

# Machine Learning in Games

## The Magic of Research in Microsoft Products

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# Overview

- Why Machine Learning and Games?
- Machine Learning in Video Games
  - Drivatars™
  - Reinforcement Learning
- Machine Learning in Online Games
  - TrueSkill™
  - Halo 3
- The Path of Go
- Conclusions

# Why

## Test

- Perfect measurement
- Perfect
- Reduced
- Reduced
- Great



AI  
hes

ent

# Games can be very hard!

- Partially observable stochastic games
  - States only partially observed
  - Multiple agents choose actions
  - Stochastic pay-offs and state transitions depend on state and all the other agents' actions
  - Goal: Optimise long term pay-off (reward)
- Just like life: complex, adversarial, uncertain, and we are in it for the long run!

# Approximations

From single player's perspective

- Partially Observable Markov Decision Process (POMDP)

Approximate Solutions

- Reinforcement Learning
- Unsupervised Learning
- Supervised Learning

What is the best AI?

- Always takes optimal actions
- Delivers best entertainment value

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# Drivatar™



Adaptive avatar for driving



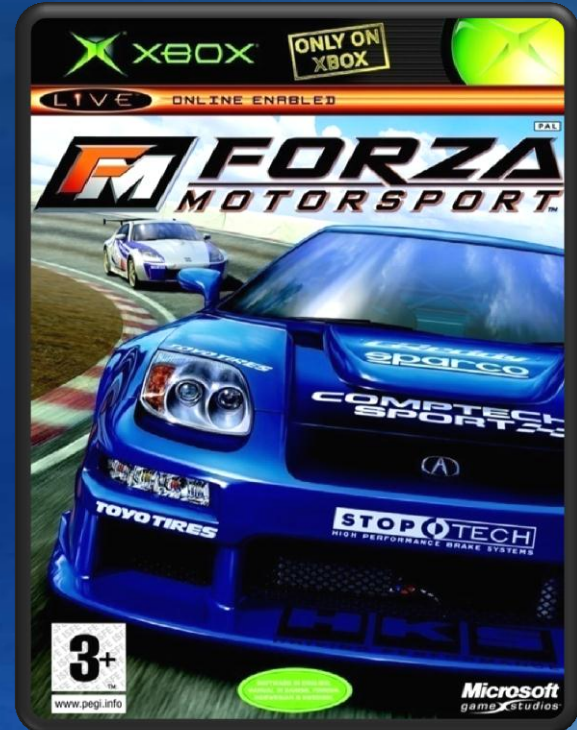
Separate game mode



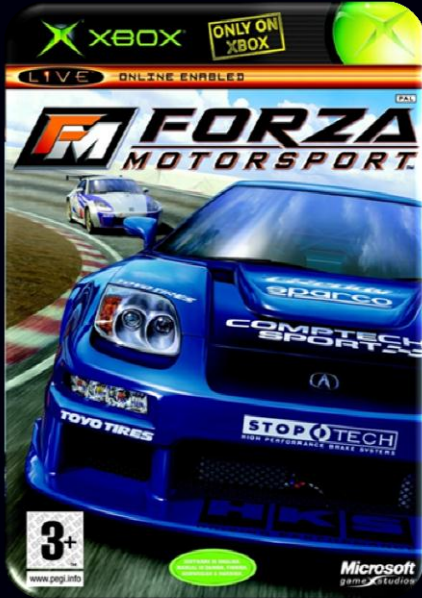
