A Memory-Based Approach to Two-Player Texas Hold'em

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http://www.cs.auckland.ac.nz/research/gameai







Introduction

- Memory-Based Approach
 - Simple approach
- Produce poker strategy
- Agent
 - Sartre

Overview

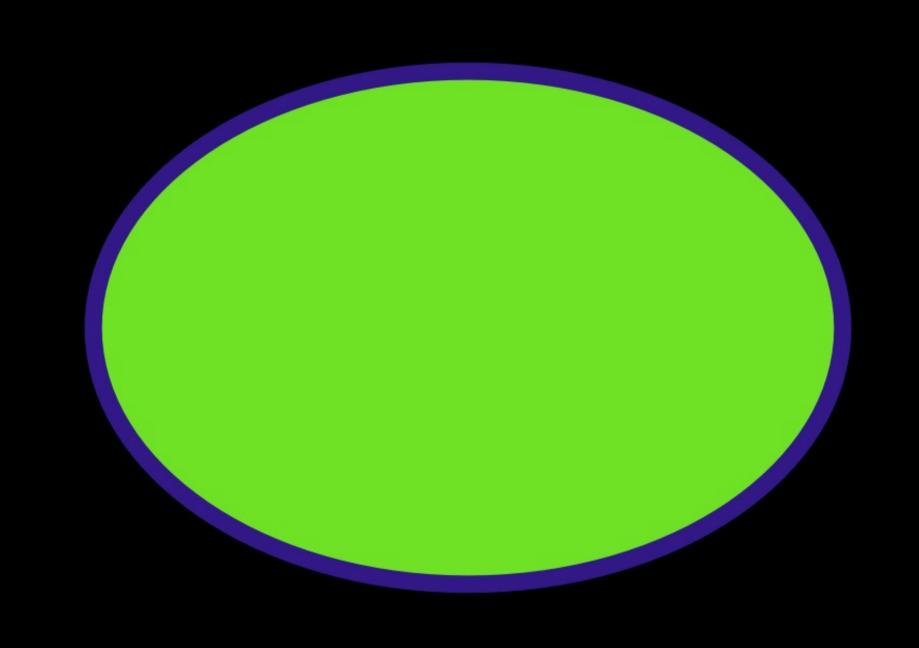
- Texas Hold'em
- Types of Strategies
- Related Approaches
- Memory-Based Approach
- Experimental Results
- Conclusions
- Future Work

