

AI-Assisted Game Design: a diverse research challenge

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Outline

- The Challenge
 - “How can AI be used to assist in game design?”
- Potential Research Topics and Examples:
 - Procedural Content Generation
 - “How can AI be used to generate diverse worlds and elements therein?”
 - Game Analysis for Game Improvement
 - “How can AI techniques leverage mathematical insight from game data to improve a game’s design?”
 - AI-Assisted Game Invention
 - “How can AI search a space of possible combinations of rules to find a set with good quality play?”
 - Other applications of AI techniques to Game Design
- Conclusion

What Is AI-Assisted Game Design?

- What distinguishes **AI-Assisted Game Design (AIAGD)** from **AI game play** is that the results of an AI technique are applied to the *design of the game itself*.
- AIAGD *includes* but is not the same as Procedural Generation
- AIAGD also includes:
 - Game Analysis for Game Improvement
 - AI-Assisted Game Invention
 - Other application of AI techniques to Game Design

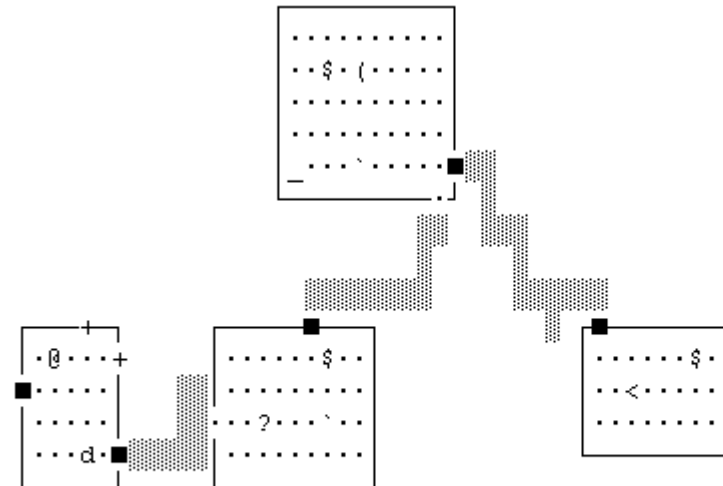
AAAI/EAAI Mentored Undergraduate Research Challenge

- Teams:
 - At least one faculty member serving as a mentor and peer-reviewer.
 - At least one undergraduate student researcher being mentored.
 - Mixed teams with graduate students are permitted.
- Submission deadline: ~September 2021 at the AAAI-22 full paper deadline
- Accepted papers will be presented at EAAI-22:
 - Collocated with AAAI-22: <https://aaai.org/Conferences/AAAI-22/>
 - Vancouver Convention Centre, Vancouver, British Columbia, Canada, February 22 - March 1, 2022

Procedural Content Generation

- Name some games you have played where the computer has generated content of the game, e.g. landscape, world elements, etc.?
- Excellent examples from Prof. Presser's colloquium on PCG:

<http://cs.gettysburg.edu/~cpresser/colloquia/pcg/PCG.pdf>



Joe the Tenderfoot St:14 Dx:9 Co:13 In:14 Wi:17 Ch:8 Chaotic
Divl:1 \$:0 HP:14(14) Pw:3(3) AC:7 Exp:1 T:82

https://en.wikipedia.org/wiki/Procedural_generation#/media/File:Nethackscreen.gif

Common Examples



Minecraft



No Man's Sky

Uncommon Example: Dwarf Fortress

- Simulates History: civilizations, wars, religion etc.



Nerfing and Buffing

- Have you ever played a game where a game element was significantly over-/under-powered?
- Have you ever had a game element “nerfed” (made weaker), or “buffed” (made stronger) in a new game version/patch?
- How has this affected your play experience?

Game Data Analysis for Game Improvement

- Discern over-/under-powered (over-/under-used) game elements to nerf/buff
 - Magic the Gathering – recent card bannings (20 in the last 2 years)
 - Fortnite – nerfing/buffing over-/under-powered weapons



Should Players be Primary Playtesters?

- While players being involved in development feedback and design is an important modern dynamic, e.g. Valheim, there are many games that are released with a feel of being unfinished or not thoroughly tested.
- There is great potential for AI players to find and exploit bugs in game physics:
 - OpenAI Hide and Seek
<https://www.youtube.com/watch?v=Lu56xVIZ40M&t=66s> (1:06 – 3:18)
 - Q*Bert: <https://youtu.be/wm8tK91k37U?t=131> (2:11 – 2:39)
- In general, AI players could preemptively expose bugs, under-/over-used game elements, and other ways in which the design is “broken”.

