

Poker-CNN: Learning Pattern in Poker Games Using a Convolutional Network

Nikolai Yakovenko – Twitter Cortex

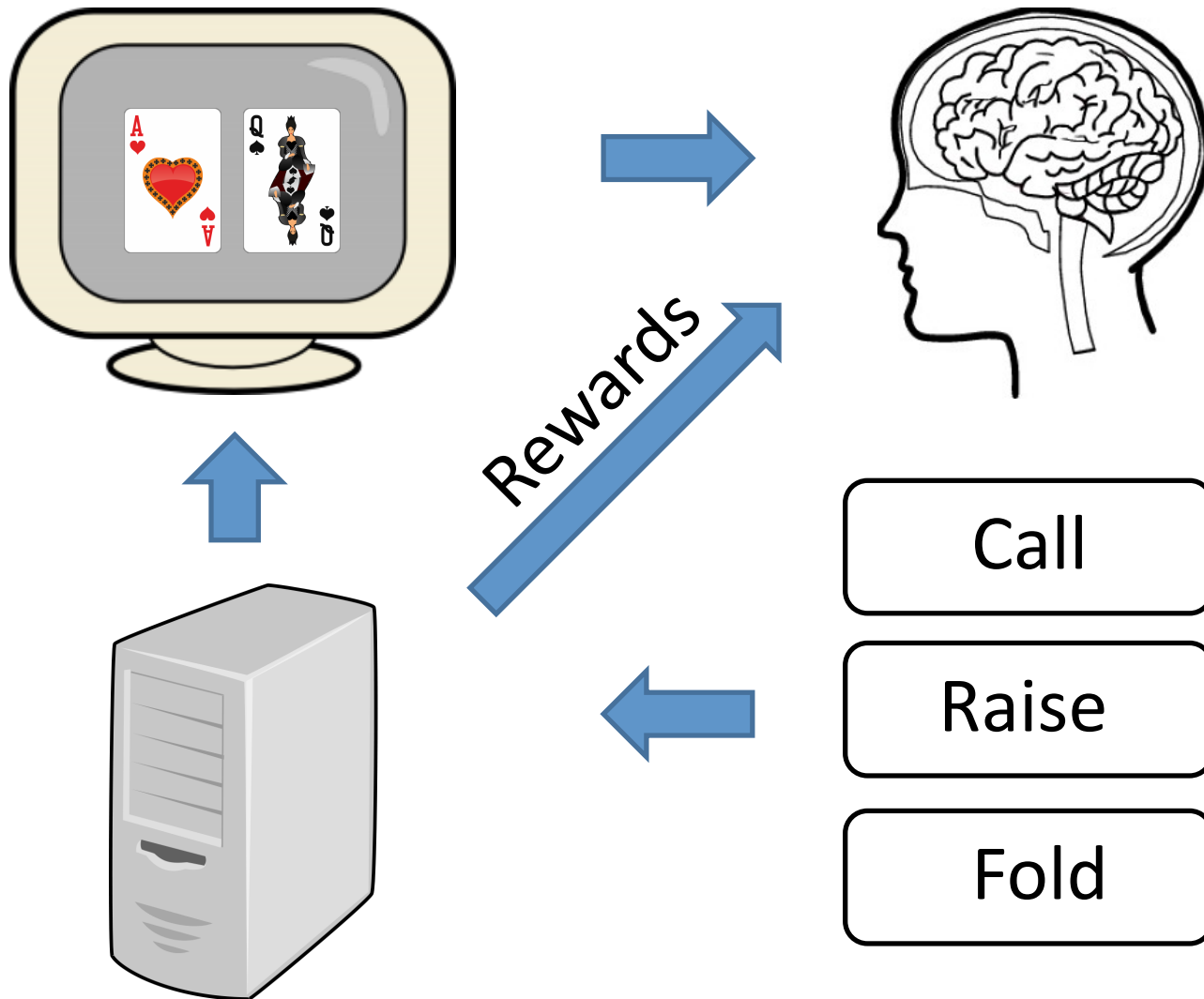
Liangliang Cao – Yahoo Labs & Columbia University