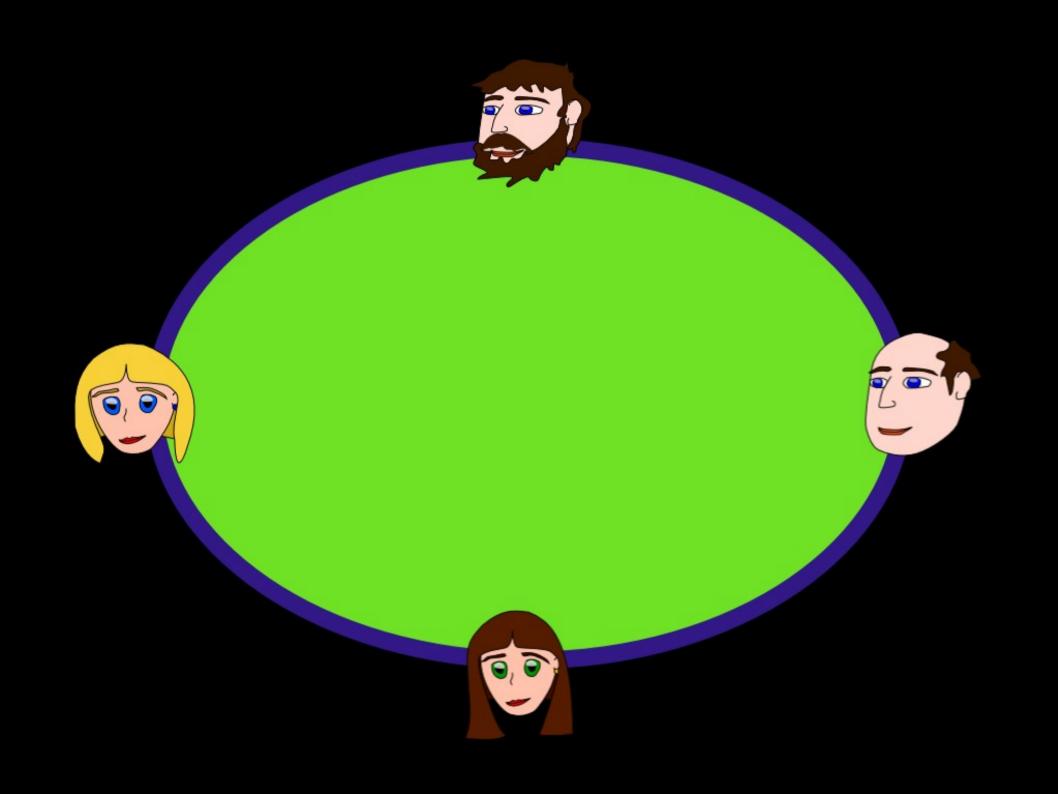
Introduction

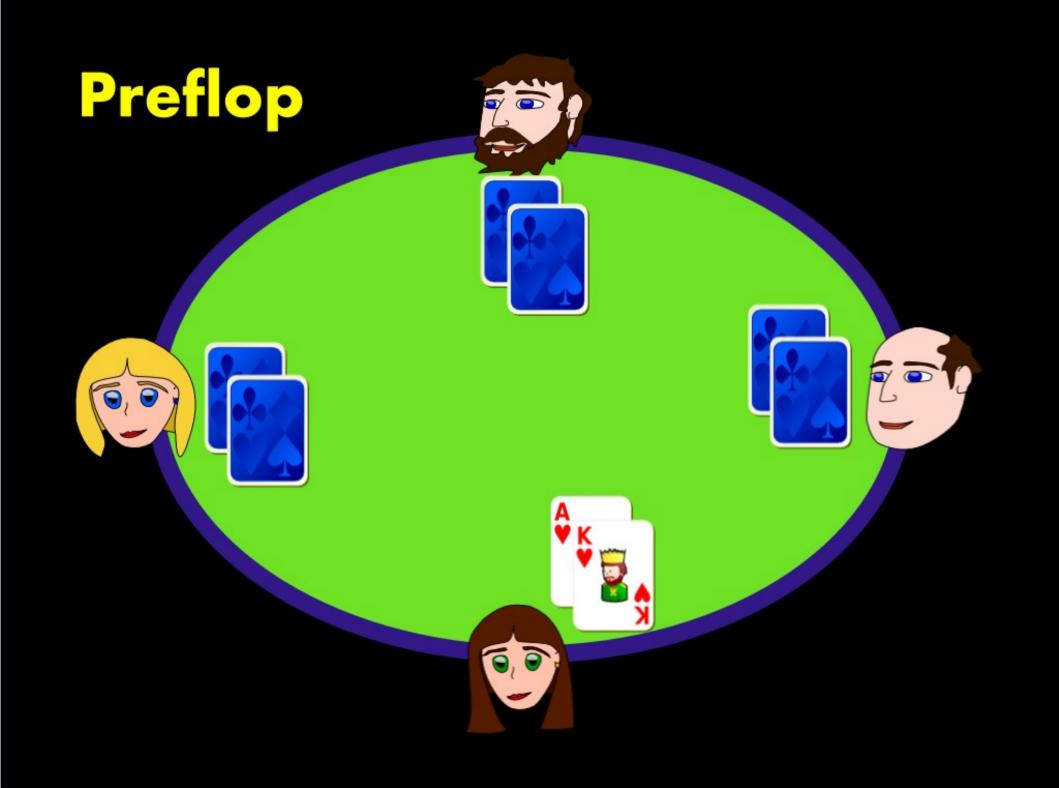
The Poker Domain

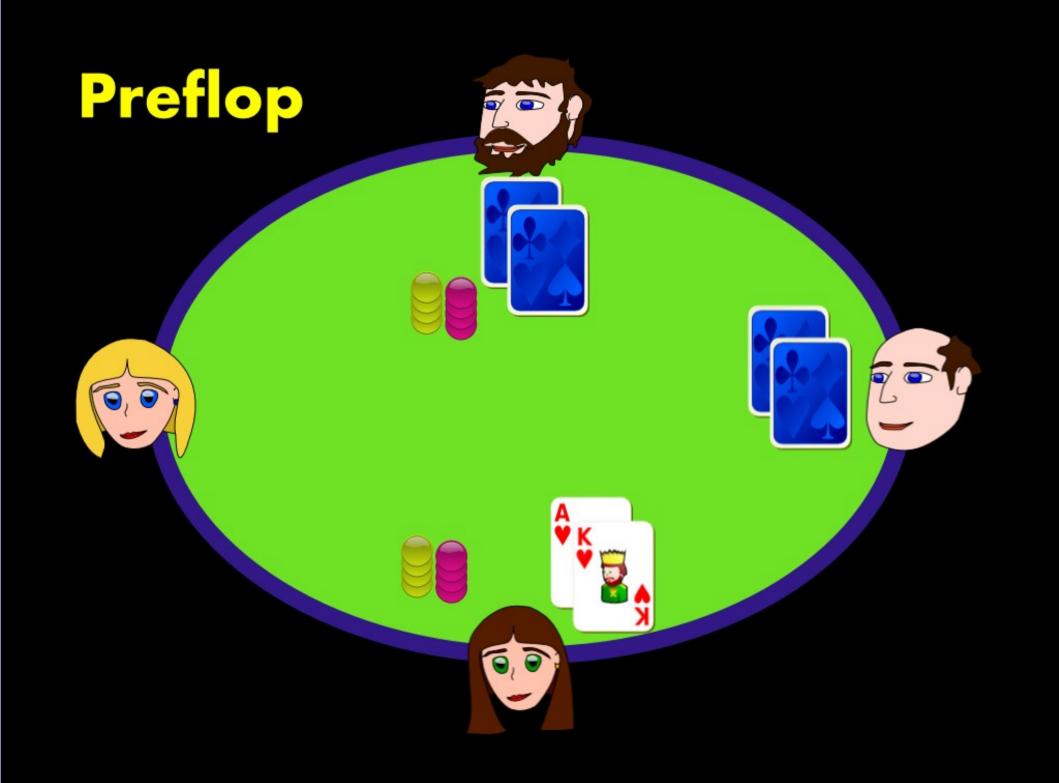
- Imperfect Information
- Chance events
- Rules and boundaries
- Performance evaluation
- Increasingly popular
 - AAAI Computer Poker Competition