Basis of all in-game AI



Basis of “dynamic” racing line



# Demo: Forza Motorsport



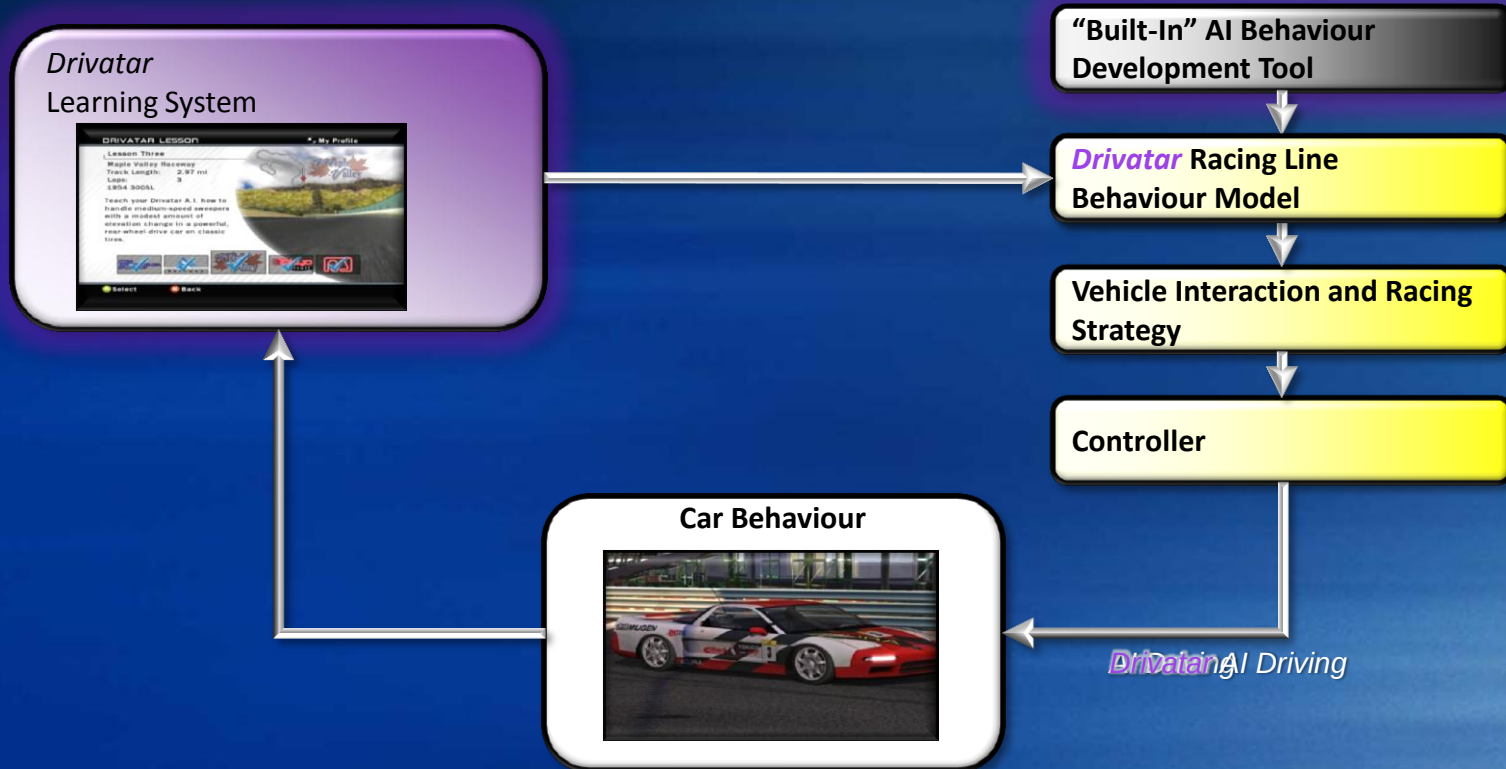
## XBOX Game

- Dynamic Racing Line
- Learning a Drivatar
- Using a Drivatar

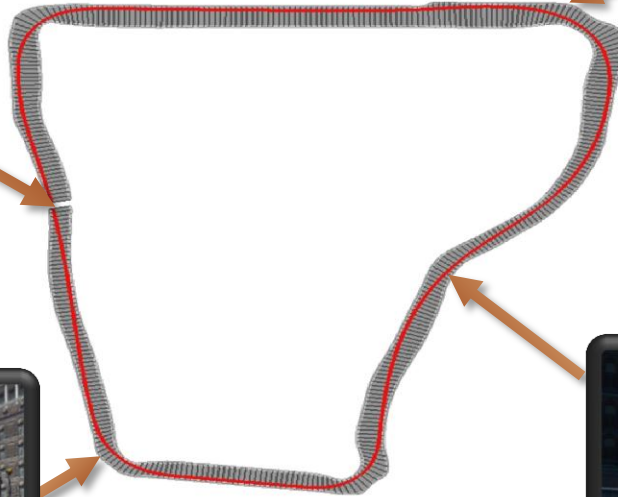




# Drivatars Unplugged



# The Racing Line Model



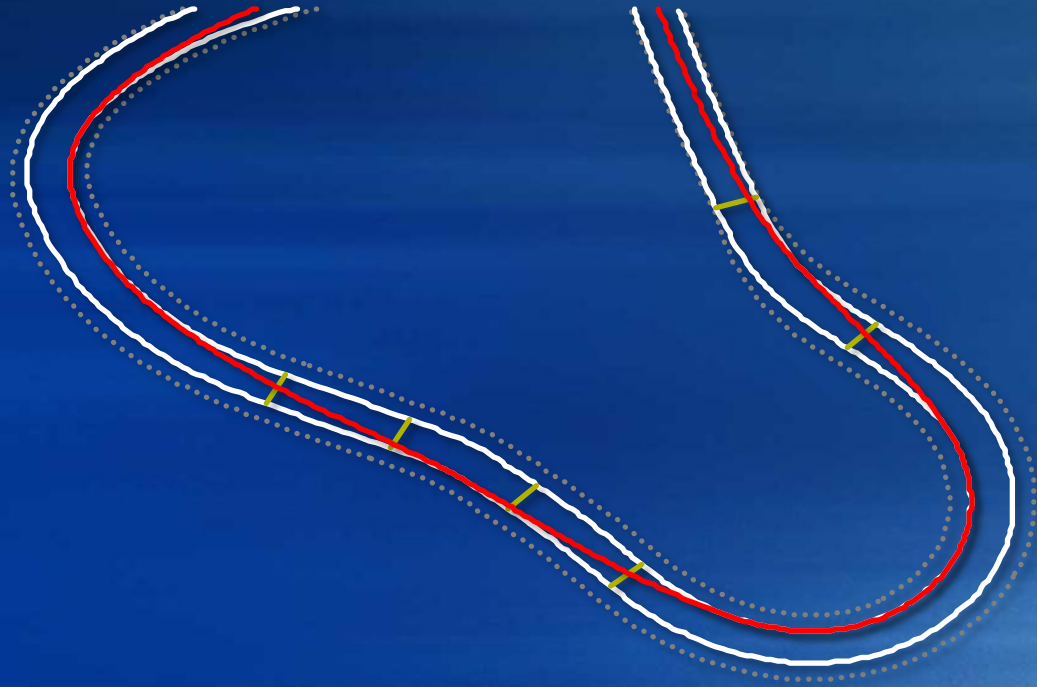
# Drivatars: Main Idea

- Two phase process:
  1. Pre-generate possible racing lines prior to the race from a (compressed) racing table.
  2. Switch the lines during the race to add variability.
- Compression reduces the memory needs per racing line segment
- Switching makes smoother racing lines.

