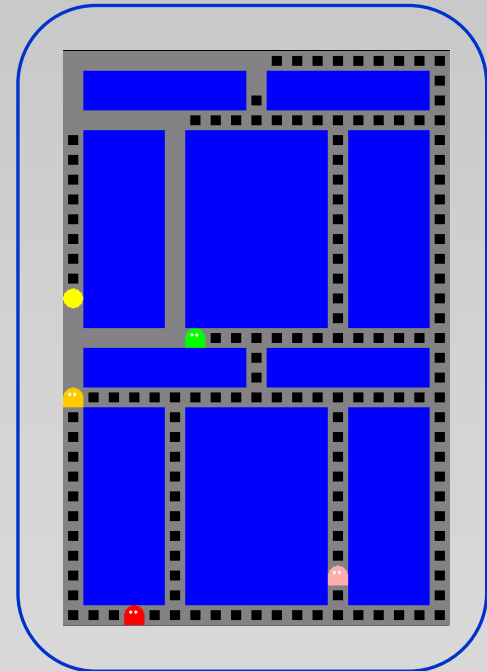


→ Tutorial on “Measuring and Optimizing Player Satisfaction” Computational Intelligence and Games '08 Perth, Australia, December 15, 2008

→ **Georgios N. Yannakakis**
Center for Computer Games Research (CGR)
IT-University of Copenhagen (ITU)
Denmark

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Dalle-Molle Institute for AI (IDSIA)
Switzerland



→ Why a Tutorial on Player Satisfaction?

- Increasing interest (different disciplines)
- Growing Community
- High quality publications (more)
- Game developers believe is hot!
- Organization:
 - IEEE Task force (game.itu.dk/PSM/)
 - 2 Workshops: SAB'06, AIIDE'07
 - Special Session at CIG'08
- Time for more CIG people to see the potential
- That's essentially our "objective function"
- Goal of the whole game design process (?)

→ Who is this tutorial for...?

- People
 - With CI backgrounds and/or
 - Using games as their domain
- People thinking...
 - ...it's too early for such a direction
 - ...that sounds cool! Can I really capture "fun"?
- That's CI&G people.. but also
- Psychologists/Usability testers/evolutionary & emergent design researchers/ game experience researchers
- Game Developers

→ Tutorial Structure

- **PART I** – NPC Behavior (Georgios)
 - Introduction
 - What is Fun? Review on Entertainment Capture
 - Modeling Player Satisfaction
 - Custom-designed metrics
 - Machine Learning
 - Data Source: Keyboard, Physiology
 - Optimizing Player Satisfaction
 - Gameplay – Test-beds:
 - Screen-Based Games
 - Physical Interactive Games
 - Open Research Questions

→ Tutorial Structure

- **PART II** – Game design (Julian)
 - Static vs. Dynamic Predictors
 - Environments
 - Narratives
 - Rules and parameters
 - Open questions



Introduction

→ Human Experience and Fun

- Play Modes: Screen-based, physical interactive, mixed–reality..
- Several experiences emerge during HCI
- Why entertainment, fun, player satisfaction?
 - The most essential part of play!
 - Fun is a term easily interpretable by humans
 - The more the fun – the more the learning – the higher the quality of the interaction!
- Definition of Fun?
 - No way!
 - Approximation of human response instead
 - Non-linear!

→ Human Experience and Fun

- Challenges towards capturing “fun”? many!
 - Complicated mental process
 - Unique (subjectivity)
 - Augmented experimental (report, hardware) noise
 - ...
- Ways of modeling fun?... Some (ideas are welcome)
- Overall purpose?
 - Make something useful with all this data/exploit multimodality
 - Richer HCI
 - Personalized HCI systems
 - Better understanding of humans (and games)
 - ...

→ “Fun” and commercial Game Development

- Intelligence can be generated easily through FSMs!
- CI in game development?
 - Not much...but why?
 - Unpredictable behaviors
 - Debugging issues
 - Emergence – Wow, can you repeat that?
 - Expressiveness – Wow, how did he do that?
 - Game engine (h-FSM) compatibility
 - ...
 - ...

→ “Fun” and commercial Game Development

- However, need for believability and more fun in real-time... how?
- CI is here to provide
 - Believable Characters
 - More Fun in Real-time
 - That’s the purpose of this tutorial!

→ “Fun” and academia

- Emerging research direction
- Optimizing **performance** of NPCs (intelligence)
 - That is implicitly more “fun”
- Optimizing for “**fun**” is a hard and interesting problem
 - Answers the key question of AI in Games

→ After this tutorial...

Some answers to the following

- Which are the features/criteria that collectively determine enjoyment (in games).
- How to quantitatively **measure** the player's satisfaction (entertainment, fun) in real-time.
- How to **increase** a game's low entertainment value and/or how to maintain a high value of entertainment.



What is “Fun”?



→ Entertainment Capture

What is “fun” and how to measure?

G. N. Yannakakis, “**How to Model and Augment Player Satisfaction: A Review**,” in *Proceedings of the 1st Workshop on Child, Computer and Interaction (WOCCI)*, 2008.

■ Qualitative

- Malone, 1981 → Challenge, Curiosity, Fantasy
- Kline and Alridge, 2003; Lazzaro, 2004 → Malone + Socialization
- Read et al., 2002 → Expectations, Engagement, Endurability
- Vorderer *et al.*, 2003 → Entertainment is all about competition
- Koster, 2005 → Theory of Fun
- Sweetser and Wyeth, 2005 → GameFlow (*theory of flow*)
- Pagulayan et al., 2007 → extensive outline of *game testing* methods for effective user-centered game design.
- Ijssellstein et al., 2007, → challenges of measuring game experiences and highlight the concepts of *immersion* and *flow*
- Ryan et al., 2006 → perceived in-game autonomy and competence are associated with game enjoyment.

→ Entertainment Capture

What is “fun” and how to measure?

G. N. Yannakakis, "How to Model and Augment Player Satisfaction: A Review," in *WOCC/Proceedings*, 2008.

- Limitations of Qualitative Approaches
 - Based on
 - empirical observations or
 - linear correlations of user input (interaction and physiological data) with reported emotions
 - Likert scale questionnaires are used
- We get inspiration from those
- Focus on quantifying “fun”

→ Entertainment Capture

What is “fun” and how to measure?

G. N. Yannakakis, "How to Model and Augment Player Satisfaction: A Review," in *WOCC/Proceedings*, 2008.

■ Quantitative

■ Player-Game Interaction Data (Focus on Player-NPC Interaction)

- Iida, 2003 → Entertainment Metrics for Board games
- Yannakakis and Hallam, 2004 → Entertainment Custom-Designed Metrics for Prey/Predator games (match human entertainment)
- Yannakakis and Hallam, 2006 → ANNs and Fuzzy-NN models get closer to human notion of entertainment (Prey/Predator games)
- Yannakakis and Hallam, 2007 → Accurate ANN models of fun preference (physical interactive games)

■ Introduction of *comparative fun analysis*, opposed to Likert scale methodology

→ Entertainment Capture

What is “fun” and how to measure?

G. N. Yannakakis, "How to Model and Augment Player Satisfaction: A Review," in *WOCC/Proceedings*, 2008.

■ Quantitative (cont.)

■ Physiological Data

- Mandryk et al., 2006 → Correlations between GSR, jaw EMG, respiration, cardiovascular measures and reported experiences
- Fuzzy model indicates high arousal and positive valence (i.e. possibly ‘fun’) when a smiling (jaw EMG) player has high HR and GSR (Mandryk et al., 2007).
- Hazlett (2006) → use of facial EMG to distinguish positive and negative emotional valence during interaction with racing video games.
- Rani et al. (2006) → player anxiety detection which adjusts the level of challenge (e.g. speed) in the game of ‘Pong’.
- Yannakakis et al, (2007, 2008) → Identification of physiological features corresponding to “fun” in physical interactive games; construction of very accurate “fun” models

→ Optimize Player Satisfaction

A brief review

G. N. Yannakakis, "How to Model and Augment Player Satisfaction: A Review," in *WOCC/Proceedings*, 2008.

■ Implicit

■ Challenge-based entertainment modeling through:

- RL (Andrade et al., 2005): Dynamic game balance;
- GAs (Verma and McOwan, 2005);
- Probabilistic models (Hunicke and Chapman, 2004);
- Dynamic Scripting (Spronck et al., 2004)
- rtNEAT: Dynamic Game Balance in RTS games (Olesen et al., CIG'08)

Not cross-validated against human players yet (?).

→ Optimize Player Satisfaction

A brief review

G. N. Yannakakis, "How to Model and Augment Player Satisfaction: A Review," in *WOCC/Proceedings*, 2008.

■ Explicit

- Yannakakis and Hallam, 2007 → Real-time Neuro-Evolution for optimizing an "interest" value (prey/predator games)
- Player Modeling + Neuro-Evolution (Yannakakis and Maragoudakis, 2005)
- Gradient-search: adaptive physical interactive games (Yannakakis and Hallam, CIG'08)



How to Capture Fun?

Custom-Designed “Fun” Metrics

→ Custom-Designed “Fun” Metric

Step-by-step

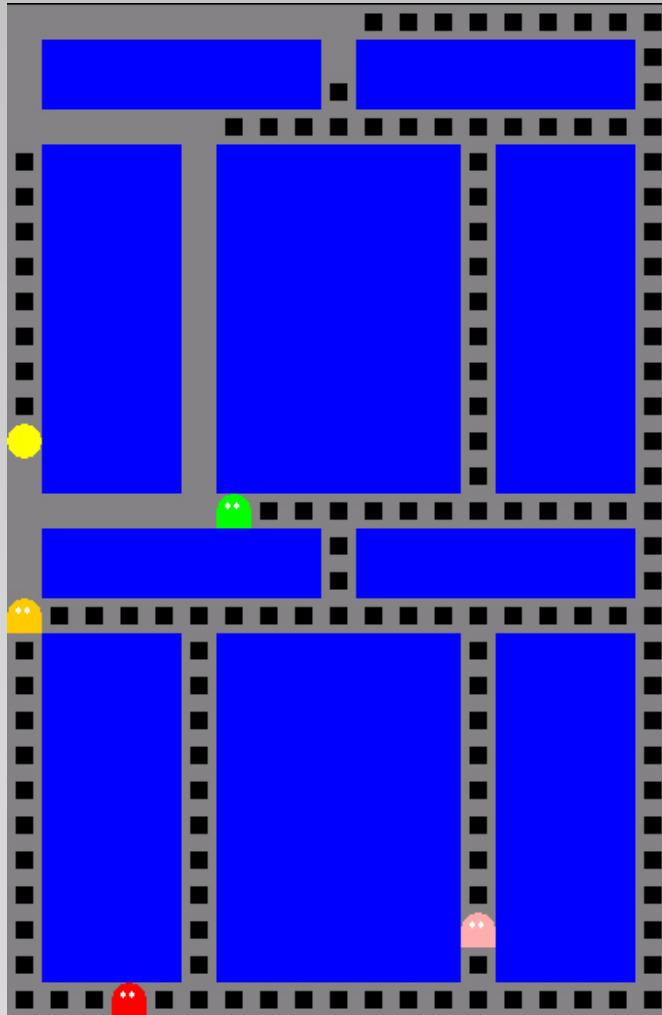
G. N. Yannakakis, and J. Hallam, "Towards Optimizing Entertainment in Computer Games," *Applied Artificial Intelligence*, 21:933-971, 2007.