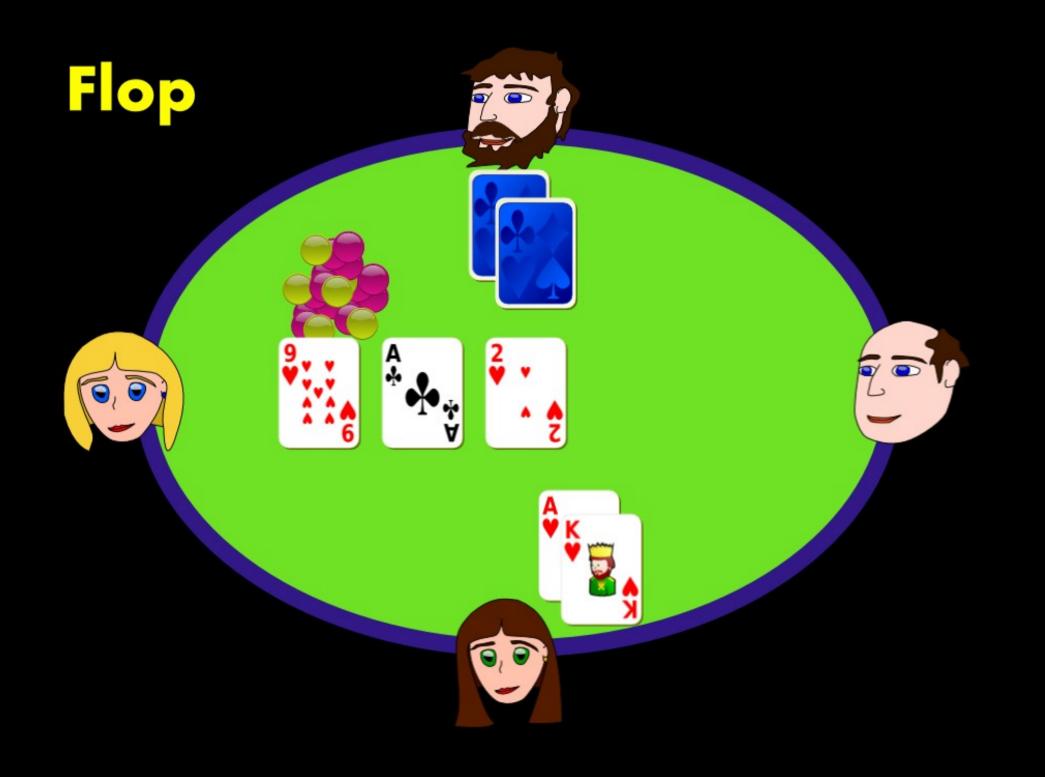
The Rules of Texas Hold'em

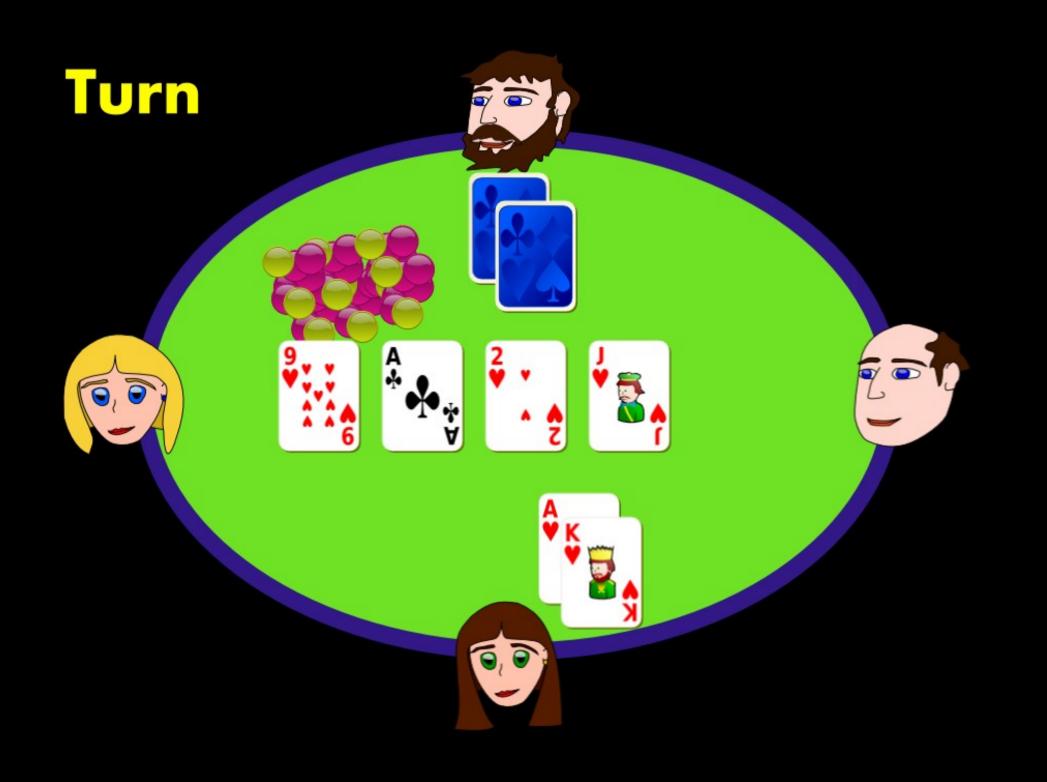


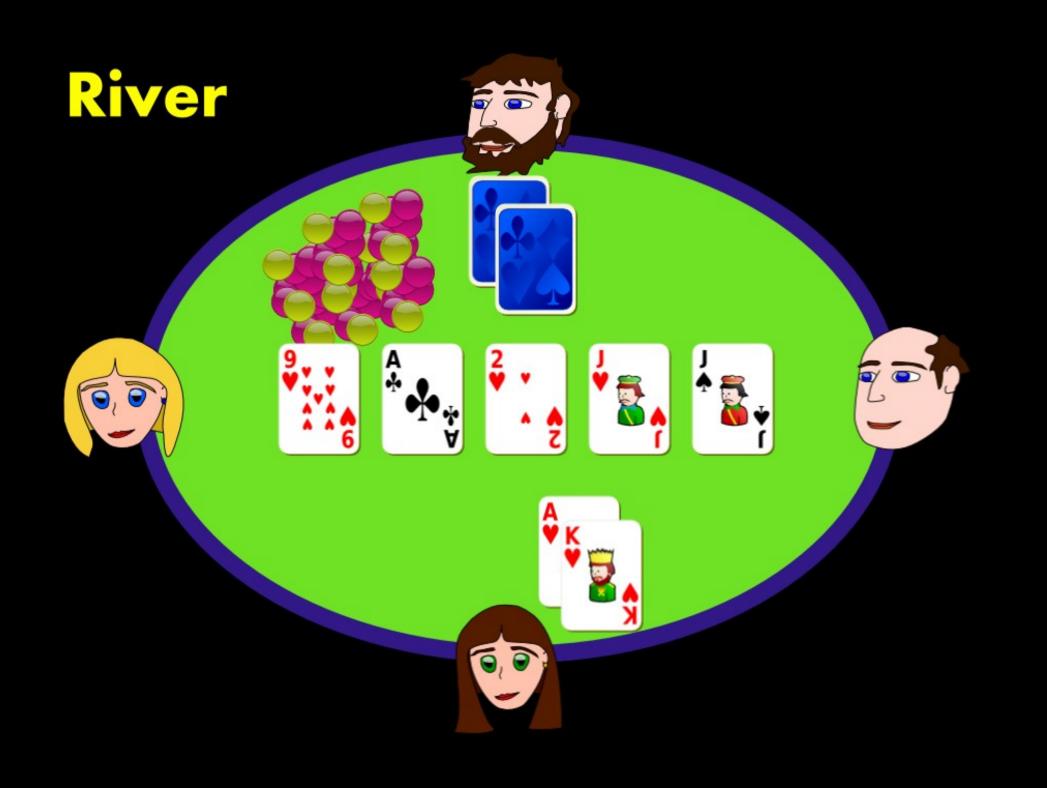


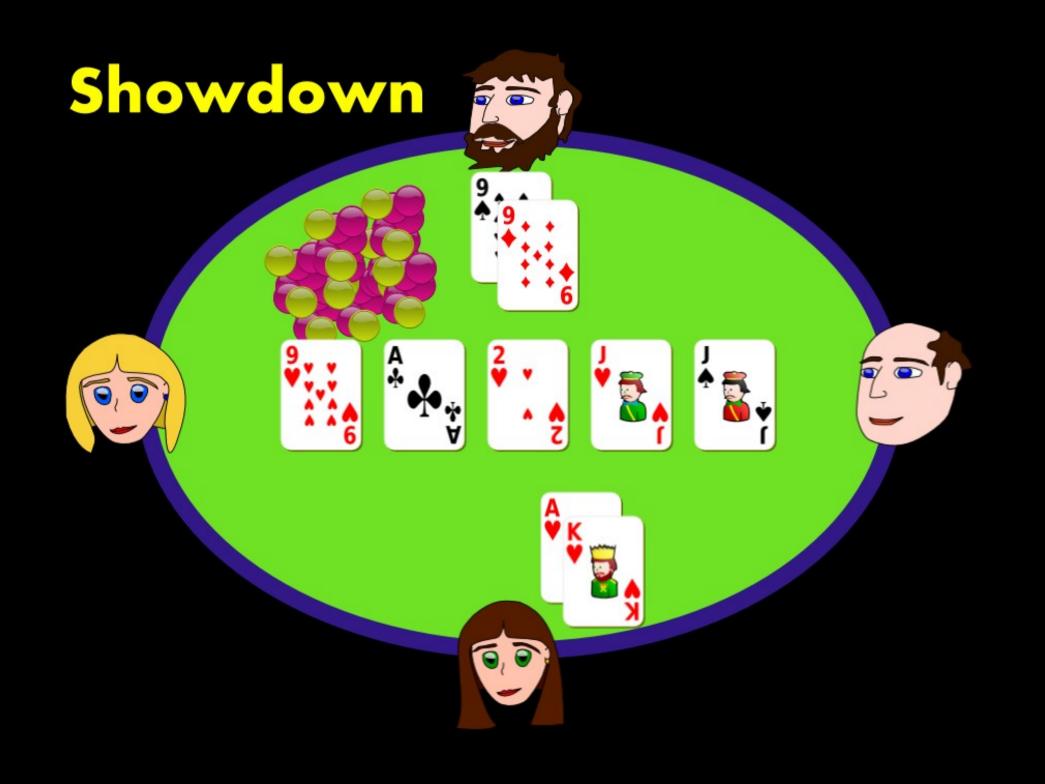












Texas Hold'em

Current Focus

- Heads up (2 players)
- Limit betting
 - \$2/\$4 Hold'em

Poker Strategies

A Poker Strategy

 At every decision point a probability triple is required that indicates the proportion of the time a player should either fold, call or raise