Colin Raffel – Columbia University

James Fan – Columbia University

AAAI 2016

Poker is a Turn-Based Video Game



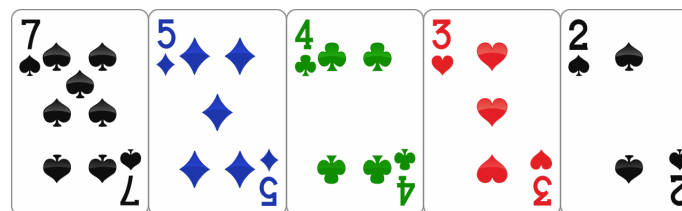
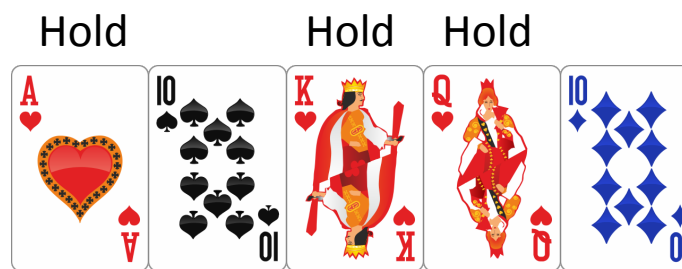
Unified Convnet Poker Framework

Single Draw Video Poker
98.5% payout

2-7 Lowball Triple Draw
Competitive against experts
(from self-play only)

Limit Hold'em
Competitive against experts
(from self-play only)

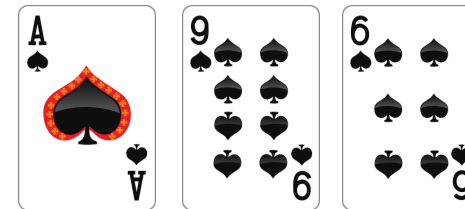
No Limit Hold'em
Tied-5th 2016 ACPC
(\$29/hand behind the winner)



Private cards



Public cards



Texas Hold'em

Private cards

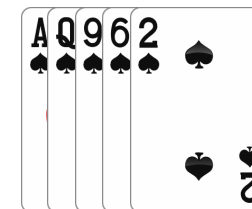
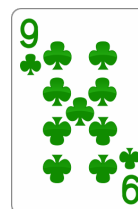
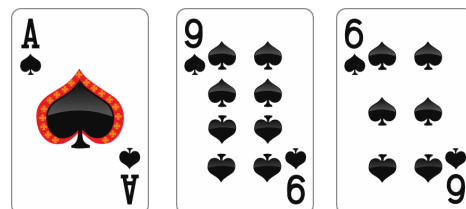
Flop (public)

Turn

River

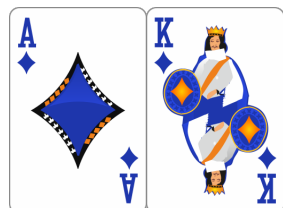
Showdown

Hero

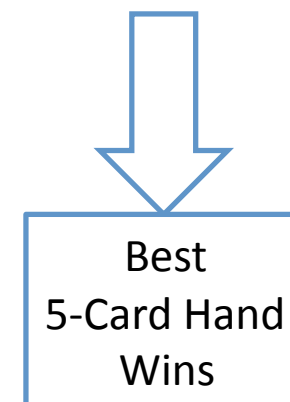
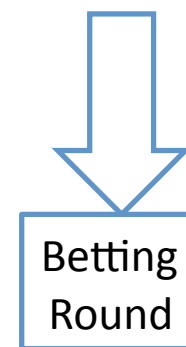
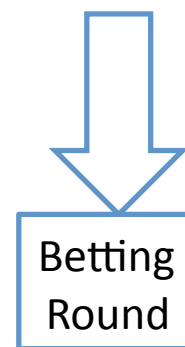
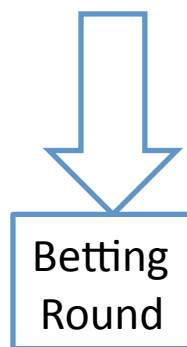