- Collect criteria features
 - Ask players, visit game forums
 - **NPCs**, Storyline, Graphics, Mechanics...? What?
- Quantify criteria and combine them into a “fun” formula
- Adjust formula for meaningful “fun” values
- Devise survey experiment
 - Cross-Validate your formula against human preferences of fun
 - Does it work? → Use it as your objective function
 - It doesn't... → Get feedback, go back to your criteria

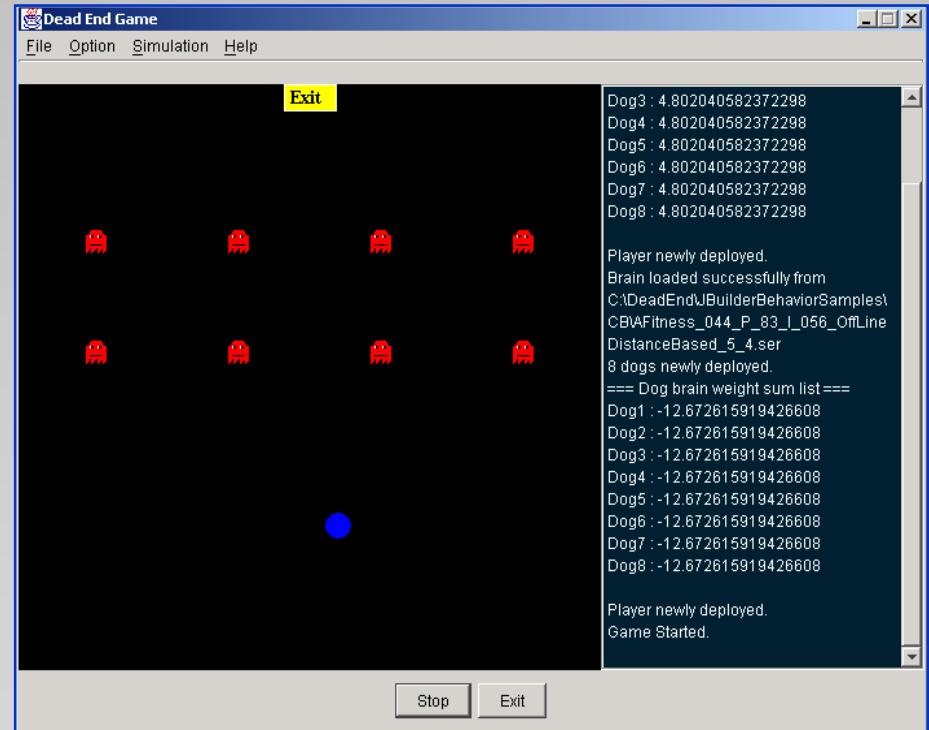
→ Case Study: Prey/Predator Games

G. N. Yannakakis, and J. Hallam, "Towards Optimizing Entertainment in Computer Games," *Applied Artificial Intelligence*, 21:933-971, 2007.

PacMan



DeadEnd



→ Case Study: Prey/Predator Games

G. N. Yannakakis, and J. Hallam, "Towards Optimizing Entertainment in Computer Games," *Applied Artificial Intelligence*, 21:933-971, 2007.

“**Interesting**” (Fun) refers (qualitatively) to interest primarily generated by **opponent** behavior

Graphics, Physics, Sound, Storyline,... etc.?

T to measure **appropriate level of challenge**

S to measure **behavior diversity**

H to measure **spatial diversity**

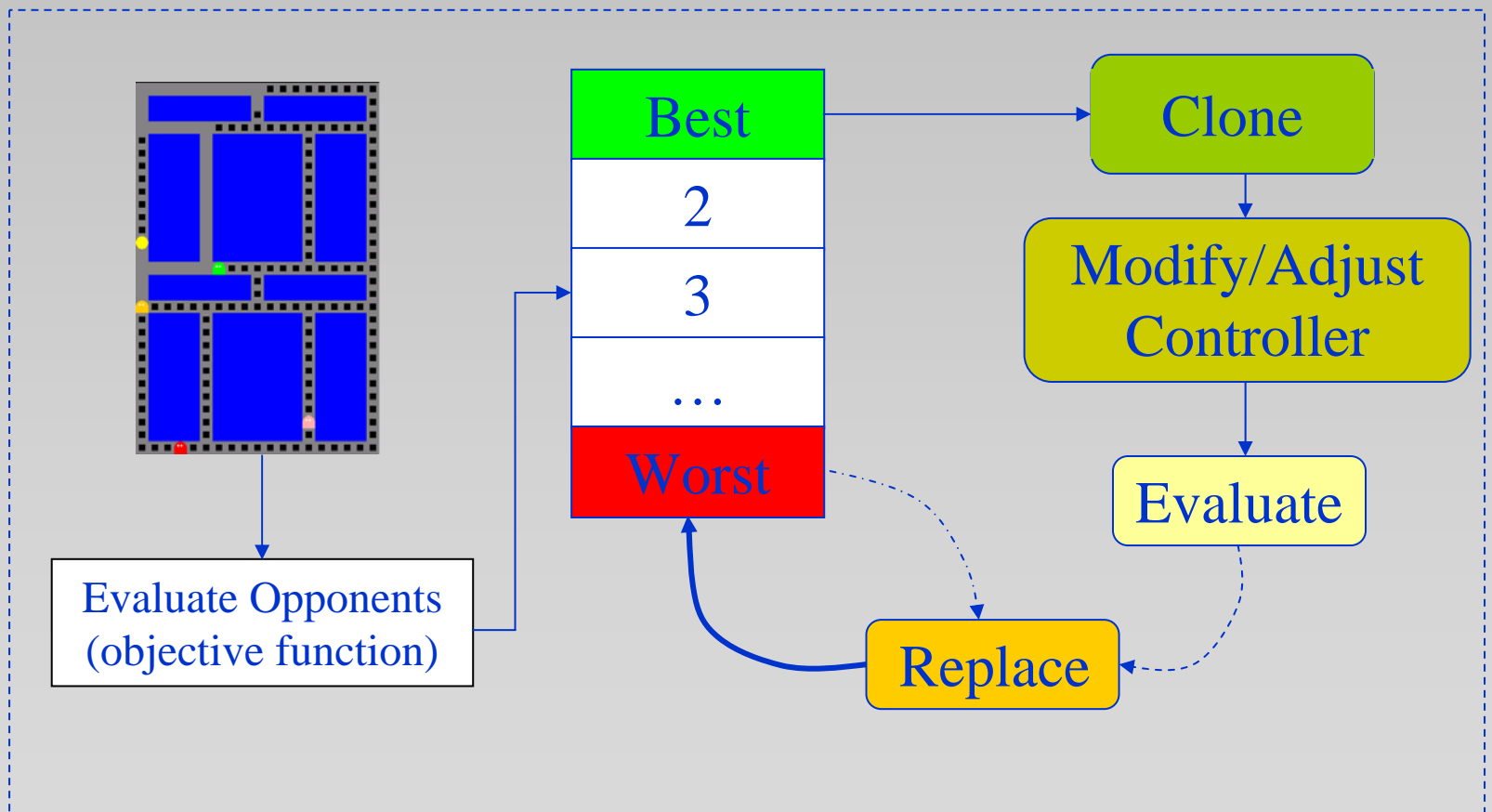
Real-Time Interest Metric (Cross-validated against Humans; $r = 0.4444$, p-value = $1.17 \cdot 10^{-8}$):

$$I = \frac{\gamma T + \delta S + \varepsilon H}{\gamma + \delta + \varepsilon}$$