 $(f,c,r) \to (0, 0.5, 0.5)$

Types of Strategies

Nash Equilibria

 Robust strategies that attempt not to lose to any type of opponent

Exploitive Strategies

- Attempts to react to an opponent's play in a way that allows maximum exploitability of that opponent
- Requires opponent modeling

Rock-Paper-Scissors Example

Nash equilibrium

- $(R,P,S) \rightarrow (1/3, 1/3, 1/3)$

 The Nash player will never lose against any player in the long run

Along comes Jimmy who only ever plays Paper



Rock-Paper-Scissors Example

- The Nash player will continue to play
 - (1/3, 1/3, 1/3)
 - Lose 33%, Win 33%, Draw 33%
 - The Nash player will still only draw against Jimmy

Rock-Paper-Scissors Example

- However, because we know Jimmy's strategy an exploitive player would be better off using the strategy
 - (0, 0, 1.0)
 - i.e. a best response that maximally exploits Jimmy at every decision point
- Now, against Jimmy the exploitive player will win
 - Consequence is that the exploitive player plays off the equilibrium, and is hence subject to potential exploitation itself

Approaches to creating poker agents

e-Nash Equilibrium

- Linear Programming
 - Constructs matrices that act as constraints within an optimization problem
- Iterative approaches
 - Basic idea: Two players begin with arbitrary strategies, play many repetitions of a game and modify their strategies in a way that improves their strategy against their opponent.
 - As the number of iterations increases the strategies approach a Nash equilibrium
 - e.g. Fictitious Play, Counterfactual Regret Minimization

e-Nash Equilibrium

- A Nash equilibrium can easily be computed for Rock-Paper-Scissors
- However, the poker game tree is much to large to find exact Nash equilibria
 - Abstractions required
- Can only approximate Nash-equilibria
 - e-Nash Equilibria
 - e specifies a lower bound on how exploitable the equilibrium strategy is

Exploitive Strategies

Miximax search

- Similar to minimax in perfect information games
- Maintains an opponent model used during game tree search to inform expected value calculations of taking certain betting actions
- Restricted Nash Response (RNR) & Data Biased Response (DBR)
 - Somewhere between an *e*-Nash equilibrium and a best response to an opponent's static strategy

Our Approach



 Investigate whether hand histories from strong poker players can be reused within a Case-Based Reasoning framework to achieve a similar performance?

Case-Based Reasoning (CBR)

- Solutions of past problems are reused or adapted to handle solutions for novel problems
- Lazy Learning approach
- Stores a memory of cases along with their solutions and outcomes
- When a new problem is encountered similar cases are retrieved from the case-base and their solutions are reused to solve the problem

A Memory-Based Approach

- Casper (CASe-based Poker playER)
 - Past poker agent for 10-player Texas Hold'em
- Sartre (Similarity Assessment Reasoning for Texas hold'em via Recall of Experience)
 - Our latest agent
 - Specialised for heads-up limit hold'em

Overview

- Cases are attribute-value pairs
- Separate case-bases are used for each different round (preflop, flop, turn, river)
- When a decision is required a case is created to describe the current state of the game and the appropriate case-base is searched to find similar cases
- The solution of the similar cases are reused for the current situation

Case Representation

Attribute	Туре	Example
Hand Type	Class	Missed, Pair, Two- Pair, Set, Flush,Flush-Draw, Straight-Draw,
Betting Sequence	String	<i>rc-c, crrc-crrc-cc-</i> <i>r,</i>
Board Texture	Class	No-Salient, Flush- Possible, Straight- Possible, Flush- Highly-Possible,
Solution	Char	<i>f, c, r</i>
Outcome	Numerical	+14, -1, -5, +20,

Case Retrieval

- Current version of Sartre uses All-or-Nothing local similarity, i.e. either attribute values are entirely similar or dissimilar
 - Baseline for future improvements
- Number of retrieved cases varies from 0 to 1000s
- If 0 cases retrieved Sartre adopts a default strategy
 - Allways-Call

Solution Reuse

- Many cases retrieved which betting action to make?
- 3 solution reuse policies
 - 1) Reuse the majority decision
 - 2) Probabilistically select actions*
 - 3) Reuse solution which achieved the greatest outcome