Red Light as AI-Assisted Improvement of Fowl Play

- T. Neller, M. Malec, C. Presser, F. Jacobs. **Optimal, Approximately Optimal, and Fair Play of the Fowl Play Card Game**, in the Proceedings of the 8th International Conference on Computers and Games CG2013
 - Resulted in approximately fair Poker chip variation “Red Light”

Fowl Play

- Designed by Robert Bushnell and published by Gamewright in 2002
- Object: Be the first of 2+ players to reach a given goal score.
 - We consider 2 players, 50 point goal case.
- Materials: Shuffled 48 card deck with
 - 42 chicken cards (each incrementing turn total)
 - 6 wolf cards (each causing loss of turn total)



Fowl Play (cont.)

- On each turn, draw one or more cards until:
 - you draw a wolf and score no points, or
 - you hold and score the number of chickens you've drawn.
- After the last (6th) wolf is drawn, reshuffle all cards.



Optimizing Fowl Play for Fairness

- Given an ability to analyze perfect play and win probabilities with Value Iteration, we can
 - vary the number of good (chicken) and bad (wolf) outcomes,
 - compute optimal play for each game,
 - compute komi optimizing fairness, and
 - thus optimize the game design for fairness.
- With our optimized game of Red Light, the first player wins 50.001% assuming optimal play.

Red Light

- Object: Be the first of 2 players to reach 50 points. The 2nd player begins with 1 point.
- Materials: 28 poker chips in a bag with
 - 24 green “green light” chips (each incrementing turn total)
 - 4 red “red light” chips (each causing loss of turn total)



Red Light (cont.)

- On each turn, draw one or more chips until:
 - you draw a red light and score no points, or
 - you hold and score the number of green lights you've drawn.
- After the last (4th) red light is drawn, all chips are returned to the bag and reshuffled.
- Alternate materials: Standard ("French") deck of playing cards using Ace (red light) and 2-7 (green lights) of each suit.



Design as Optimization: Rook Jumping Mazes

- Todd Neller, Adrian Fisher, Munyaradzi T. Choga '12, Samir M. Lalvani '13, and Kyle D. McCarty '11. [Rook Jumping Maze Design Considerations](#), in van den Herik, H. Jaap, Iida, Hiroyuki, and Plaat, Aske, eds., *LNCS 6515: Computers and Games, 7th International Conference, CG 2010, Kanazawa, Japan, September 24-26, 2010, Revised Selected Papers*, Springer, 2011, pp. 188-198.
- Some attribute the creation of the Rook Jumping Mazes (RJM) to the great puzzle innovator Sam Loyd, who published a Queen Jumping Maze allowing diagonal moves in 1898. Few maze designers have the skill to create them.
- Let's look at an example...

Example Rook Jumping Maze

- ▶ **Specification:** grid size, start state (square), goal state, jump numbers for each non-goal state.
- ▶ **Jump number:** Move *exactly* that many squares up, down, left, right. (*Not* diagonally.)
- ▶ **Objectives:**
 - Find a path from start to goal.
 - Find the shortest of these paths.