→ Learning in Real-Time

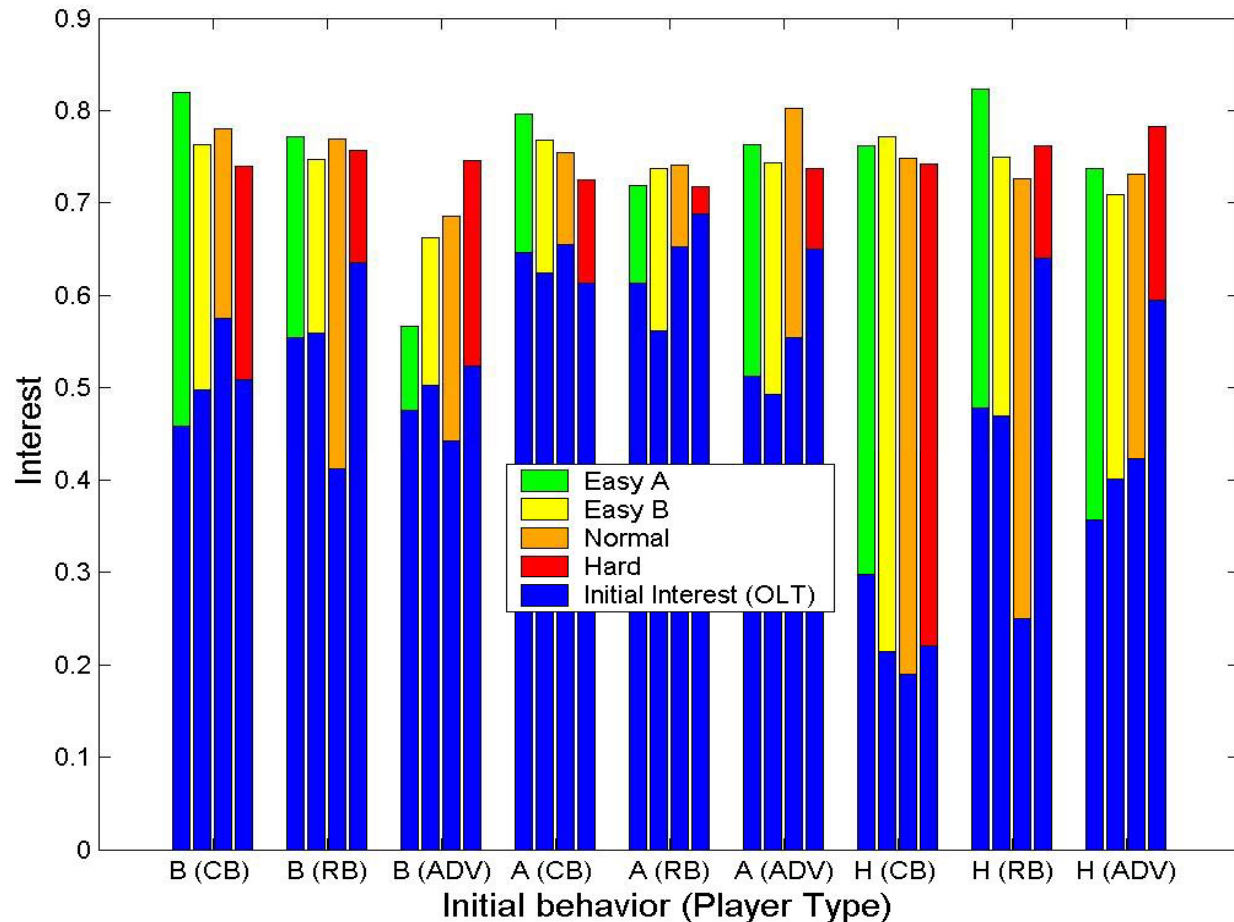
G. N. Yannakakis, and J. Hallam, "Towards Optimizing Entertainment in Computer Games," *Applied Artificial Intelligence*, 21:933-971, 2007.

Every n ticks – While the game is Played



→ Learning in Real-Time: Pac-Man

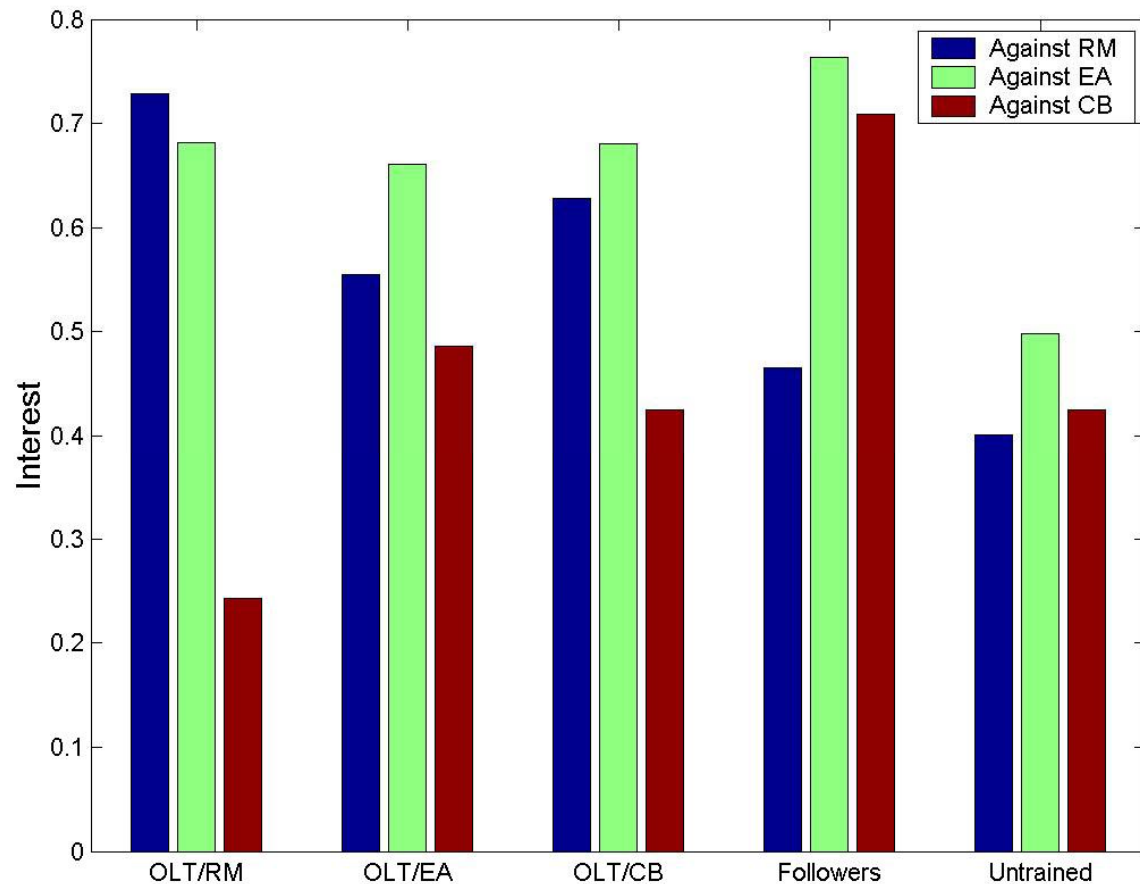
G. N. Yannakakis, and J. Hallam, "A Generic Approach for Obtaining Higher Entertainment in Predator/Prey Computer Games," *Journal of Game Development*, vol. 1, issue 3, 23-50, 2007.



→ Learning in Real-Time: Dead-End

G. N. Yannakakis, and J. Hallam, "A Generic Approach for Obtaining Higher Entertainment in Predator/Prey Computer Games," *Journal of Game Development*, vol. 1, issue 3, 23-50, 2007.

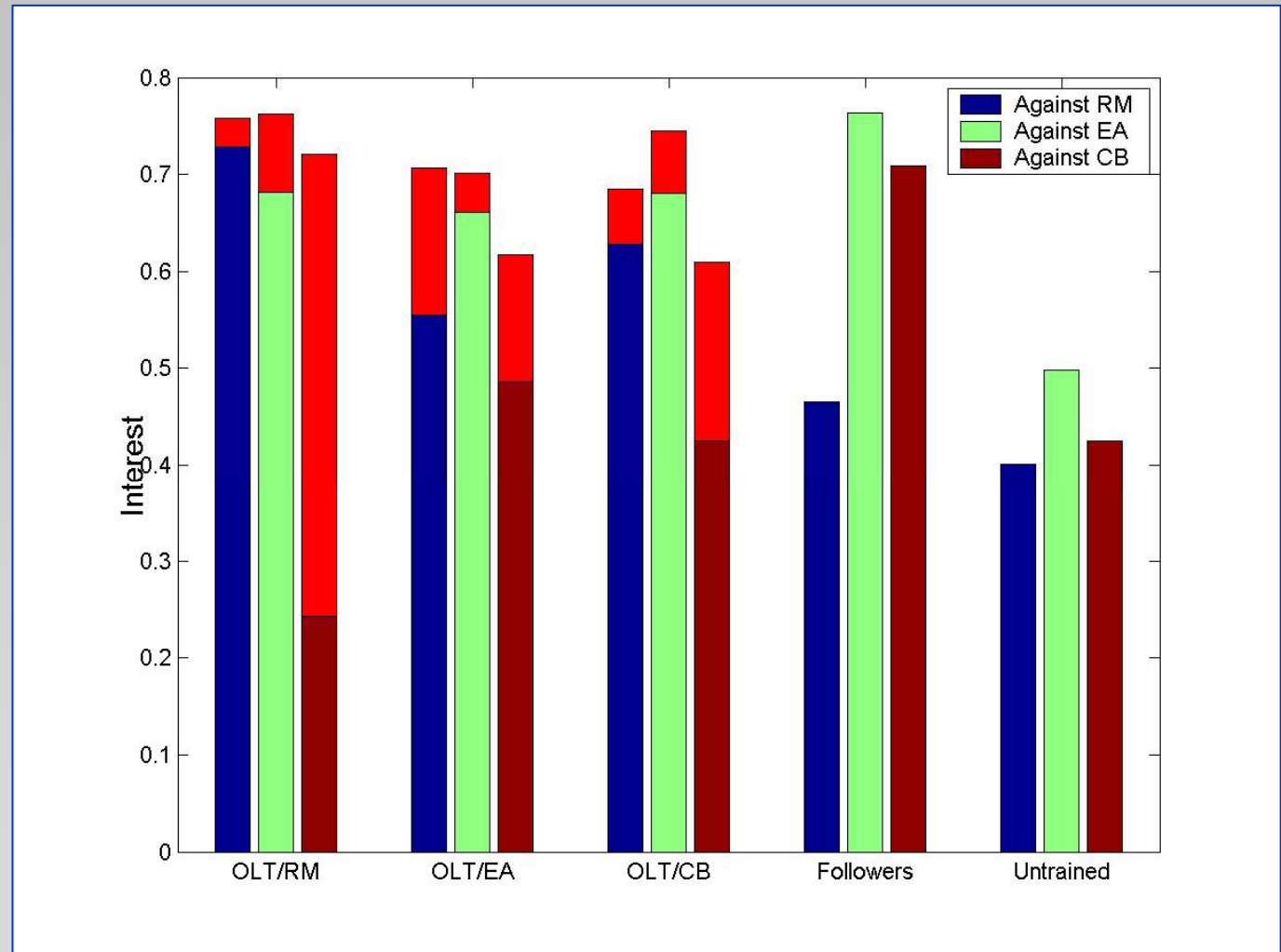
Learning Off-Line



→ Learning in Real-Time: Dead-End

G. N. Yannakakis, and J. Hallam, "A Generic Approach for Obtaining Higher Entertainment in Predator/Prey Computer Games," *Journal of Game Development*, vol. 1, issue 3, 23-50, 2007.

Learning On-Line





How to Capture Fun?