Flush

Oppn



Two Pairs



CFR: Equilibrium Balancing

- Abstract Hold'em game to smaller state-space
- Cycle over ever game state.
- Converges to Nash equilibrium in simplified game.
- Winners of every Annual Computer Poker Competition (ACPC) since 2007.
 - Limit Hold'em: 1% of unexploitable
 - No Limit Hold'em: statistical tie vs pro players

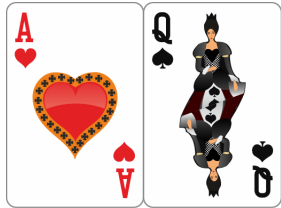
No Limit Hold'em Complexity

- 2016 ACPC 1st place winner Slumbot:
 - 9.7×10^{11} information sets
 - Each state encoded as single byte
 - 2 months to train
 - \$7,000 on AWS
- Tartanian (CMU)
 - Statistical tie vs best poker pros
 - 2 TB online lookup table

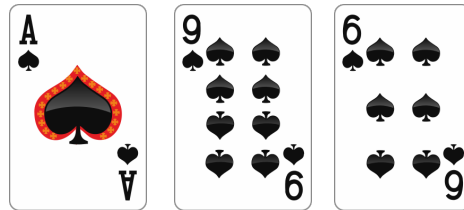
Can we train a strong poker player
with a much smaller strategy?

Poker-CNN: Cards as 2D Tensors

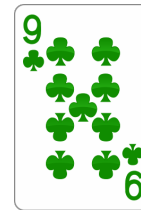
Private cards



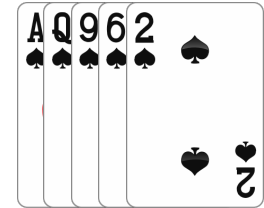
Flop (public)



Turn



River Showdown



Flush

[AhQs]

x	2	3	4	5	6	7	8	9	T	J	Q	K	A
c
d
h	1	.
s	1	.

[AhQs]+[As9s6s]

x	2	3	4	5	6	7	8	9	T	J	Q	K	A
c
d
h	1	.
s	.	.	1	.	1	.	1	.	1	.	1	.	.

Flush draw

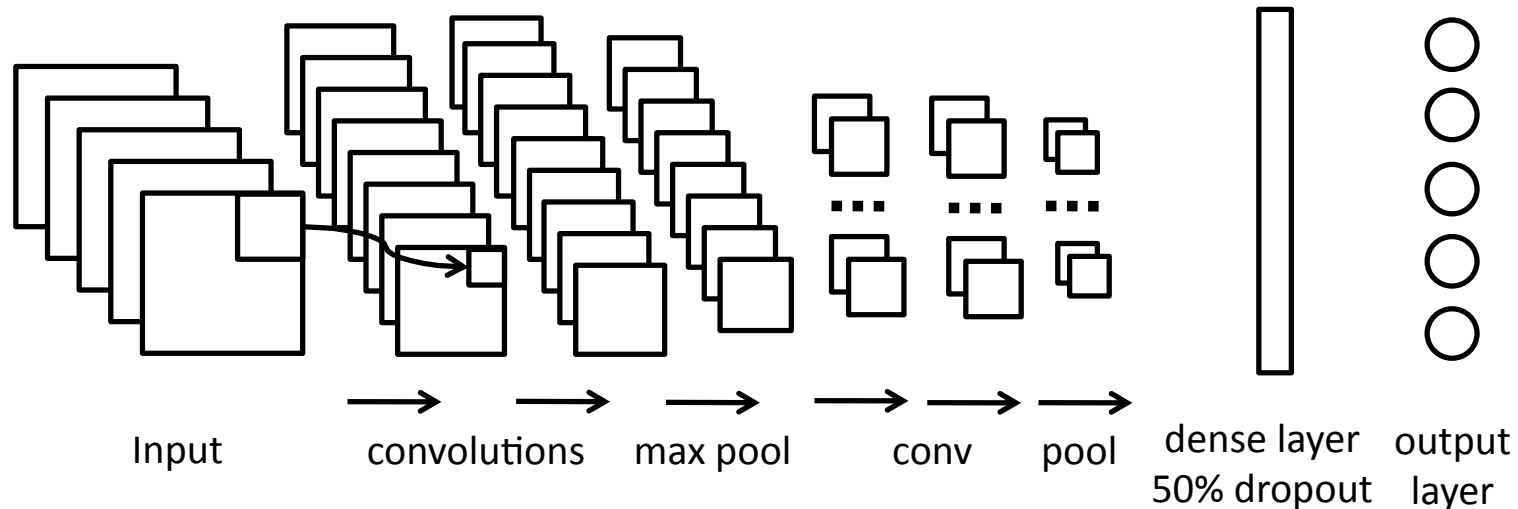
Pair (of Aces)

