Evolutionary Art

No Institute Given

Abstract. <Text of the summary of your article>

1 Introduction

This paper is organized as follows: In the next Section, a brief review on Evolutionary Art is presented.

2 Evolutionary Art

Creative evolutionary systems are used to evolve aesthetically pleasing or innovative structures [?].

2.1 Art Representation for Evolutive Art

- alalala

2.2 Aesthetic measures for evolutive art

MAIN CHALLENGE -> HOW TO MEASURE AESTHETICS.

Definition Two modes of aesthetics measures can be defined [?]:

- 1. Aesthetics evaluations are expected to simulate, predict or cater to humans notions of beauty and taste.
- 2. Is an aspect of meta-aesthetic exploration and usually involves aesthetic standards created by software agents in artificial worlds.

According to Galanter [?], computational aesthetics measures can be classified in the following categories:

- Based in Design Principles. Like the rule of thirds or theory of color (e.g., opposite colors) [?].
- Based in Neural Networks and Connective Models.
- Based in Evolutionary Systems:
 - Interactive Evolutionary Computation. The fitness of the individuals is determined by human agents.
 - Performance based goals. Certain properties of the art piece are evaluated and optimized based in performance measures (e.g., usable surface in furniture design generator).
 - Error relative to Exemplars. The individual fitness is measured using a real-world example (e.g., a photography or painting) [?].
 - Complexity measures. This type of measures is based in the idea the complexity is directly related to aesthetics, following the path firstly stablished by Birkhoff [?].

- Multi-objective. Given the multidimensional nature of aesthetics judgement, multi-objective EAs are a clear option in order to deal with this multidimensionality.
- Extensions to EA (such as, coevolution, agent swarm behavior, etc.).
- Complexity Based Models

En [?], Li et al. proponen las siguiente métricas para el aprendizaje estético:

- Color ingredient.
- Image complexity.
- Image order.
- MC metric.
- BL Metric.

En [?], presenta una comparación de tres métricas estéticas:

- Benford Law.
- Global Contrast Factor.
- Information Theory.

En [?], presenta una comparación de cuatro métricas estéticas:

- Machado and Cardoso.
- Ross and Ralph.
- Fractal Dimension.
- A weighted sum of the above mentioned metrics.

En [?] se presenta una aproximación multi-objetivo para arte evolutivo. Las tres funciones de fitness utilizadas son:

- Benford Law.
- Global Contrast Factor.
- Ross and Ralph (bell curve).

En [?] se presenta un AE para crear arte evolutiva a partir de imágenes vectorizadas. La función de fitness utilizada es la diferencia de tono entre distintas regiones de la imagen a distintas resoluciones.

En [?] they present an automatic fitness function specific to portrait painting based in four scores:

- Resemblance.
- Composition (face vs background).
- Tonality.
- Color.

3 Genetic Operators

- 3.1 Representation
- 3.2 Initialization
- 3.3 Mutation
- 3.4 Crossover
- 3.5 Fitness Functions

Histogram HISTOGRAMA DEF: a graphical representation of the tonal distribution in an image.

Image Matching

4 Experimental Results

5 Conclusions and Future Work

Aknowledments.

References

- 1. George David Birkhoff. Aesthetic Measure. Harvard University Press, 1933.
- 2. E Den Heijer and A Eiben. Comparing aesthetic measures for evolutionary art. Applications of Evolutionary Computation, pages 311–320, 2010.
- 3. E Den Heijer and A Eiben. Evolving art using multiple aesthetic measures. Applications of Evolutionary Computation, pages 234–243, 2011.
- 4. E den Heijer and A Eiben. Evolving pop art using scalable vector graphics. Evolutionary and Biologically Inspired Music, Sound, Art and Design, pages 48–59, 2012.
- 5. E den Heijer and AE Eiben. Using aesthetic measures to evolve art. In Evolutionary Computation (CEC), 2010 IEEE Congress on, pages 1–8. IEEE, 2010.
- Steve DiPaola and Liane Gabora. Incorporating characteristics of human creativity into an evolutionary art algorithm. Genetic Programming and Evolvable Machines, 10(2):97–110, 2009.
- Philip Galanter. Computational aesthetic evaluation: past and future. In Computers and Creativity, pages 255–293. Springer, 2012.
- 8. Yang Li, Changjun Hu, Ming Chen, and Jingyuan Hu. Investigating aesthetic features to model human preference in evolutionary art. *Evolutionary and Biologically Inspired Music, Sound, Art and Design*, pages 153–164, 2012.