

# Playing Hanabi Near-Optimally

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# Outline

- The game of Hanabi
- Previous work
- The hat principle
- Artificial players
- Experiments
- Results
- Conclusions and future work



# Hanabi Game Set



# Example

NP=3 players, NCPP=4 cards per player

Fireworks	0				1				3				1	2			
Deck	22				Blue Tok. 4				Red Tok. 3				score 7				
Trash	1	1	2	3	1	4	2	1	4								
Player 1	1		2		3		1										
Information	Not red		red		Not red		red										
Player 2	4		5		2		1										
Information	Not w.		Not w.		white		?										
Player 3	2		3		2		5										
Information	2		?		2		Not 2										

# My own cards are hidden

NP=3 players, NCPP=4 cards per player

Fireworks	0				1				3				1				2	
Deck	22				Blue Tok.		4				Red Tok.		3				score	7
Trash	1	1	2	3	1	4	2	1	4									
Player 1	X	X	X	X														
Information	Not red	red	Not red	red														
Player 2	4	5	2	1														
Information	Not w.	Not w.	white	?														
Player 3	2	3	2	5														
Information	2	?	2	Not 2														

# 3 kinds of move

- **Play** a card
- **Discard** a card
- **Inform** a player with either a color or a height

# I choose to play card number 2

NP=3 players, NCPP=4 cards per player

Fireworks	0			1				3			1		2		
Deck	22		Blue Tok. 4				Red Tok. 3			score 7					
Trash	1	1	2	3	1	4	2	1	4						
Player 1	X	X	X	X											
Information	Not red	red	Not red	red											
Player 2	4	5	2	1											
Information	Not w.	Not w.	white	?											
Player 3	2	3	2	5											
Information	2	?	2	Not 2											

# Oops, it was red 2 ==> penalty

NP=3 players, NCPP=4 cards per player

Fireworks	0				1				3			1				2
Deck	21				Blue Tok.	4						Red Tok.	2			score 7
Trash	1	1	2	3	1	4	2	1	4	2						
Player 1	X	X	X	X												
Information	Not red	?	Not red	red												
Player 2	4	5	2	1												
Information	Not w.	Not w.	white	?												
Player 3	2	3	2	5												
Information	2	?	2	Not 2												



# Player 2 to move

NP=3 players, NCPP=4 cards per player

Fireworks	0				1				3				1				2			
Deck	21				Blue Tok. 4				Red Tok. 2				score 7							
Trash	1	1	2	3	1	4	2	1	4	2										
Player 1	1		4		3		1													
Information	Not red		?		Not red		red													
Player 2	X		X		X		X													
Information	Not w.		Not w.		white		?													
Player 3	2		3		2		5													
Information	2		?		2		Not 2													

# P2 informs p3 with color = red

NP=3 players, NCPP=4 cards per player

Fireworks	0			1			3			1	2		
Deck	21			Blue Tok. 3			Red Tok. 2			score 7			
Trash	1	1	2	3	1	4	2	1	4	2			
Player 1	1		4		3		1						
Information	Not red		?		Not red		red						
Player 2	X		X		X		X						
Information	Not w.		Not w.		white		?						
Player 3	2		3		2		5						
Information	2 Red		Not red		2 not red		Not 2 Red						

# P3 informs p1 with height = 1

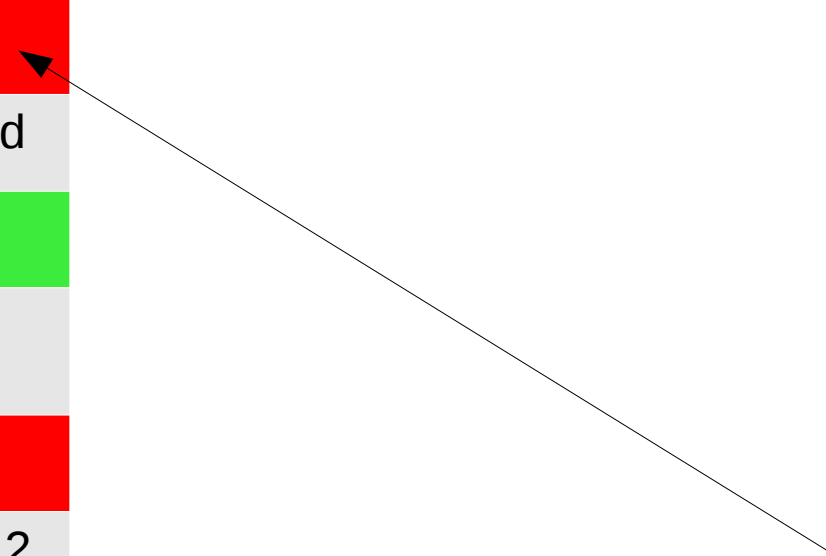
NP=3 players, NCPP=4 cards per player

Fireworks	0		1		3		1		2	
Deck	21		Blue Tok. 2		Red Tok. 2		score 7			
Trash	1	1	2	3	1	4	2	1	4	2
Player 1	1		4		3		1			
Information	1 Not red		Not 1		Not red not 1		1 red			
Player 2	4		5		2		1			
Information	Not w.		Not w.		white		?			
Player 3	2		X		2		X			
Information	2 Red		Not red		2 not red		Not 2 Red			