3	4	1	3	1
3	3	3	G	2
3	1	2	2	3
4	2	3	3	3
4	1	4	3	2

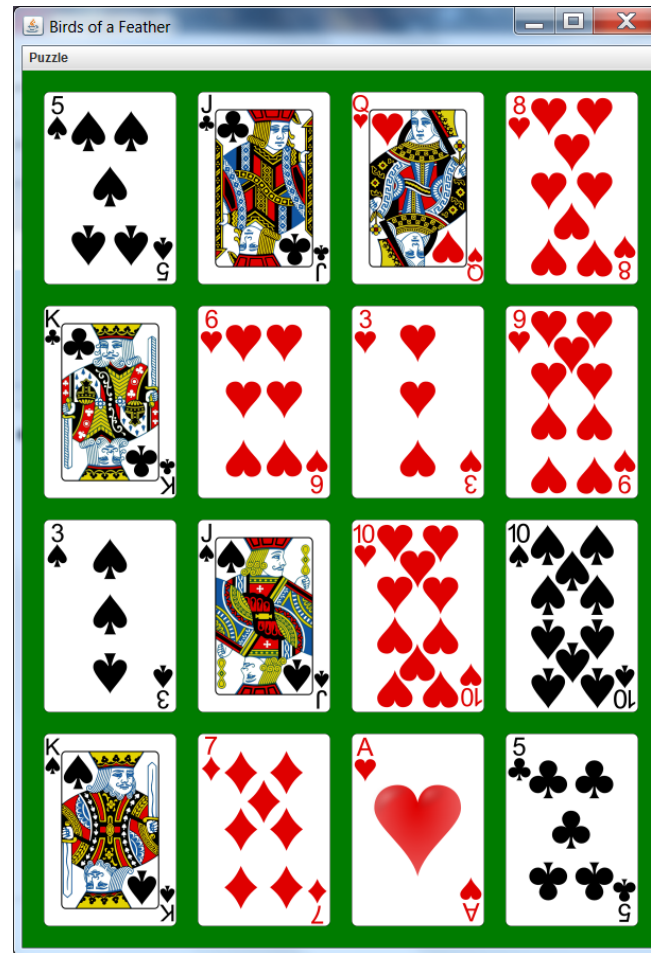
Computer Generation of RJMs

- Stochastic Local Search: Iteratively change a single random jump number, accepting improvements, and usually rejecting changes for the worse according to an “energy” measure we seek to minimize.
- Energy = puzzle badness; subjective and hard to express
- Iterative puzzle design generations suggesting refinements to “energy”:
 - Goal Reachability, Reaching States, Reachable States, Black Holes, and White Holes
 - Shortest Solution Uniqueness
 - Minimum Solution Path Length
 - Etc.

Birds of a Feather (BoaF)

- [“Birds of a feather flock together.”](#)
- Designed August 9, 2016
- **Materials:** a standard, shuffled 52-card deck
- **Setup:** Deal cards singly, face-up into a grid (e.g. 4-by-4).
- **Object:** Form a single stack of all cards.
- **Play:** A player may move one stack of cards onto another stack of cards in the *same row or column* if the cards on top of the stacks have either
 - (1) the *same suits*, or
 - (2) the *same or adjacent ranks*. Aces are low and not adjacent to kings, so rank adjacency is according the ordering A, 2, 3, ..., J, Q, K.

Birds of a Feather Example with Solution



TS-9H AH-TH AH-3H
AH-QH 6H-7D JS-JC
KS-3S KS-KC 5S-KS
6H-5C 5S-TS AH-8H
6H-AH 5S-6H 5S-JS

Generation of Difficult Interesting Boaf Deals

- Todd Neller, Daniel Ziegler '21. [Computer Generation of Birds of a Feather Puzzles](#), in the Proceedings of the 33rd AAAI Conference on Artificial Intelligence (AAAI-19), January 27 – February 1, 2019, Honolulu, Hawaii, USA, AAAI Press, Palo Alto, California, USA, pp. 9693-9699.
- Key insight: The most difficult, interesting Boaf puzzles we generated maximized puzzle tension at the halfway point to the solution.
- Many good puzzles have an “hourglass” structure of easy-tense-easy effort through the solving process, e.g. Sudoku, FreeCell, the Devil’s Claw.



Game Tension

- Game Tension, according to Cameron Browne, can be estimated as the probability that a random player will make a wrong choice.

Browne informally

described tension as follows:

Tension is related to the amount of impact that players' choices have on the outcome of the game.

Browne further proposes a simple metric for estimating the tension T in puzzle state S :