# Racing Tables

| Segments                                                                          | $a_1$                                                                             | $a_2$                                                                               | $a_3$                                                                               | $a_4$                                                                               |
|-----------------------------------------------------------------------------------|-----------------------------------------------------------------------------------|-------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------|-------------------------------------------------------------------------------------|
|  |  |  |  |  |
|  |  |   |  |  |
|  |  |   |  |  |
|  |  |  |  |  |

# Minimal Curvature Lines

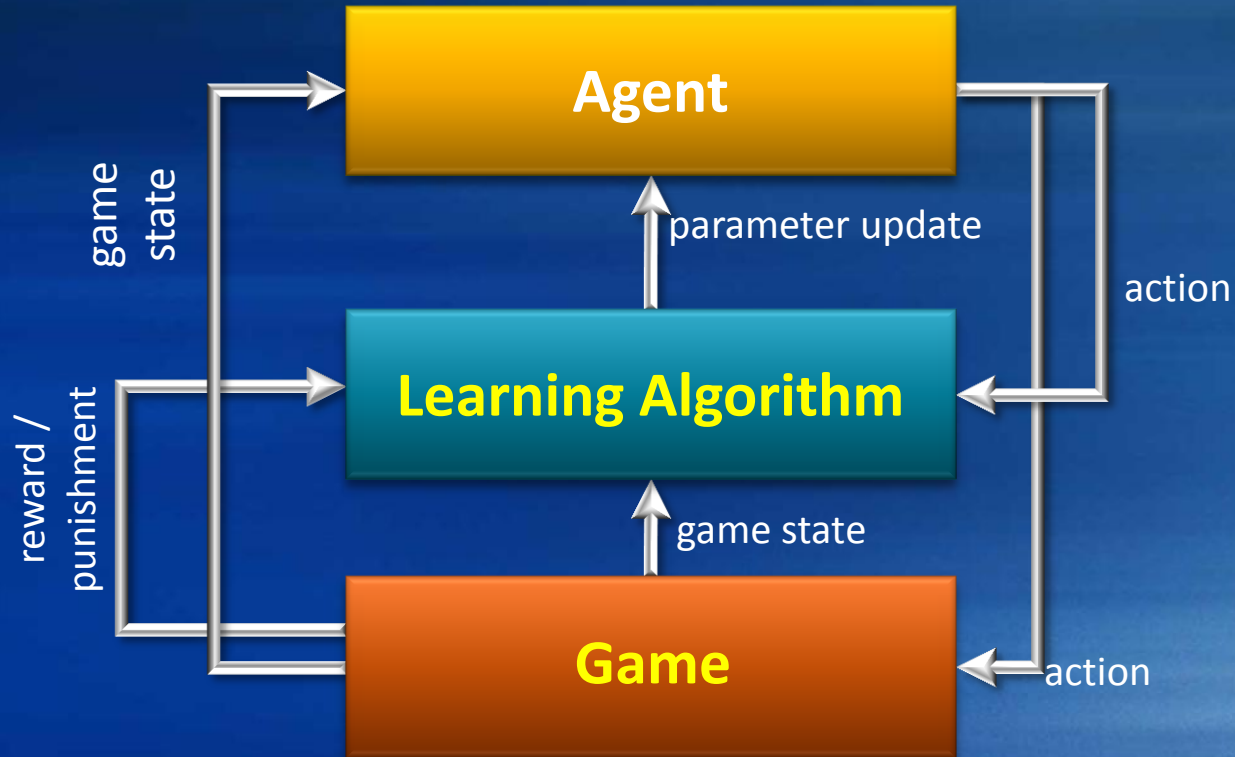




# Overview

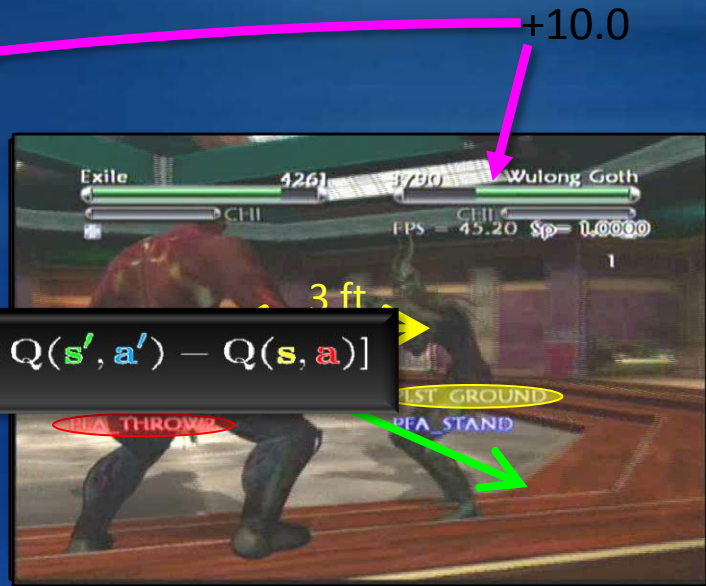
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# Reinforcement Learning