# P1 chooses to play card 4

NP=3 players, NCPP=4 cards per player

Fireworks	0			1			3			1	2		
Deck	21			Blue Tok. 2			Red Tok. 2			score 7			
Trash	1	1	2	3	1	4	2	1	4	2			
Player 1	X	X	X	1									
Information	1 Not red	Not 1	Not red not 1	1 red									
Player 2	4	5	2	1									
Information	Not w.	Not w.	white	?									
Player 3	2	3	2	5									
Information	2 Red	Not red	2 not red	Not 2 Red									



# Success !

NP=3 players, NCPP=4 cards per player

Fireworks	1			1		3		1	2				
Deck	20			Blue Tok.	2			Red Tok.	2			score	8
Trash	1	1	2	3	1	4	2	1	4	2			
Player 1	X	X	X	X									
Information	1 Not red	Not 1	Not red not 1	?									
Player 2	4	5	2	1									
Information	Not w.	Not w.	white	?									
Player 3	2	3	2	5									
Information	2 Red	Not red	2 not red	Not 2 Red									

The diagram shows arrows indicating the flow of information and card placement. One arrow points from the '1' in the first red firework to the '1' in the first blue trash card. Another arrow points from the '1' in the first blue trash card to the '?' in the information for Player 1. A third arrow points from the '1' in the first green trash card to the '?' in the information for Player 2. A fourth arrow points from the '1' in the first yellow trash card to the '?' in the information for Player 3. A fifth arrow points from the '2' in the first yellow firework to the '8' in the score.

# Player 2 chooses to discard card 2

NP=3 players, NCPP=4 cards per player

Fireworks	1				1				3				1				2			
Deck	20				Blue Tok. 2				Red Tok. 2				score 8							
Trash	1	1	2	3	1	4	2	1	4	2										
Player 1	1		4		3		3													
Information	1 Not red		Not 1		Not red not 1		?													
Player 2	X		X		X		X													
Information	Not w.		Not w.		white		?													
Player 3	2		3		2		5													
Information	2 Red		Not red		2 not red		Not 2 Red													

# One blue token is added

NP=3 players, NCPP=4 cards per player

Fireworks	1			1			3			1	2		
Deck	19				Blue Tok. 3			Red Tok. 2			score 8		
Trash	1	1	2	3	1	4	2	1	4	2	5		
Player 1	1	4	3	3									
Information	1 Not red	Not 1	Not red not 1	?									
Player 2	X	X	X	X									
Information	Not w.	?	white	?									
Player 3	2	3	2	5									
Information	2 Red	Not red	2 not red	Not 2 Red									

# Hanabi

- The game ends when
  - The number of red tokens is zero
  - The score is 25
  - Each player has played once since the deck is empty
- Main features
  - Cooperative, N players
  - Hidden information
  - Finite episode
  - Explicit information moves





# Previous work

- Osawa 2015
  - Partner models, NP=2, NCPP=5, <score> = 15.9
- Kusters, Van den berghe 2016
  - Miscellaneous, NP=3, NCPP=5, <score> = 15.4
- Franz 2016
  - MCTS, NP=4, NCPP=5, <score> = 17
- Cox 2015
  - Hat principle, NP=5, NCPP=4, <score> = 24.5

# Goal

- Implement the hat principle
- Improve Cox's result
- Generalize to other NP and NCPP values



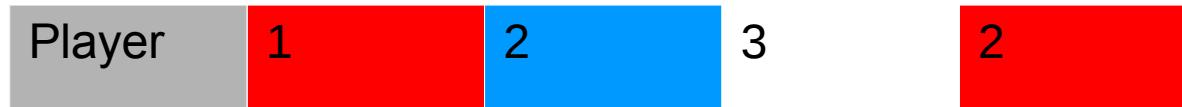
# The hat principle

- Recreational mathematics
- Each hand is represented with a number – a hat – computed with a public program P1
- Each information move emitted by player P corresponds to the sum of hats player P sees. (public program P2)
- « Recommendation » version with NP=4
  - Each value of h corresponds to a « recommendation » (play card 1, play card 2, play card 3, play card 4, discard card 1, discard card 2, discard card 3, discard card 4)
  - 4 playing moves + 4 discarding moves = 8 moves
  - Hat h, such that  $0 \leq h < 8$
- Information move by player P :
  - Compute S(P) the sum of hats that P sees.
  - Perform the corresponding information move.
  - With a subtraction, players Q different from P can deduce their own hat.