Machine Learning

→ Machine Learning Approach

Step by step guidelines for Entertainment Modeling

- Before you even start...
 - Go through psychological/qualitative fun approaches
 - Provide quantitative measurements of entertainment for qualitative factors (e.g. Malone's **Challenge** and **Curiosity**)
 - Generate game variants
 - Devise survey experiment for effective and “clean” data collection
 - Be aware of all those experiment effects (cultural, day-dependency, order of play)

→ Machine Learning Approach

Step by step guidelines for Entertainment Modeling

- Within your survey experiment
 - Extract player features (as many as possible) through
 - Game-player interaction
 - Physiology
 - Speech
 - ...
 - Question subject preferences on game variants (remember: **always compare** for reliable cognitive models!)

→ Machine Learning Approach

Step by step guidelines for Entertainment Modeling

- After your experiment
 - Feature Selection
 - Preference Learning
 - **Neuro-evolution**
 - Large Margin classifiers
 - Bayesian Learning
 - Built a model (function) of user/game characteristics and user preferences of fun
 - Enhance/Optimize entertainment value in real-time based on that model

→ Tools

Which Features though? → Feature Selection

- What does feature selection do ?
 - appropriate input vector of the model
- Selection methods
 - n-Best feature selection (nBest)
 - Sequential Forward Selection (SFS)
 - Fisher Projection
 - ...

→ Tools

Preference Learning

■ Preference Learning

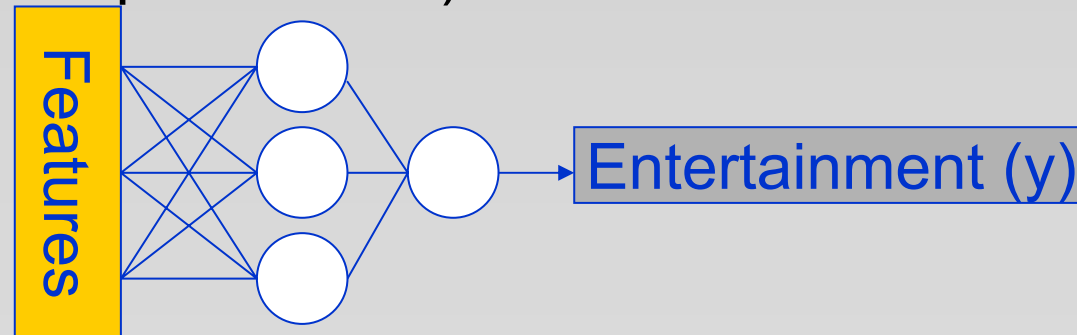
Learn the mapping between selected feature subset and entertainment preferences.

(Assumption: y of a given game is an unknown function of player features-characteristics)

■ Neuro-Evolution: GA shapes

- ANNs (connection weights)

- Fuzzy-NN (rule weights and membership function parameters)



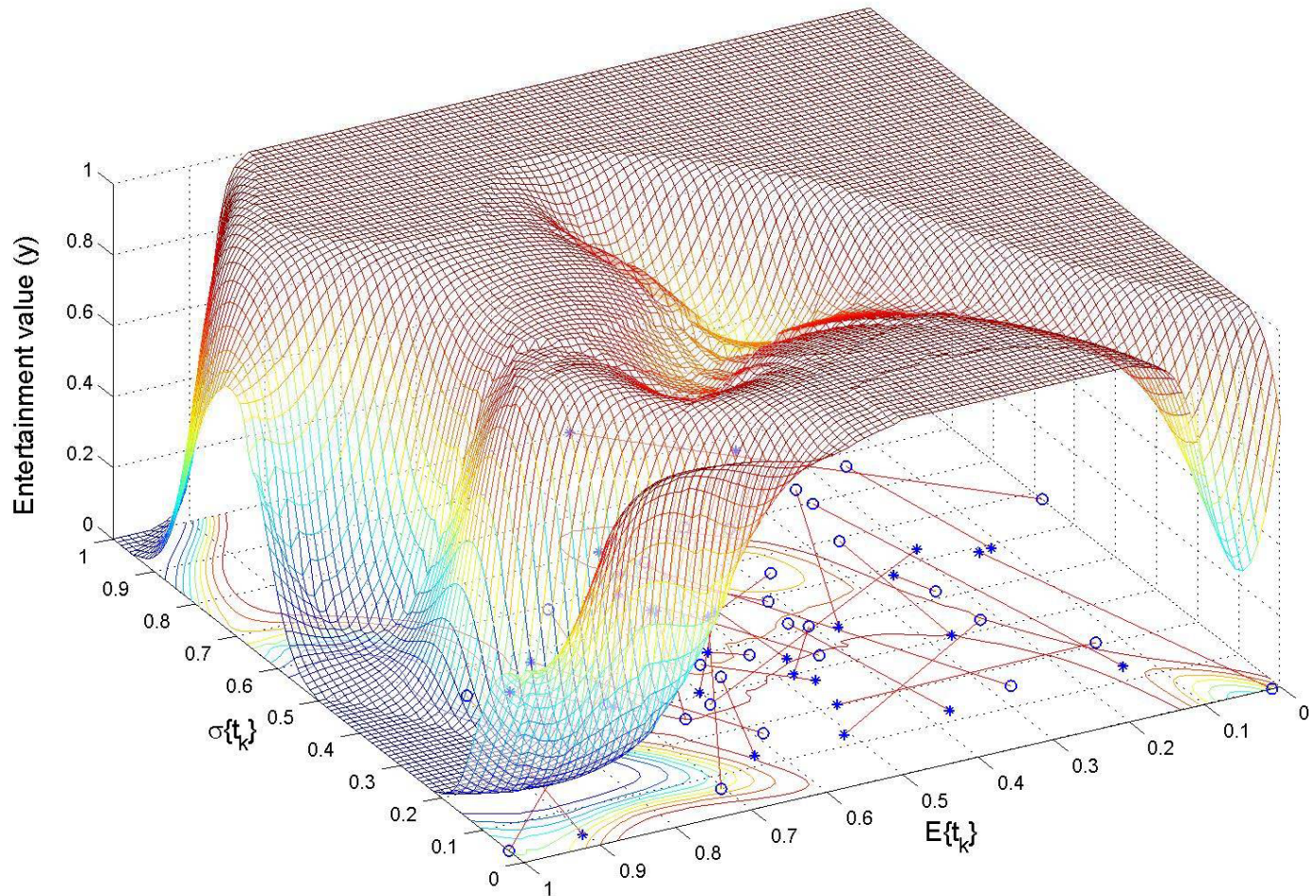
→ Case Study: Pac-Man

G. N. Yannakakis, and J. Hallam, "Modeling and Augmenting Game Entertainment through Challenge and Curiosity," *International Journal on Artificial Intelligence Tools*, vol. 16, issue 6, pp. 981-999, December 2007.

- 30 subjects (44% female, 56% male; 90% Danish, 10% Greek)
- Each subject plays two variants of the game (A and B)
- Each subject is asked which on of the two (A or B) was more **fun** to play (2-alternative forced choice).

→ Case Study: Pac-Man

G. N. Yannakakis, and J. Hallam, "Modeling and Augmenting Game Entertainment through Challenge and Curiosity," *International Journal on Artificial Intelligence Tools*, vol. 16, issue 6, pp. 981-999, December 2007.



→ Conclusions

G. N. Yannakakis, and J. Hallam, "Modeling and Augmenting Game Entertainment through Challenge and Curiosity," *International Journal on Artificial Intelligence Tools*, vol. 16, issue 6, pp. 981-999, December 2007.

- Modelling and Enhancement of Player Satisfaction:
 - **Possible** for **simple** 2D arcade games
- Better use ML rather than custom-design of “fun”
 - The ANN model gets closer to the human notion of entertainment ($r = 0.5432$, $p\text{-value} = 3.89 \cdot 10^{-12}$)



From Screen to Physical Interaction

→ Physical Interactive Games

Computer Games: Rich form of Interactivity, roles in a virtual (fantasy) world

Traditional play: physical activity, socialization, freedom: new games and rules.

Exertainment:
Advantages from both worlds, play out of screen, partly solve reported health (obesity) problems..?



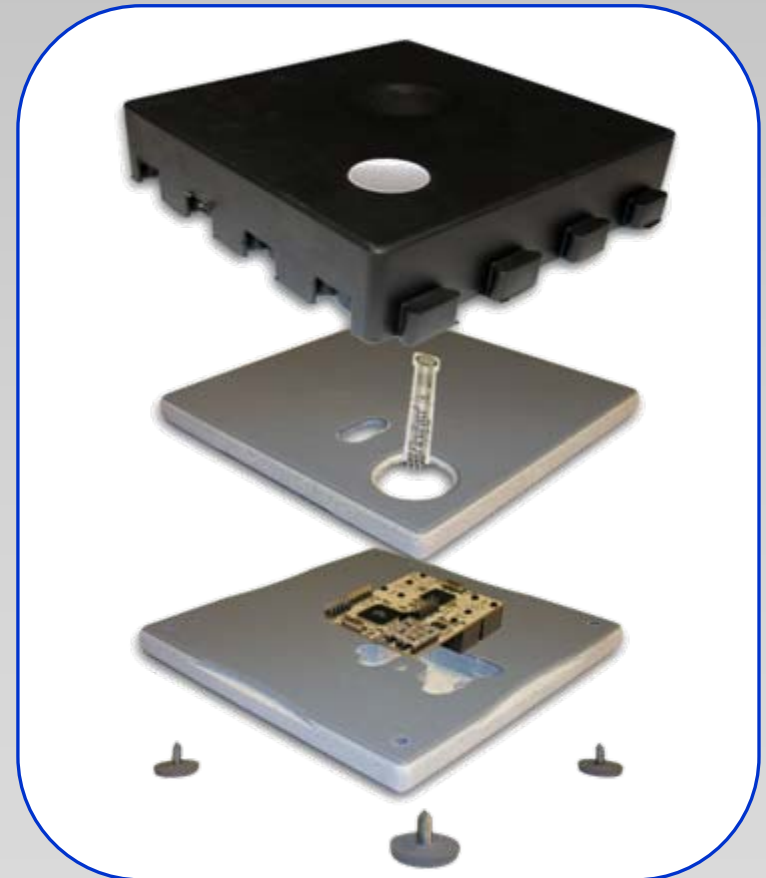
→ Playware Playground

What is it?