# Tabular Q-Learning

|             | Q-Table       | actions |      |       |
|-------------|---------------|---------|------|-------|
|             |               | THROW   | KICK | STAND |
| game states | 1ft / GROUND  |         |      |       |
|             | 2ft / GROUND  |         |      |       |
|             | 3ft / GROUND  | 13.2    | 10.2 | -1.3  |
|             | 4ft / GROUND  |         |      |       |
|             | 5ft / GROUND  |         |      |       |
|             | 6ft / GROUND  |         |      |       |
|             | 1ft / KNOCKED |         |      |       |
|             | 2ft / KNOCKED |         |      |       |
|             | 3ft / KNOCKED |         |      |       |
|             | 4ft / KNOCKED |         |      |       |
|             | 5ft / KNOCKED | 3.2     | 6.0  | 4.0   |
|             | 6ft / KNOCKED |         |      |       |



# Results

## ● Game state features

- Separation (5 binned ranges)
- Last action (6 categories)
- Mode (ground, air, knocked)
- Proximity to obstacle

## ● Available Actions

- 19 aggressive (kick, punch)
- 10 defensive (block, lunge)
- 8 neutral (run)

## ● Q-Function Representation

- One layer neural net (tanh)

### Reinforcement Learner



### In-Game AI Code



# Learning Aggressive Fighting

Reward for decrease in Wulong Goth's health

Early in the learning process ...



... after 15 minutes of learning





# Learning “Aikido” Style Fighting

Punishment for decrease in either player's health

Early in the learning process ...



... after 15 minutes of learning



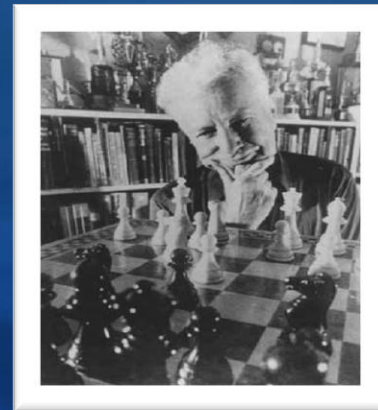


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# Motivation

- Competition is central to our lives
  - Innate biological trait
  - Driving principle of many sports
- Chess Rating for fair competition
  - ELO: Developed in 1960 by Árpád Imre Élő
  - Matchmaking system for tournaments
- Challenges of online gaming
  - Learn from few match outcomes efficiently
  - Support multiple teams and multiple players per team



# The Skill Rating Problem

- Given:

- Match outcomes: Orderings among  $k$  teams consisting of  $n_1, n_2, \dots, n_k$  players, respectively

- Questions:

