



# Evaluation in Computational Creativity

**(Some views.**

**Anna Jourdanous provides a wider perspective)**



# Evaluation is important – and difficult

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- Evaluation of creativity allows us to compare methods and control progress
- However, evaluation of creativity is very difficult
  - No precise definition of creativity
  - Various goals (novelty, value, originality, ...)
  - Context-dependence
  - Cost of evaluation
  - Evolution of (social) systems
  - ...



# What to evaluate?

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- Machine creativity:  
Creative performance of creative programs
- Computer-supported creativity:  
Increase in creativity of humans using CC tools
- Creativity studies: Increase in knowledge about creative processes
  
- Focus here: evaluation of machine creativity



# Evaluation of Machine Creativity

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Two possible targets in evaluation of machine creativity (Colton 2008):

- Artefact-based evaluation: are the results creative?
  - e.g: novelty and value of results
- Process-based evaluation: is the process creative?
  - e.g: combinatorial/ exploratory/ transformational creativity; generation vs. creativity by Ventura; creative acts of the FACE model



# **Ritchie's Framework for Artefact Based Evaluation**

**Ritchie (2007)**



# Essential properties

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Consider a set  $R$  of artefacts produced by a system.

Primitive properties that can be considered:

- **Typicality**: Is the artefact a typical/ recognizable example of the target genre?
- **Novelty**: How (dis)similar is the artefact to existing examples of its genre?
- **Quality** [= Value]



# Formal definitions

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- $\text{typ}(a)$  = amount of typicality associated to artefact  $a$
- $\text{val}(a)$  = amount of quality associated to  $a$
- $T_{\alpha,\beta}(X) = \{a \in X \mid \alpha \leq \text{typ}(a) \leq \beta\}$ 
  - Set of artefacts  $a$  with typicality between  $\alpha$  and  $\beta$
- $V_{\alpha,\beta}(X) = \{a \in X \mid \alpha \leq \text{val}(a) \leq \beta\}$ 
  - Set of artefacts  $a$  with value between  $\alpha$  and  $\beta$
- $\text{size}(X)$  = number of elements of  $X$
- $\text{ratio}(X,Y) = \text{size}(X) / \text{size}(Y)$



## Some criteria

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Criterion 2  $ratio(T_{\alpha,1}(R), R) > \theta$

- at least fraction  $\theta$  of results  $R$  have high typicality ( $>\alpha$ )

Criterion 4  $ratio(V_{\gamma,1}(R), R) > \theta$

- at least fraction  $\theta$  of results  $R$  have high value ( $>\gamma$ )

Criterion 5  $ratio(V_{\gamma,1}(R) \cap T_{\alpha,1}(R), T_{\alpha,1}(R)) > \theta$

- at least fraction  $\theta$  of results  $R$  have both high value ( $>\gamma$ ) and high typicality ( $>\alpha$ )





# Inspiring set

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- Any creative system is based on some existing examples, in one way or another. These can – and should – be taken into account.
- The *inspiring set* consists of all the relevant artefacts known to the program designer, or items which the program is designed to replicate, or a knowledge base of known examples which drives the computation within the program
- Inspiring set  $\approx$  training set in ML/DM



## Some more criteria

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Criterion 9  $ratio(I \cap R, I) > \theta$

- Results R reproduce at least fraction  $\theta$  of the inspiring set I
- Is the system able to reproduce its training examples?

Criterion 10  $ratio(R, I \cap R) > \theta$

- Results R contain at least  $\theta-1$  times as many items outside the inspiring set I as inside it
- Can the system extrapolate outside the training examples?



# Novelty vs. typicality?

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Novelty and typicality are subtly different:

- Not recognizable as a member of the genre  
→ low typicality
- Very different from the inspiring set (but possibly very clearly within the genre) → high novelty



# Comments

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Note: Ritchie does not prescribe a set of criteria. Instead, the criteria must be designed and chosen according to the goals and needs of each work; Ritchie gives examples of some of the possible criteria that one may want to use.



# **FACE Model for Process-Based Evaluation**

**Pease and Colton (2011)**



## F, A, C, E

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- Focus on *creative processes*, not their results
- In the FACE model, systems can be characterized by their creative acts
- The four aspects of the model:
  - F – framing
  - A – aesthetics
  - C – concept
  - E – expression
- Here we present a simplified version



# FACE aspects

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- C: the *concept* or the idea of the artefact
  - E.g. use of excessive rhyming in poetry
- E: a concrete *expression* of the concept
  - E.g. a poem that uses excessive rhyming
- A: a measure of *aesthetics* of the work of art
  - E.g. emotionality etc. of a poem
- F: background information about the piece (*framing*)
  - E.g. a description of why excessive rhyming could be interesting, and what the poem expresses



# Framing

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- Framing is especially important for computational creativity
- It is difficult to appreciate the output (expression) without knowing anything about the process, its goals, etc.
- E.g., is the resulting image pretty just by chance? Or did the system produce it based on some specific criteria and goals? Was the process complicated? Is there some intention, e.g., a message that is being conveyed?





# Ground level of FACE

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- Ground-level generative acts and their products
  - Act  $F^g \rightarrow$  generates an item of framing information
  - Act  $A^g \rightarrow$  generates an aesthetic measure
  - Act  $C^g \rightarrow$  generates a concept
  - Act  $E^g \rightarrow$  generates an expression of a concept
- Any system can now be described in terms of who carries out these acts, and how
  - A simple generative system only performs  $E^g$
  - A system that learns to evaluate also performs  $A^g$
  - (The programmer and other humans probably perform the other acts)



# Meta-level of FACE

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- FACE also has a meta-level: processes that produce ground-level generators
- Process-level acts and their outputs:
  - Act  $F^p \rightarrow$  generates a method for generating framing information
  - Act  $A^p \rightarrow$  generates a method for generating aesthetic measures
  - Act  $C^p \rightarrow$  generates a method for generating concepts
  - Act  $E^p \rightarrow$  generates a method for generating expressions of a concept



# Example from Pease et al, 2011

## The Upside Downs by Verbeek

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## FACE Upsidedowns

$F^P$ : Methods for generating the contextual history of this genre of art

$F^g$ : The contextual history of this genre of art, motivation, justification, etc.

$A^P$ : Methods for generating the idea of art having multiple meanings when viewing from multiple perspectives

$A^g$ : The idea of art having multiple meanings when viewing from multiple perspectives

$C^P$ : Methods for generating new perspectives from which the art might make sense

$C^g$ : The constraint that a picture must make sense when upside down

$E^P$ : Methods for generating expressions of art which have a different meaning when viewed upsidedown

$E^g$ : Expressions of art which have a different meaning when viewed upsidedown (see figure 1)