[AhQsAs9s6s9c2s]

x	2	3	4	5	6	7	8	9	T	J	Q	K	A
c	1
d
h	1	.
s	1	.	.	1	.	1	.	1	.	1	.	.	.

Flush!

Convnet: Predict Anything You Want



Inputs:

- Private cards
- Public cards
- Pot size
- Position
- Previous bets history

(31 x 17 x 17 3D tensor)

Predict action value:

- Bet, call, fold values
- Action probabilities
- Value by bet size

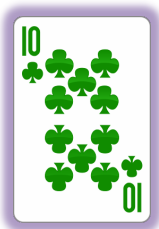
Surrogate tasks:

- All in odds
- Opponent hand distribution

- single-trial \$ win/loss

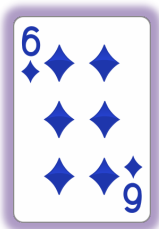
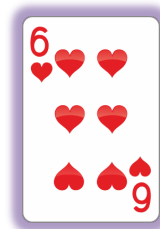
- no gradient for bets not made

- no Monte Carlo or tree search required



Big Blind
\$100

Small Blind
\$50



\$20,000



\$20,000

Raise	81.1%
Call	18.9%
Fold	0.0%

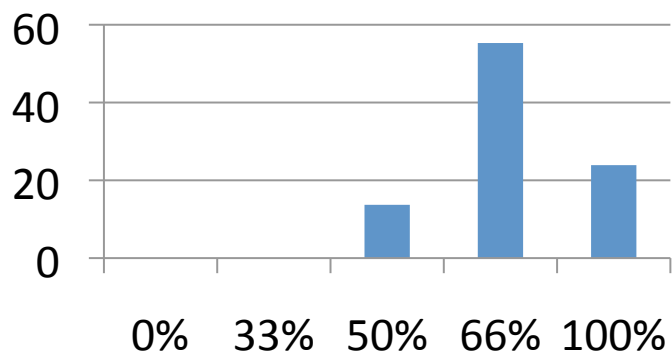
+\$2616

+\$265

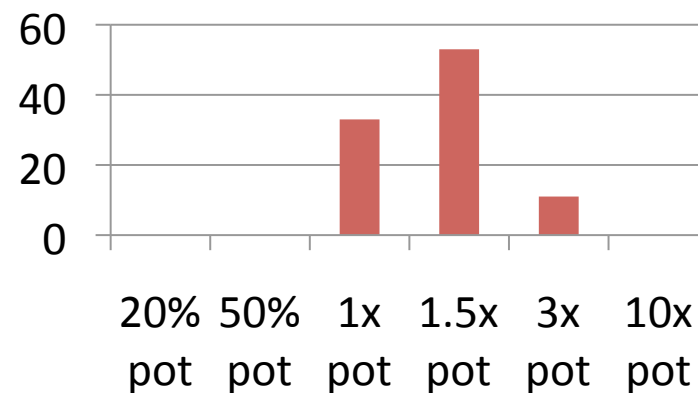
Raise	84.2%
Call	15.8%
Fold	0.0%

(Call)

Odds vs Opponent

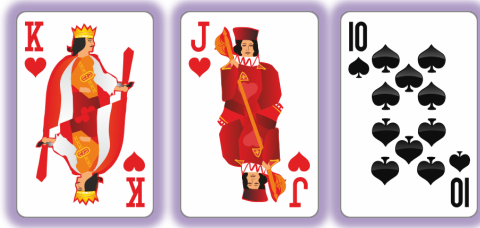


Bet Size





\$17,285



\$5,430



\$17,285

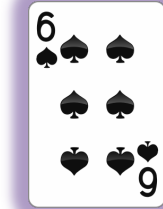
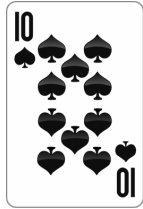
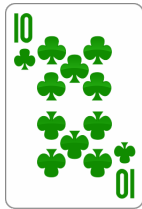
Bet 30.0%
Check 70.0%