# The hat principle

- Number of information moves (NIM)
  - $NIM = (NP-1) NIMP$
  - $NIMP = 10$ 
    - 5 colors + 5 heights (many work)
  - $NIMP = 2$ 
    - Color or height (Cox's work)
- Importance of the rule set
  - Informing a player with an empty set : allowed or not
  - $NIM \geq H$

# Allowing all information moves or not ?



- Wikipedia and many sources including our work
  - No forbidden information moves
  - NIMP = 10
- Cox 2015
  - No corresponding card in the player's hand ==> forbidden information moves
    - Color = Green
    - Color = Yellow
    - Height = 4
    - Height = 5
  - NIMP = 2
- Commercial ruleset
  - Not mentioned (!)

# The hat principle

- « Information » version
  - A hat -> a (subset of) value(s) of a card
  - Each hand has a specific card to be informed
    - Given by a public program P3
      - (Highest playing probability,
      - Left most non informed card)
- Ruleset
  - If NIM  $\geq 25$  : Each hat -> unique card value
  - Otherwise a hat may correspond to a subset of card values
- Effect
  - A player is quickly informed with its cards' values.
  - As if the players could see their own cards





# Artificial players

- **Certainty** player
  - Play or discard totally informed cards only
- **Confidence** player
  - Without proof of the contrary, assumes an informed card is playable
- **Seer** player
  - Sees its own card but not the deck
- **Hat** players
  - Recommendation player
  - Information player
- **Depth-one tree search** player
  - Use an above player as a policy in a depth-one Monte-Carlo search
  - Uses NCD plausible card distributions
  - (Kuhn 1955) polynomial time assignment problem algorithm

# Experiments

- Team made up with NP copies of the same player
- Test set
  - NG games (each with one card distribution)
  - NG = 100 for tree search players
  - NG = 10,000 for knowledge-based players
- « near-optimality » :
  - approaching the seer empirical score on a given test set.
- Settings
  - 3 Ghz , 10 minutes / game at most
  - No memory issue
  - NCD = 1, 10, 100, 1k, 10k.

# Results (knowledge based players)

Certainty (Cert), Confidence (Conf), Hat recommendation (Hrec) and Hat information (Hinf)  
 For NP = 2, 3, 4, 5 ; NCPP = 3, 4, 5 ; NG = 10,000

NP	Cert			Conf			Hrec			Hinf		
	3	4	5	3	4	5	3	4	5	3	4	5
2	10.3	10.7	11.1	16.9	16.7	15.8	15.8	16.9	17.8	5.9	6.4	6.7
3	12.9	13.0	13.5	19.4	19.2	17.9	22.8	23.8	23.8	18.7	19.4	18.9
4	14.4	14.7	14.1	20.3	19.7	17.9	23.2	23.5	22.8	24.3	24.6	24.4
5	15.2	14.4	12.8	20.6	19.2	16.8	23.2	22.6	21.0	24.6	24.7	24.3

Hat information, NP=5 NCPP=4, histogram of scores, NG = 10,000

Score	19	20	21	22	23	24	25
%	0.01	0.05	0.17	1.19	3.62	13.66	81.30

# Results (depth-one tree search players)

Tree search players using :

Confidence (Conf), Hat recommendation (Hrec), Hat information (Hinf), Seer

For NP = 2, 3, 4, 5 ; NCPP = 3, 4, 5 ; NG = 100 ; NCD = 100, 1k, 10k

NP	Conf			Hrec			Hinf			Seer		
	3	4	5	3	4	5	3	4	5	3	4	5
2	19.2	19.4	19.0	16.40	17.38	18.53				23.10	24.46	24.91
3	20.7	21.1	20.4	23.96	24.56	24.70				24.62	24.97	25.00
4	21.5	21.0	19.7	24.34	24.60	24.45	24.72	24.96	24.91	24.91	25.00	24.99
5	22.0	20.4	18.0	24.26	24.30	22.68	24.85	24.92	24.76	24.96	24.98	24.96

Tree search + Hat information, NP=5 NCPP=4, Histogram of scores , NG = 100

Score	19	20	21	22	23	24	25
%	0	0	0	0	0	8	92



# Conclusions and future work

- Summary
  - The hat convention « kills » the game for the computer
  - Current work :
    - Small step upward with depth-one search
      - 75 % → 90 % for NP=5 and NCPP=4
    - Generalization to NP = 2, 3, 4, 5 and NCPP = 3, 4, 5
    - Use of the hungarian method of the card assignment problem
- Future work :
  - Neural Network approach :
  - Is it possible to learn the hat convention ?
    - In self play ? What is the architecture ?
    - With a teacher ?

# Thank you for your attention!

Questions ?

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