New game generation: adaptive, intelligent interactive physical playground (augmented-reality)

Theoretical background:
Embodied AI (Brain-Body)

Plastic Tiles (Building blocks): CPU, input (FSR sensor), output (LEDs, sound), communication



→ Test-bed games

Bug-Smasher / Space-Invaders

■ Bug-Smasher

Topology: 6 x 6



■ Space-Invaders

Topology: 10 x 5



→ Experiment

Protocol

- 72 Danish, normal-weighted children (Age Group: 8-10)
- Each child plays a pair of game variants (*A* and *B*).
- The child is asked whether *A* or *B* was more “*fun*”

Naive interviewer, no interviewed questions →
minimization of interviewing effects.

Order of play effects? No!

→ Features Extracted

- **Game** (controllable) Features [2]:
 - Challenge (S)
 - Curiosity (H)
- **Player** (personalized) Features [9]:
 - Based on 3 measurable features (child-game interaction):
 - **State** (position and LEDs color) of a pressed tile
 - **Time** that a tile-press event took place
 - **Pressure force** on a pressed tile

→ Results

G. N. Yannakakis and J. Hallam, "Game and Player Feature Selection for Entertainment Capture," in *Proceedings of the IEEE Symposium on Computational Intelligence and Games*, pp. 244-251, Hawaii, USA, 2007.

n-Best Feature Selection		Sequential Forward Selection	
Feature Subset	Validation (%)	Feature Subset	Validation (%)
Average response time	62.22	Average response time	62.22
Variance of response times	58.88	Variance of pressure forces	67.77
Variance of pressed tile-bugs distances	44.44	Curiosity (H)	68.88
No. of interactions	46.67	No. of interactions	77.77
Curiosity	52.22	Variance of response times	63.33

p-value=0.0019

➔ Real-time Entertainment Augmentation

G. N. Yannakakis, and J. Hallam, "Real-time Adaptation of Augmented-Reality Games for Optimizing Player Satisfaction," in *Proceedings of the IEEE Symposium on Computational Intelligence and Games*, Perth, December, 2008.

- Use this model to...
- ...**adjust** opponents (e.g. bugs) according to the playing style/preferences of each player:
 - Gradient-ascent $\partial y / \partial H$
 - Adaptive game: Simple rule-based system for adjusting H every 15"
 - 24 Subjects
 - "Fun" Comparison between static and adaptive game variants
 - Children prefer the adaptive over the static game in **76%** of game comparisons



Entertainment Modeling: Going physiological

→ Physiology of Entertainment...?

Entertainment is a complex mental process. However, some of its elements (sympathetic arousal) can be measured through physiological indices:

- Heart Rate
- Skin Conductance
- Blood Volume Pulse/Photoplethysmography
- Skin temperature
- Jaw-Electromyography
- ...

→ Physiology of Entertainment

Heart Rate Experiment

G. N. Yannakakis, J. Hallam and H. H. Lund, "Entertainment Capture through Heart Rate Activity in Physical Interactive Playgrounds," *User Modeling and User-Adapted Interaction, Special Issue on Affective Modeling and Adaptation*, vol. 18, no. 1-2, pp. 207-243, February 2008.

- Extracted features from **56** children
 - Games: Bug-Smasher + Space Invaders
 - Heart Rate (HR)
- 14 features** in total..

→ Results

Heart Rate Experiment

G. N. Yannakakis, J. Hallam and H. H. Lund, "Entertainment Capture through Heart Rate Activity in Physical Interactive Playgrounds," *User Modeling and User-Adapted Interaction, Special Issue on Affective Modeling and Adaptation*, vol. 18, no. 1-2, pp. 207-243, February 2008.

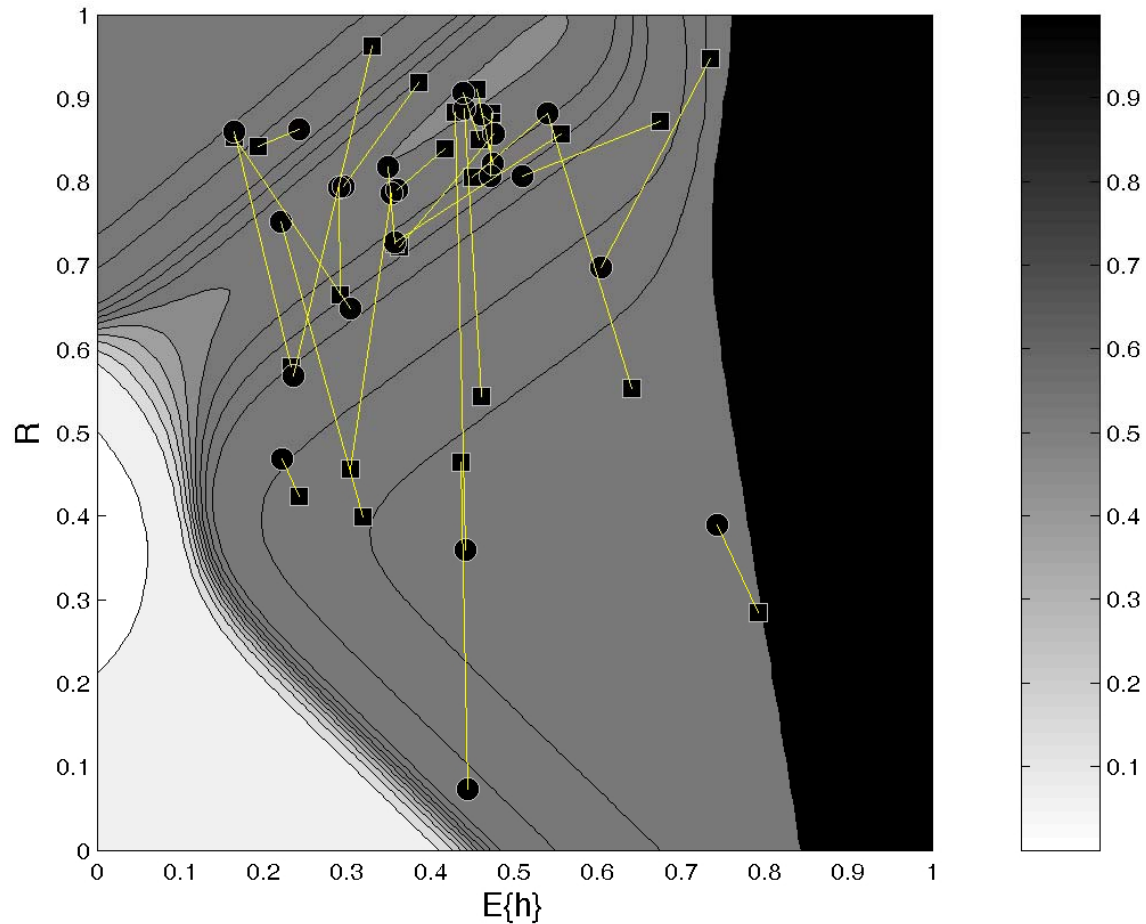
n-Best Feature Selection		Sequential Forward Selection	
Feature Subset	Validation (%)	Feature Subset	Validation (%)
Correlation (R)	72.00	Correlation (R)	72.00
Linear slope	70.66	Average HR ($E\{h\}$)	76.00
Average HR	72.00	Max{HR}-min{HR}	74.66

p-value=0.0014

→ Results

Hear Rate Experiment

G. N. Yannakakis, J. Hallam and H. H. Lund, "Entertainment Capture through Heart Rate Activity in Physical Interactive Playgrounds," *User Modeling and User-Adapted Interaction, Special Issue on Affective Modeling and Adaptation*, vol. 18, no. 1-2, pp. 207-243, February 2008.



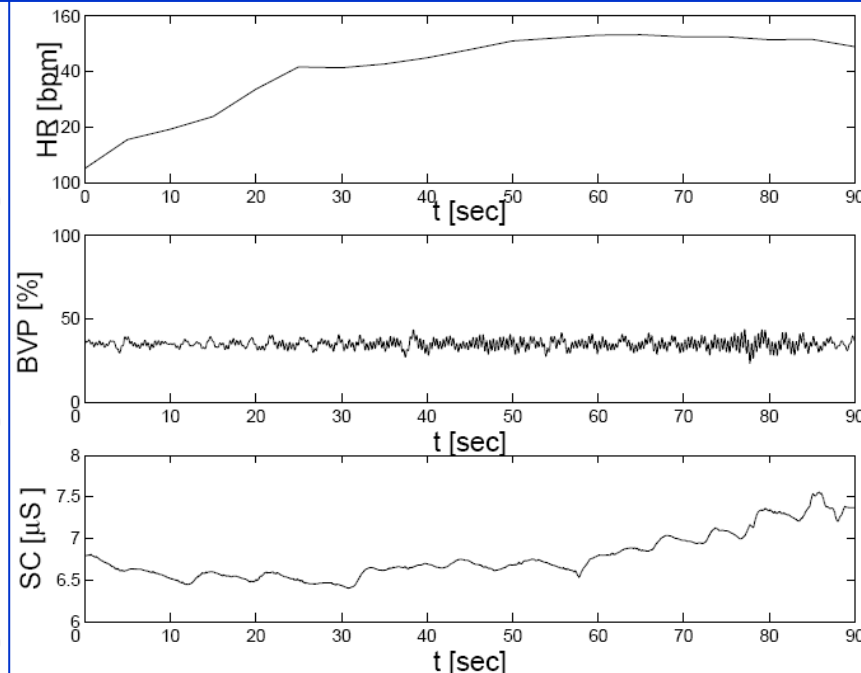
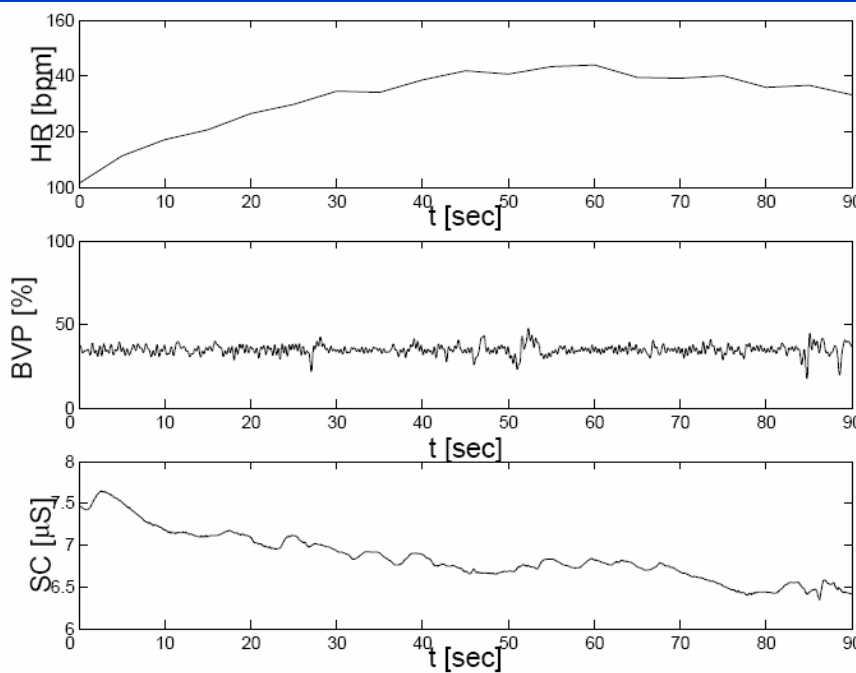
→ More Physiological Signals?