(Check) (Check)

Bet 25.9%
Check 74.1%

Value vs random 91.3%
Value vs oppon 85.6%

Value vs random 52.9%
Value vs oppon 32.6%



\$5,430



\$17,285



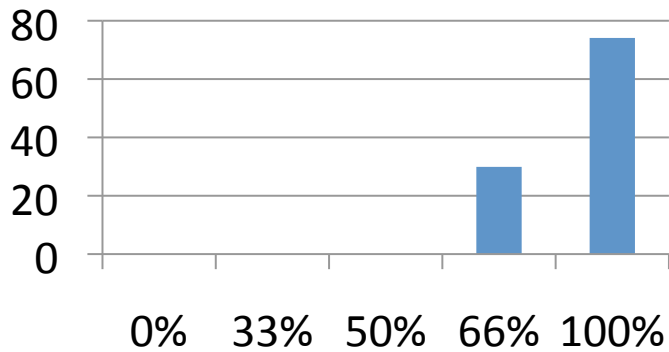
\$17,285

Bet 59.0%
Check 41.0% (Check)

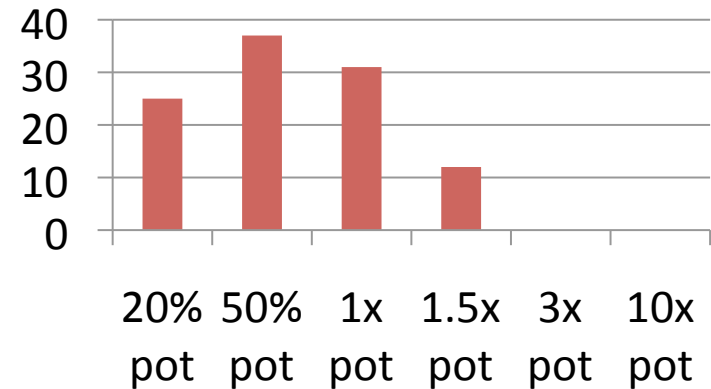
+\$3,967

Bet 86.6%
Check 13.4%

Odds vs Opponent



Bet Size





Raise 26.6%
 Call 73.4%
 Fold 0.0%

+\$17,285 (allin)

Call 32.4%
 Fold 67.6%

Value vs random 91.3%
 Value vs oppon 68.0%

Value vs random 84.7%
 Value vs oppon 30.6%

(\$13,000 allin call, to win \$26,000)

33.3% odds = break-even

Takeaways

- Pattern matching pretty good, with enough data
 - Naïve network design
 - Training \approx 4 million 2014 ACPC hands
 - No batch shuffle
 - No reinforcement learning
 - No LSTM memory units
- Struggles with rare cases
 - Under-weights outliers
 - Out of sample
- Struggles in big pots
 - Large importance for average results
 - Sparse data
- No attempt to avoid exploitability

Easy to Start, Easy to Modify

- Small 12-layer network in Theano
 - 2MB of model parameters
- Trains overnight on a single AWS GPU
- Easy to add surrogate tasks
 - For explanation
 - For better generalization
 - Small game-state abstractions for CFR?
- Relatively easy to train a new poker game

Future Work

- More games, more contexts
 - 3-player No Limit Hold'em
 - Pot-Limit Omaha
 - Tournament Hold'em
- Distill a CFR by learning internal parameters?
- Personal model against an opponent
- Hyper-parameter tuning...
 - 100,000 hands per experiment

References

- Paper on ArXiv <http://arxiv.org/abs/1509.06731>
- Code & models (admitted needs cleanup) at https://github.com/moscow25/deep_draw
- Annual Computer Poker Competition <http://www.computerpokercompetition.org/>
- Thanks to Eric Jackson of Slumbot for CFR code and advice.
- Thanks to poker pros Randy Ohel, Rep Porter & Neehar Banerji for help with human baselines.