Details



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## Self-Portrait of AI Artist



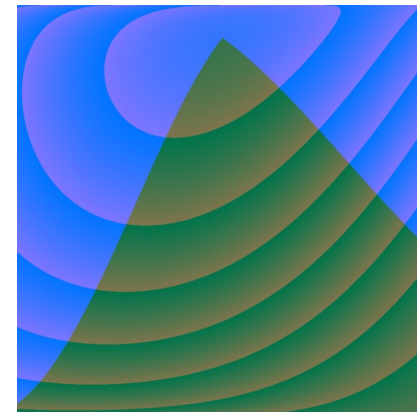
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## Conclusion

- ▶ adversarial training / coevolution is a powerful technique
- ▶ coevolutionary dynamics observed in biology, evolutionary computation, game learning, sorting networks, GANs and evolutionary art
- ▶ deep neural network is a good choice for discriminator, but it is fruitful to try different kinds of generator
- ▶ possible future applications to areas such as text generation, cybersecurity, deception, modular evolving systems, credit assignment

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## Questions



Questions ?

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