G. N. Yannakakis, and J. Hallam, "Entertainment Modeling through Physiology in Physical Play," *International Journal of Human-Computer Studies* (to appear)

- Extracted features from **72** children
 - Heart Rate (HR), Blood Volume Pulse (BVP), Skin Conductance (SC)
- 44 features** in total...

Non-preferred game

Preferred Game



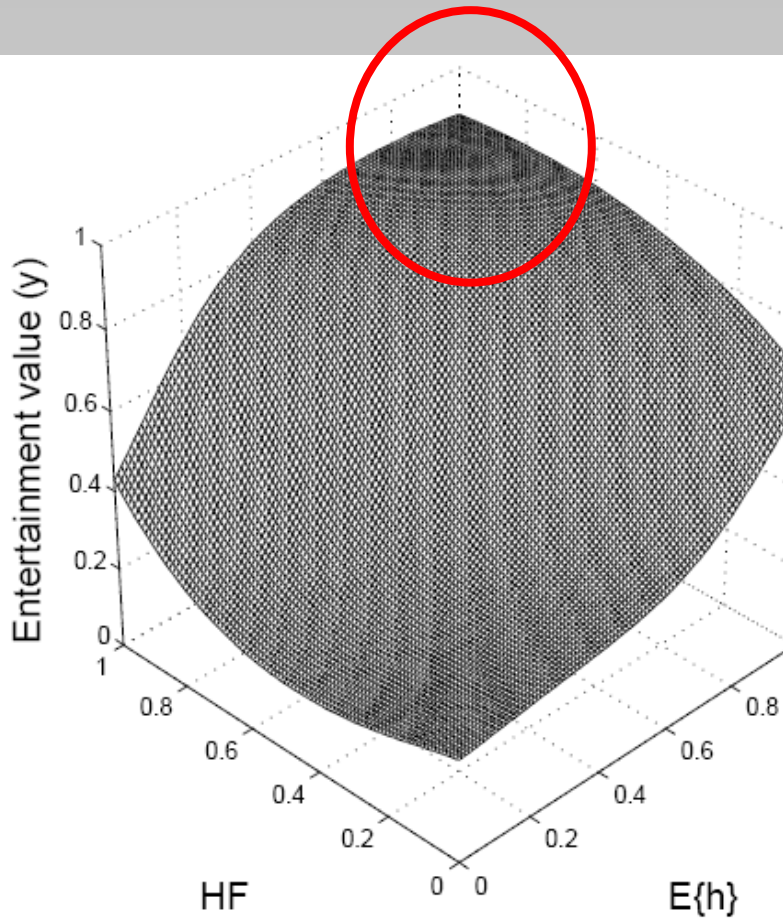
→ Results

G. N. Yannakakis, and J. Hallam, "Entertainment Modeling through Physiology in Physical Play," *International Journal of Human-Computer Studies* (to appear)

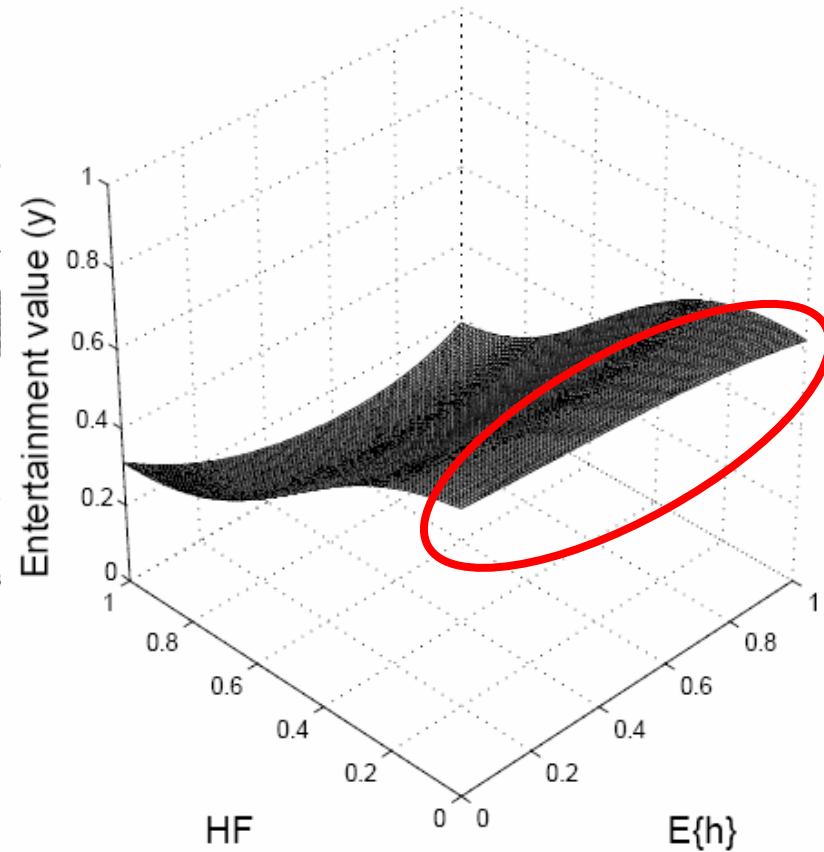
- Selected features:
 - High frequency energy of HRV (HF): mental or emotional load/effort (parasympathetic CNS)
 - average heart rate ($E\{h\}$): physical effort
 - standard deviation of RR intervals ($\sigma\{RR\}$): uniformity of heart pulses
- Model's accuracy: **79.76%** (3-fold CV)

→ Results

G. N. Yannakakis, and J. Hallam, "Entertainment Modeling through Physiology in Physical Play," *International Journal of Human-Computer Studies* (to appear)



(a) $\sigma\{RR\} = 0.0$



(d) $\sigma\{RR\} = 1.0$

→ Conclusions

Child-Platform Interaction Data

G. N. Yannakakis and J. Hallam, "Game and Player Feature Selection for Entertainment Capture," in *Proceedings of the IEEE Symposium on Computational Intelligence and Games*, pp. 244-251, Hawaii, USA, 2007.

G. N. Yannakakis, and J. Hallam, "Real-time Adaptation of Augmented-Reality Games for Optimizing Player Satisfaction," in *Proceedings of the IEEE Symposium on Computational Intelligence and Games*, Perth, December, 2008 (to appear).

- **ANN:** 3-fold cross-validation accuracy **77.77%** due to experimental noise (questionnaires, hardware failure).
- Indications that even simple gradient-ascent augments “fun” in real-time
- **Generality...**
 - **Results:** Playware action games
 - **Approach:** any computer game

→ Conclusions

Physiology

G. N. Yannakakis, and J. Hallam, "Entertainment Modeling through Physiology in Physical Play," *International Journal of Human-Computer Studies* (to appear)

- ANNs: successful predictors of children's reported entertainment grounded on physiology
- There exist features ($E\{h\}$, HF) corresponding to physical activity that can effectively capture entertainment
- Isolation of those features is **possible** in physical games

→ Key/Open Research Questions

- Strong evidence already exist; however...
 - **Generalization (different scales):** More complex environments – commercial-standard computer games.
 - **Generalization (different modes):** Design and implementation of adaptive human-centered systems of rich HCI (exertainment, edutainment, adaptive Web).
 - **Real-time augmentation** of ‘entertainment value’ of HCI systems by adjusting ***opponents***:
 - Various levels of NPC control
 - Content creation
 - Storyline/Narrative
 - Game/interface design
 - ...



Part II: Game Design

→ Part II: Game Design

■ Overview:

- Why should we try to design (aspects of) games automatically?
- How can we create predictors of player satisfaction?
 - Static approaches
 - Dynamics approaches
- Which aspects of games can we optimize?
 - Environments
 - Narratives
 - Rules and parameters

→ Why automate game design?

- It's an interesting research problem
 - interdisciplinary: optimization, supervised learning, game studies, psychology...
 - not much research done yet!
- Could save game developers money
 - Large parts of game budgets go into creating environments, tuning parameters
- Could enable new types of games
 - adaptive content creation
 - evolution might be a radical designer

→ Automatic content creation / game design: general idea

- Use optimization algorithms (e.g. evolution) to optimize some aspect of a game, not necessarily the agent
 - keep the rest of the game similar, while changing the aspect that's being optimized
- Objective: make the game more fun
 - we need a measure of fun

→ Predicting player satisfaction

- The big problem: we want to optimize (aspects of) games to be fun, but how do we know what is fun?
 - i.e. where's our fitness function?
- Using human players? (interactive)
 - takes too much time (during optimization)
 - humans don't want to play boring games (low-fitness solutions)
- Need a predictor of fun/satisfaction

→ Criteria for a predictor

- Accurate, i.e.
 - theoretically well-founded or understood
 - or empirically validated
- Fast!
- Personalizable (preferably)
- Generalizable over different types of games and content (preferably)

→ Player satisfaction prediction: overall idea

- Create a game-playing agent (NPC AI)
 - hard-coded, or
 - through some learning algorithm
 - maybe as a model of a human player
- Let the agent play a game
 - judge how much “fun” the agent had according to some theory of fun, and the behaviour of the agent
 - if the theory is right, and if the agent plays like a human, the predictor is accurate

→ Approaches to predictor design: Static fun predictors

- Assume the agent (player model) does not learn
- Based on e.g. Malone's factors
- Very often focuses on challenge - the game should not be too easy (too hard?)
 - i.e. how well does the agent play the game?
- Many similar measures possible:
 - variation in performance
 - diversity in behavior, locations visited etc.
- Georgios' model for Pacman (slide 20)

→ Approaches to predictor design: Dynamic fun predictors

- Assume the agent learns, measure learning progress / learnability
- Raph Koster: learning = fun
 - A game is fun to the extent you learn while playing
- Juergen Schmidhuber: curiosity = prediction progress
 - a curious agent chooses to explore areas which it can learn about

→ What's in a game?

