# TTIC 31230, Fundamentals of Deep Learning

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What About alpha-beta?

#### **Grand Unification**

AlphaZero unifies chess and go algorithms.

This unification of intuition (go) and calculation (chess) is surprising.

This unification grew out of go algorithms.

But are the algorithmic insights of chess algorithms really irrelevant?

### Chess Background

The first min-max computer chess program was described by Claude Shannon in 1950.

Alpha-beta pruning was invented by various people independently, including John McCarthy in the late 1950s.

Alpha-beta has been the cornerstone of all chess algorithms until AlphaZero.

#### Alpha-Beta Pruning

```
def MaxValue(s,alpha,beta):
   value = alpha
   for s2 in s.children():
     value = max(value, MinValue(s2, value, beta))
     if value >= beta: break()
   return value
def MinValue(s,alpha,beta):
   value = beta
   for s2 in s.children():
     value = min(value, MaxValue(s2,alpha,value))
     if value <= alpha: break()</pre>
   return value
```

#### **Strategies**

An optimal alpha-beta tree is the union of a root-player strategy and an opponent strategy.

A strategy for the root player is a selection of a single action for each root-player move and a response for each possible action of the opponent.

A strategy for the opponent is a selection of a single action for each opponent move and a response for each possible action of the root player.

#### **Proposal**

Simulations should be divided into root-player strategy simulations and opponent strategy simulations.

A root-player strategy simulation is optimistic for the root player and pessimistic for the opponent.

An opponent strategy simulation is optimistic for the opponent player and pessimistic for the root-player.

#### Proposal

$$U(s,a) = \begin{cases} \lambda_u \, \pi_{\Phi}(s,a) & \text{if } N(s,a) = 0\\ \hat{\mu}(s,a) + \lambda_u \, \pi_{\Phi}(s,a)/N(s,a) & \text{otherwise} \end{cases}$$
(1)

 $\lambda_u$  should be divided into  $\lambda_u^+$  and  $\lambda_u^-$  with  $\lambda_u^+ > \lambda_u^-$ .

Simulations should be divided into two types — optimistic and pessimistic.

In optimistic simulations we use  $\lambda_u^+$  for root-player moves and  $\lambda_u^-$  for opponent moves.

In pessimistic simulations we use  $\lambda_u^-$  for root-player moves and  $\lambda_u^+$  for opponent moves.

### AlphaStar

Grandmaster level in StarCraft II using multi-agent reinforcement learning, Nature Oct. 2019, Vinyals et al.

#### StarCraft:

- Players control hundreds of units.
- Individual actions are selected from  $10^{26}$  possibilities (an action is a kind of procedure call with arguments).
- Cyclic non-transitive strategies (rock-paper-scisors).
- Imperfect information the state is not fully observable.

### The Paper is Vague

It basically says the following ideas are used:

A policy gradient algorithm, auto-regressive policies, self-attention over the observation history, LSTMs, pointer-networks, scatter connections, replay buffers, asynchronous advantage actor-critic algorithms,  $\mathrm{TD}(\lambda)$  (gradients on value function Bellman error), clipped importance sampling (V-trace), a new undefined method they call UPGO that "moves policies toward trajectories with better than average reward", a value function that can see the opponents observation (training only), a "z statistic" stating a high level strategy, supervised learning from human play, a "league" of players (next slide).

## The League

The league has three classes of agents: main (M), main exploiters (E), and league exploiters (L). M and L play against everybody. E plays only against M.

# A Rube Goldberg Contraption?

## Video

https://www.youtube.com/watch?v=UuhECwm31dM

# $\mathbf{END}$