Training Data

- Trained on data from the best agent equilibrium agent from the 2008 Computer Poker Competition
 - Hyperborean-Eqm

Round	# of Cases
Preflop	201,335
Flop	300,577
Turn	281,529
River	216,597

Experimental Results

Experiments

Where possible used: Duplicate Matches

- N hands in forward + backwards direction
- Set of hands played
- Set of hands replayed, but agents receive the cards that their opponent previously received
- Reduces variance
- Small bets per hand (sb/h)

- Sartre Vs. FellOmen2
 - Sartre "expert" trained by Hyperborean-Eqm
 - Hyperborean-Eqm Vs. FellOmen2 results known
 - Compare Sartre Vs. FellOmen2 to Hyperborean Vs. FellOmen2
 - FellOmen2 2nd equal in 2008 equilibrium CPC
 - Publicly available
- 6 rounds of N = 3000 duplicate hands

Sartre Vs. FellOmen2

Round	Sartre (sb/h)	Hyperborea n (sb/h)*
1	-0.025	+0.014
2	-0.041	+0.023
3	-0.094	+0.029
4	-0.055	+0.030
5	-0.066	+0.033
6	-0.070	+0.016
Average	-0.0585 +/- 0.01 sb/h	+0.0241 +/- 0.003 sb/h

* Note: *N* = 5000

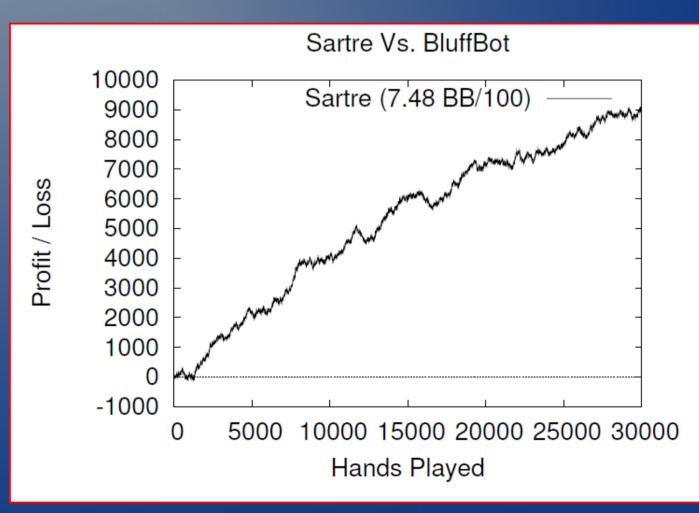
• Independent samples *t*-test gives *p* < 0.00001

Sartre Vs. BluffBot

- Further evaluation
- 2nd place in 2006 Computer Poker Competition
- Publicly available
- Duplicate match structure not available
- Straight 30,000 hands

Sartre Vs. BluffBot

- +0.150 sb/h



- 2009 IJCAI Computer Poker Competition
 - Participated in limit hold'em competition
 - Same system, but with majority-decision reuse policy
 - Chosen because of results of self-play experiments
 - 13 competitors
 - 2 divisions
 - Bankroll
 - Equilibrium

2009 IJCAI Computer Poker Competition

Limit bankroll division

Place	Agent	sb/h
1	MANZANA	0.186
2	Hyperborean-BR	0.116
3	GGValuta	0.110
4	Hyperborean-Eqm	0.116
5	Rockhopper	0.103
6	Sartre	0.097
7	Slumbot	0.096
8	GS5	0.082
9	AoBot	-0.002
10	dcurbHU	-0.07
11	LIDIA	-0.094
12	GS5Dynamic	-0.201

- 2009 IJCAI Computer Poker Competition
 - Limit equilibrium division

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12	dcurbHU
13	Tommybot

Conclusions

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- Presented a straight-forward, memory based approach for 2-player limit Texas Hold'em
- Initial results show a disparity between our memory-based system trained via "expert" and actual "expert" player results
 - With further improvements we believe we can limit this gap
- Memory-based approach still able to achieve strategies of reasonable quality
 - Consistent profit against BluffBot
 - 6th place finish in 2009 Computer Poker Competition

Future Work

Future Work

Improved similarity measures & generalization
Compare against Sartre-Baseline

Investigate Case Representation

• No limit betting

Thank you! To challenge Sartre go to: www.cs.auckland.ac.nz/poker