- Agent controllers (NPC AI)
 - this is what most CIG research is about, though usually not with an *explicit* player satisfaction perspective
- Environments
 - levels, tracks, maps, cities...
- Narrative
- Rules and parameters
- Artwork

→ Criteria for a representation

- Expressive (should be able to express good content / rules)
- Adaptive (induces a smooth search space, so can be used with evolution)
- Human-readable (can be further edited by humans)

→ Environments: mazes

- Representation: placement, orientation and lengths of walls in a grid world
- Static fitness function: maximize minimum number of turns and path length taken by agent (as determined by dynamic programming) in order to reach goal from start

– D. Ashlock, T. Manikas and K. Ashenayi, *Evolving A Diverse Collection of Robot Path Planning Problems*. CEC 2006

→ Environments: mazes

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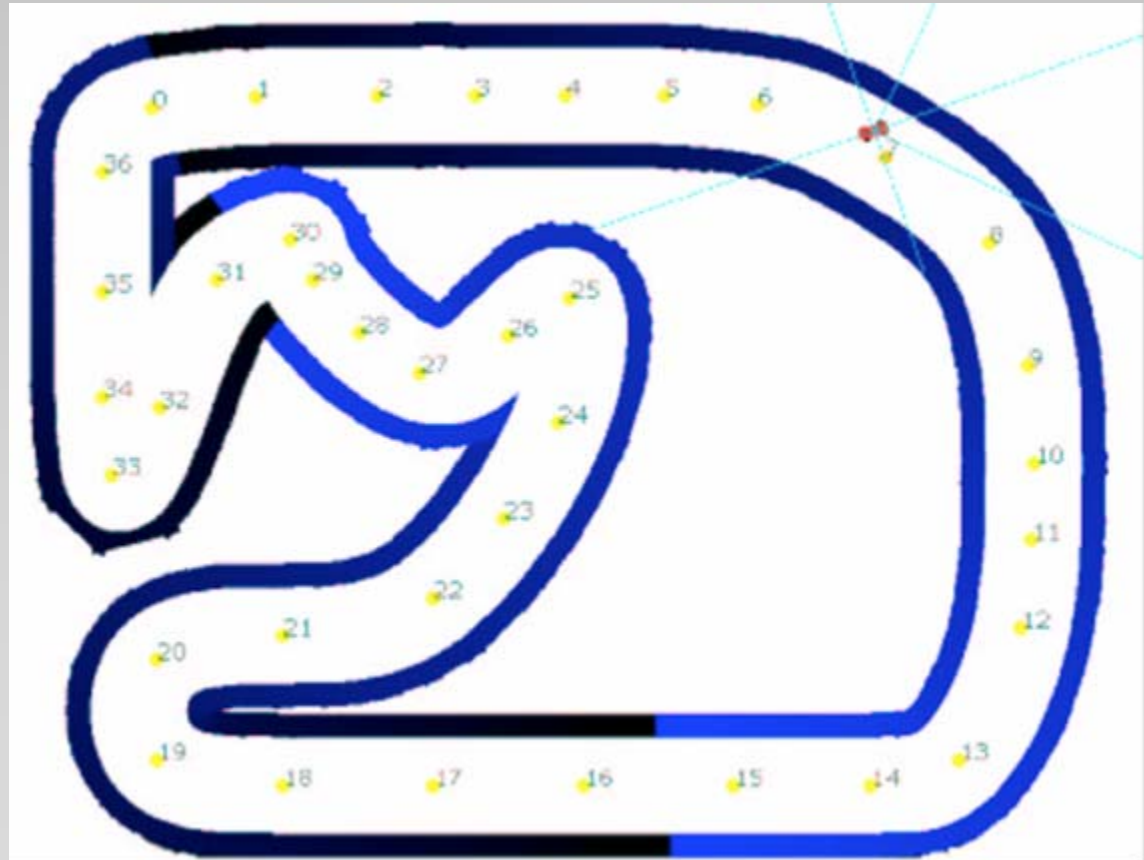
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→ Environments: racing tracks

- Static fitness function:
 - loosely based on Malone
 - right amount of progress on track, maximize variation in progress, maximize difference between max and mean speed
- Representation:
 - b-splines (sequences of Bezier curves)
 - Julian Togelius, Renzo De Nardi and Simon Lucas.
Towards automatic personalized content creation for racing games. CIG 2007

→ Environments: racing tracks

- First, create a player model through letting a human drive on a test track, and evolving a controller that mimics the driving style of the human