- Skill  $s_i$  for each

- Glob

- Fair m

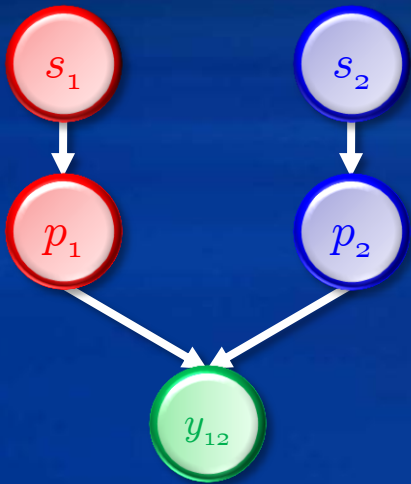
| Rank | Level | Gamertag        | Avg. Life | Best Spree | Score |
|------|-------|-----------------|-----------|------------|-------|
| 1st  | N/A   | SniperEye       | N/A       | N/A        | 25    |
| 2nd  | N/A   | xXxHALOxXx      | N/A       | N/A        | 24    |
| 3rd  | N/A   | AjaySandhu      | N/A       | N/A        | 15    |
| 3rd  | N/A   | AjaySandhu(G)   | N/A       | N/A        | 15    |
| 5th  | N/A   | Robert115       | N/A       | N/A        | 11    |
| 5th  | N/A   | TurboNegro84(G) | N/A       | N/A        | 11    |
| 7th  | N/A   | TurboNegro84    | N/A       | N/A        | 5     |
| 8th  | N/A   | SniperEye(G)    | N/A       | N/A        | 1     |

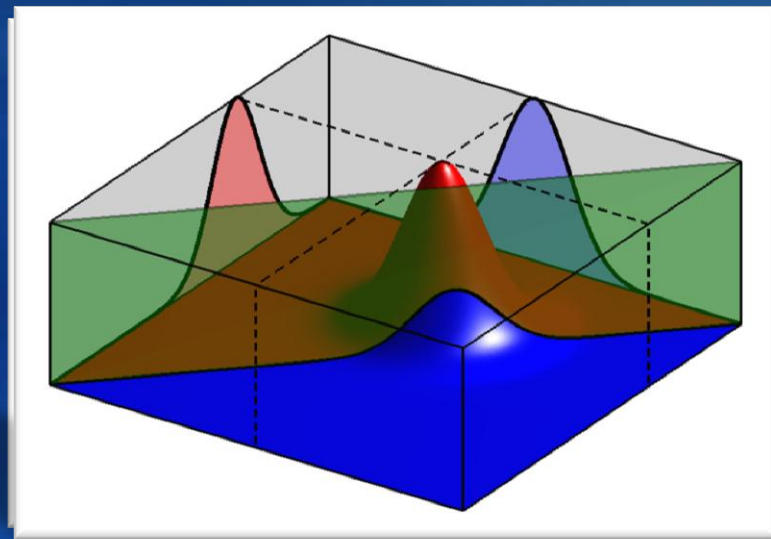
| Rank | Score | Player Name     |
|------|-------|-----------------|
| 1    | 27    | SEWICSYDE OWNS  |
| 2    | 26    | FATAL REVENGE   |
| 3    | 25    | Paranoia 1      |
| 4    | 25    | Paulk           |
| 5    | 25    | lxX OMG Xxl     |
| 6    | 25    | BittyTom        |
| 7    | 24    | brian 2007      |
| 8    | 24    | SEXY MOZES      |
| 9    | 24    | droplates       |
| 10   | 24    | jaCKdaSaMuRai   |
| 11   | 24    | ll Me ll        |
| 12   | 24    | iamNightMare    |
| 13   | 24    | a retarded007   |
| 14   | 24    | Perfected Brit  |
| 15   | 24    | THE MUFFIN MANx |
| 16   | 23    | TheVunit        |
| 17   | 23    | Mr Sushi87      |