$$T(S) = 1 - \frac{M_{S_w}}{M_S}$$

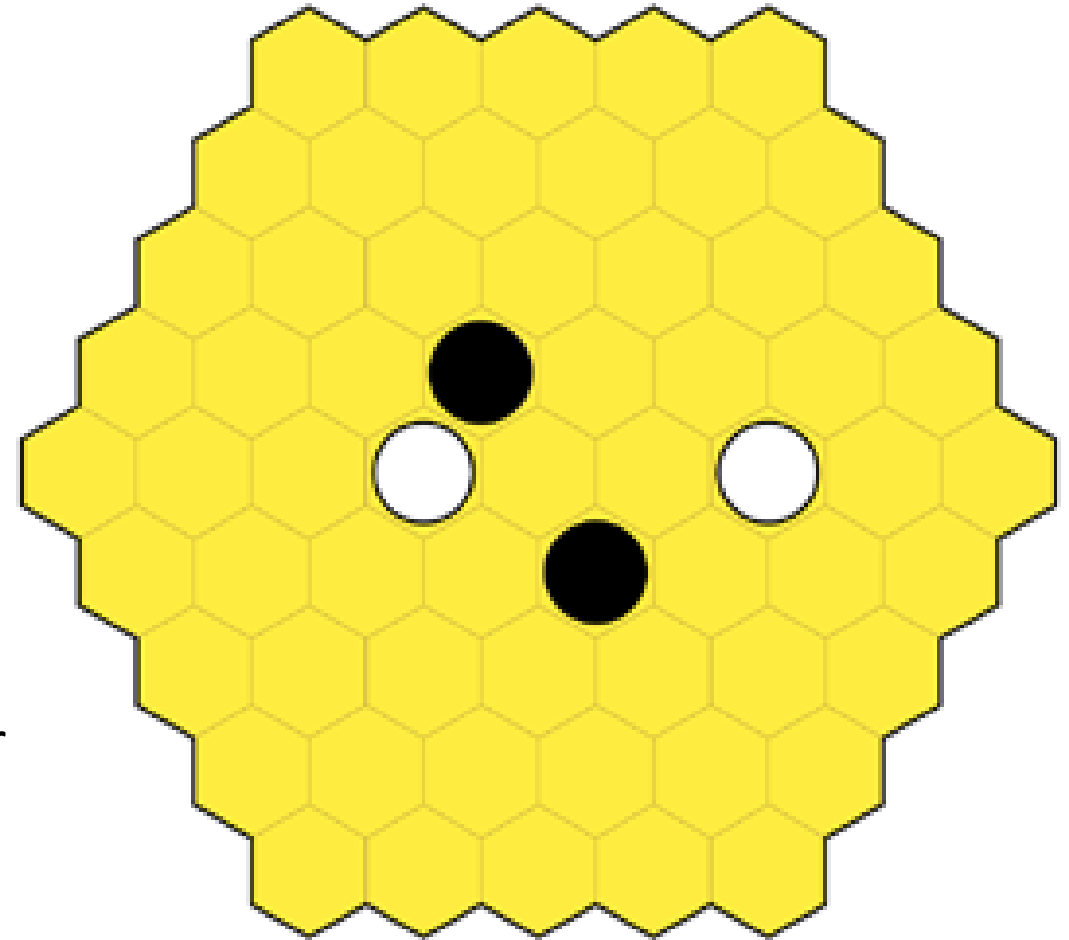
where M_S is the number of moves available in S and M_{S_w} is the number of winning moves available in S .

AI-Assisted Game Invention

- Cameron Browne's work
 - Digital Ludeme Project (<http://ludeme.eu/>) - seeks to express a wide variety of historical games according to game units called "ludemes".
 - The project seeks to "represent games as structured sets of *ludemes* (units of game-related information)", e.g. objectives (capture all pieces, reach position, form a line of length n), actions (jump capture, place piece), etc.
 - Evolutionary algorithmic approach to computationally design the game Yavalath (Browne, 2011, pp. 75-85; <http://cambolbro.com/games/yavalath/>).
 - See also Browne's *Game and Puzzle Design Journal* (<http://gapdjournal.com/>)

Yavalath

- Rules and examples from:
<http://cambolbro.com/games/yavalath/>
- Rules:
 - The board starts empty.
 - Two players take turns adding a piece of their color to an empty cell.
 - **Win** by making a line-of-4 (or more) pieces of your color.
 - **Lose** by making a line-of-3 pieces of your color beforehand.
 - **Draw** if the board otherwise fills up.
- White to win with which move?
- Annotated sample game at link above.



Mentored Undergraduate Research Challenge

- A limited number of papers will be accepted for publication and presentation at EAAI-22 that
 - exemplify high-quality scholarly writing, and
 - demonstrate creative application of AI to AI-Assisted Game Design.
- As mentioned, this can take many forms, including but not limited to:
 - Existing game improvement through AI game analysis,
 - New game design through AI search in a design space,
 - Adaptive technologies shown empirically to improve player experience, and
 - AI procedural generation of game play elements.
- Teams must include at least one faculty mentor and at least one undergraduate. Faculty mentors will be expected to review. (Contact Todd Neller <tneller@gettysburg.edu> to register teams.)

The AI-Assisted Game Design Challenge

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Or search “AI Matters”

Questions?