→ Environments: racing tracks



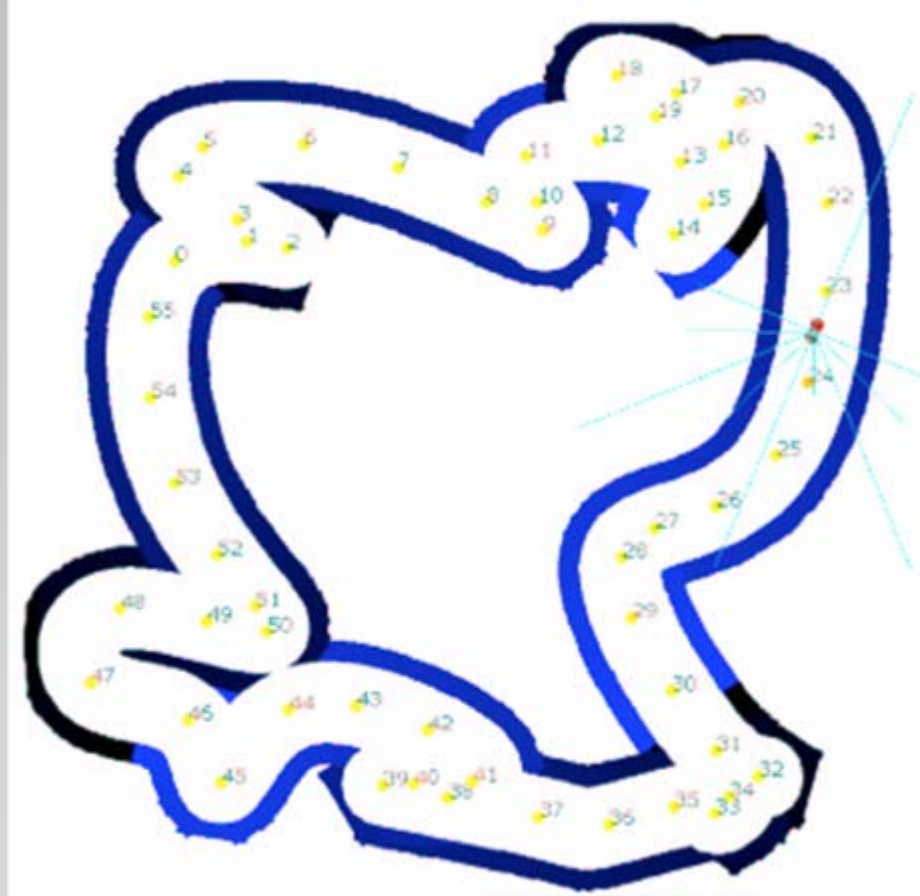
Track evolved for me

→ Environments: racing tracks



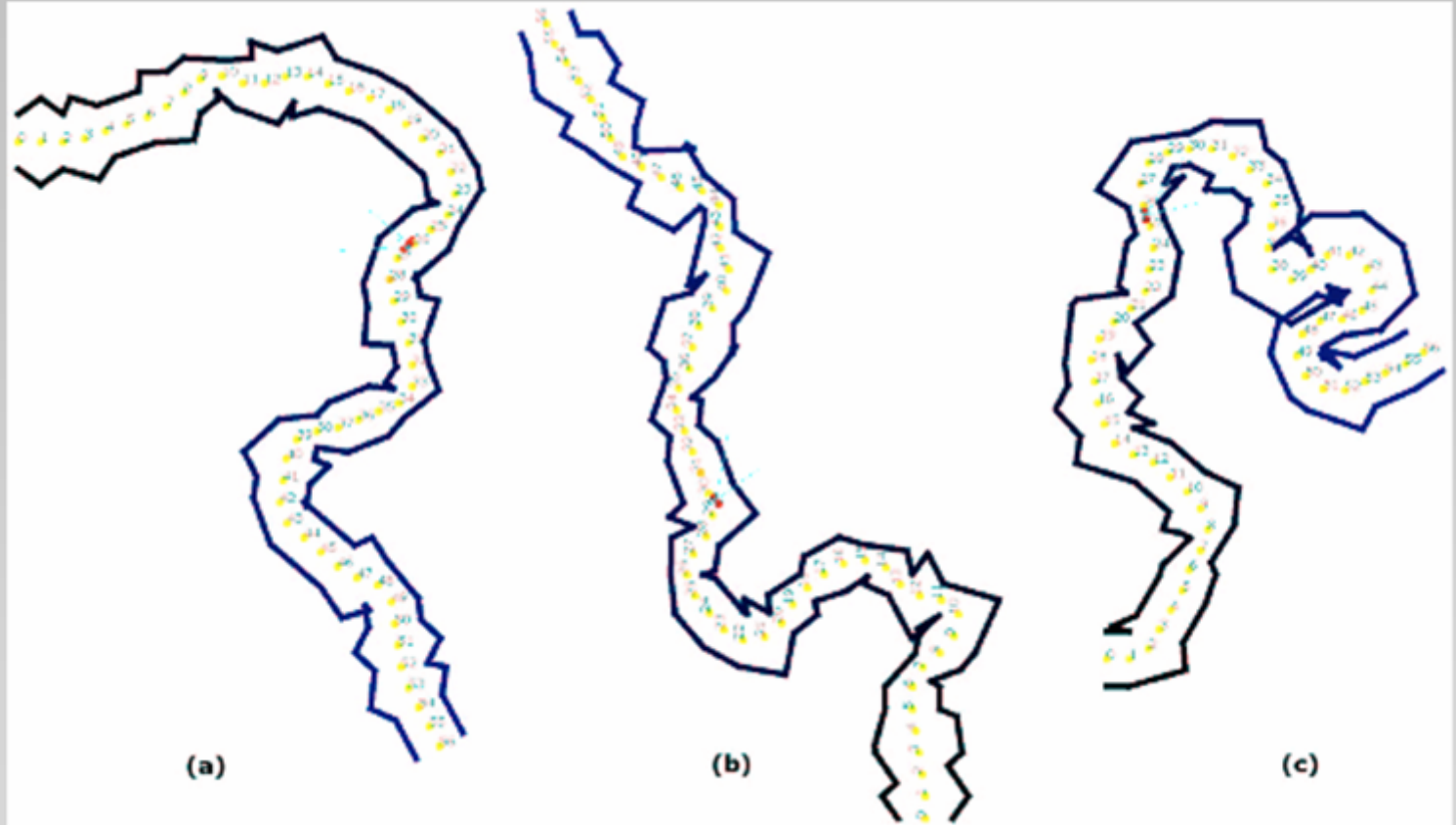
Track evolved for Renzo

→ Environments: racing tracks



Track evolved with random initialization

→ Environments: racing tracks



Tracks evolved with segment representation

→ Environments: cities

- Fitness function: none! (not used for player satisfaction optimization yet)
 - but imagine escape routes from a bank robbery in GTA...
- Representation: recipe for building the city procedurally
 - procedural representation needed to keep search space manageable (cities are big!)
 - » George Kelly, Hugh McCabe. *Citygen: An Interactive System for Procedural City Generation*. GDTW 2007

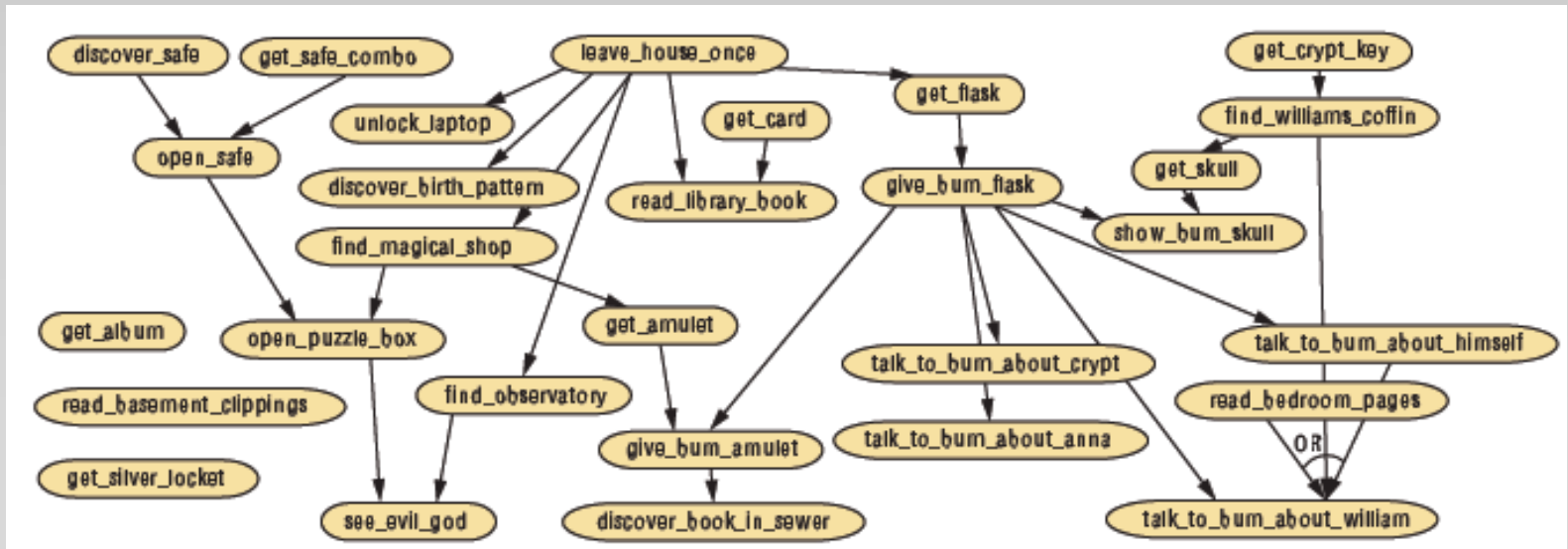
→ Narrative

- Fitness function: measures of the behaviour of an e.g. random agent
 - location flow (successive events at the same place)
 - thought flow (events related conceptually)
 - motivation (related causally)
 - » Mark Nelson, Michael Mateas, David Roberts and Charles Isbell, “*Declarative Optimization-Based Drama Management in the Interactive Fiction Anchorhead.*” IEEE Computer Graphics and Applications, vol 26, number 3, 2006, pp 32-41.

→ Narrative

■ Representation:

- deniers, causers, hints and game endings that affect a player's progression through a story



→ Rules and parameters: Board games

- Static fitness function: results of playing game with Alfa-beta search
 - Completion: most games reach a conclusion
 - Balance: no advantage to either player
 - Advantage: no first move advantage
 - Duration: games end in a reasonable number of moves
 - Additional “aesthetic measurements”
 - » Cameron Browne, *Automatic Generation and Evaluation of Recombination Games* (PhD Thesis), Queensland University of Technology 2008

→ Rules and parameters: Board games

- Representation: The *Ludi* game description language

```
(ludeme TicTacToe
  (players White Black
    (board
      (tiling square i-nbors)
      (shape square)
      (size 3 3)
    )
    (end (All win (in-a-row 3) ) )
  )
)
```

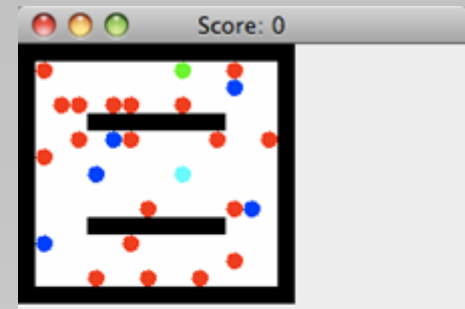
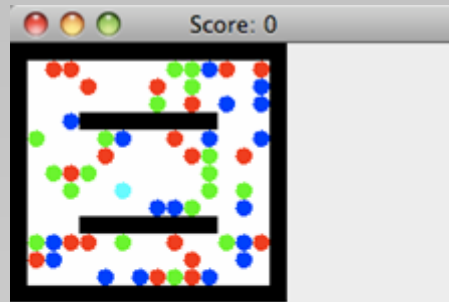
→ Rules and parameters: Agent-based games

- Dynamic fitness function: learnability of game rules
 - the game should not be too easy (winnable by a random game agent)
 - an “inner” evolutionary process should be able to learn to play the game satisfactorily
 - inspired by Koster, Schmidhuber
 - Julian Togelius and Juergen Schmidhuber. *An Experiment in Automatic Game Design*. This symposium!

→ Rules and parameters: Agent-based games

- Representation: game rules
 - number of red, green, blue things
 - movement logic for things
 - effects on things or agent when things or agents collide with each other (death, teleport etc.)
 - ending conditions (time, score)

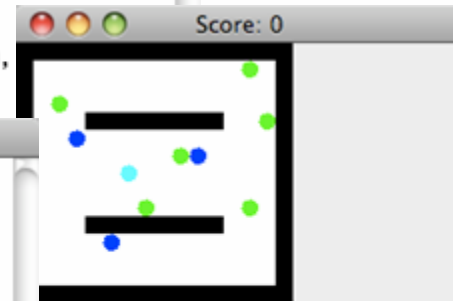
→ Rules and parameters: Agent-based games



```
Terminal — java — 62x14
julian-togelius-macbook-pro-15:classes julian$ java agd.Test
Rules:
score 10 must be reached within 85 time steps
red: 20, clockwise
green: 18, counterclockwise
blue: 18, counterclockwise
collision: Red, Green -> teleport, teleport, 0, 0
collision: Red, Blue -> death, death, -1, 0
collision: Red, Agent -> death, none, 0, 0
collision: Green, Blue -> none, death, -1, 0
collision: Green, Agent -> death, teleport, 0, 0
collision: Blue, Agent -> none, death, 1, 1
```

```
Terminal — java — 62x14
Rules:
score 5 must be reached within 93 time steps
red: 19, still
green: 1, counterclockwise
blue: 5, clockwise
collision: Red, Green -> death, teleport, 0, -1
collision: Red, Blue -> teleport, death, 0, -1
collision: Red, Agent -> death, none, 1, 0
collision: Green, Blue -> none, teleport, -1, -1
collision: Green, Agent -> death, death, -1, -1
collision: Blue, Agent -> none, none, -1, -1
```

```
Terminal — java — 62x14
julian-togelius-macbook-pro-15:classes julian$ java agd.Test
Rules:
score 5 must be reached within 51 time steps
red: 0, clockwise
green: 6, random short
blue: 3, still
collision: Red, Green -> none, none, 1, 1
collision: Red, Blue -> death, none, -1, 1
collision: Red, Agent -> death, teleport, 1, -1
collision: Green, Blue -> teleport, teleport, 0, -1
collision: Green, Agent -> death, none, -1, -1
collision: Blue, Agent -> death, teleport, -1, 0
```



→ Open research questions

- Pretty much everything!
- Does it work?
 - need empirical validation with human players
- Does it scale?
 - only relatively small spaces searched so far, for relatively simple games
- Variations: online / offline, static / dynamic, personalized / generic, invention / fine tuning etc.

→ Some obvious ideas

- Multi-objectivity
 - fun can be measured in many ways, maybe we want to optimize different experiences
 - MOEAs can explore trade-offs
- Model learning
 - dynamic predictors are based on learnability; we need a learning process that learns like a human player
- Parameters for games

→ Questions...?

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Papers in the proceedings

IEEE Task Force on Player Satisfaction Modeling
game.itu.dk/PSM/

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