# Two Player Match Outcome Model

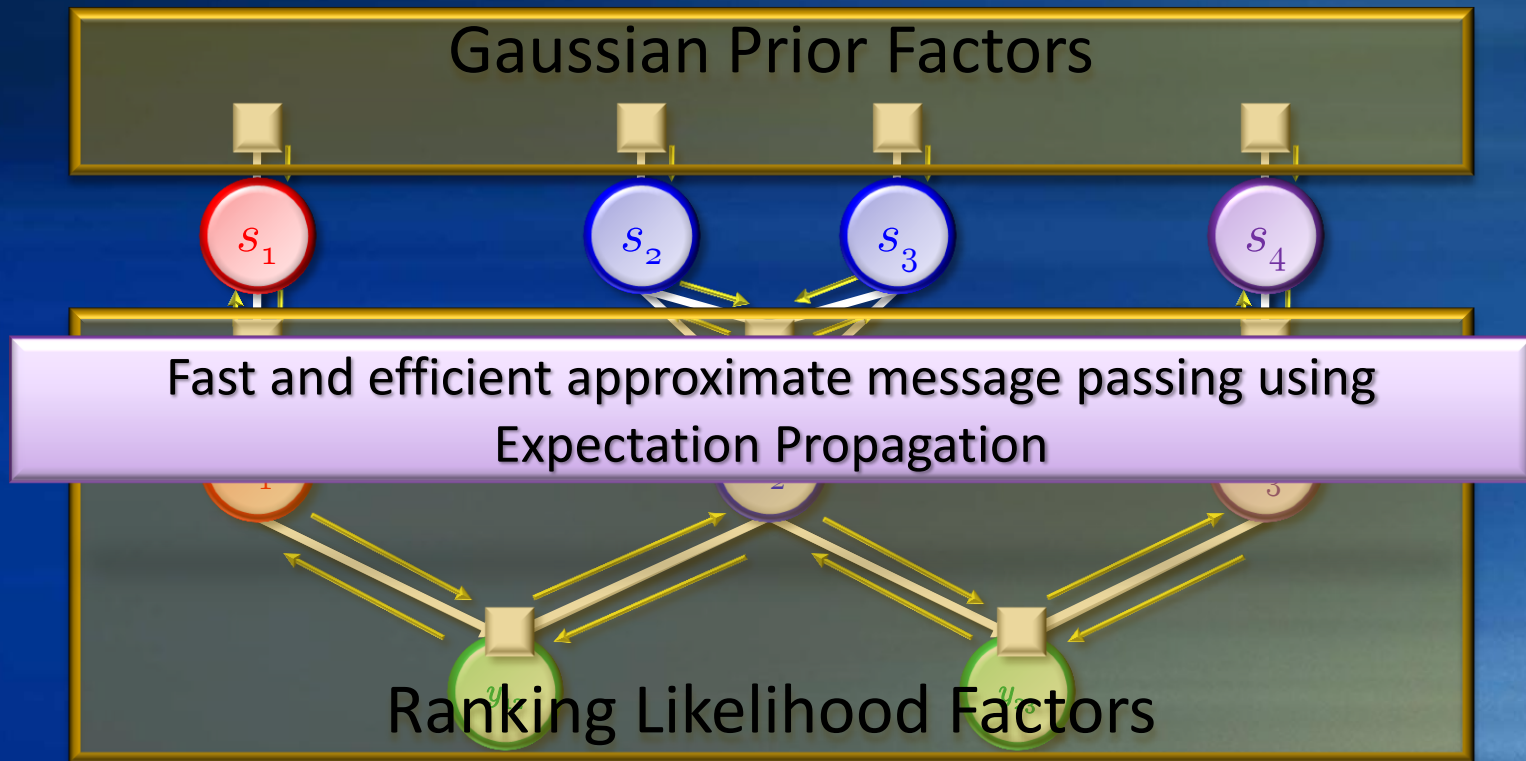
- Latent Gaussian performance model for fixed skills
- Possible outcomes: Player 1 wins over 2 (and vice versa)



$$\mathbf{P}(y_{12} = (1, 2) | p_1, p_2) = \mathbb{I}(p_1 > p_2)$$



# Efficient Approximate Inference





# Applications to Online Gaming

- **Leaderboard**

- Global ranking of all players

- **Matchmaking**

- For gamers: Most uncertain outcome
  - For inference: Most informative
  - Both a

$$\mu_i - 3 \cdot \sigma_i$$

|     | Level | Gamertag        | Avg. Life | Best Spree | Score |
|-----|-------|-----------------|-----------|------------|-------|
| 1st | 10    | BlueBot         | 00:00:49  | 6          | 15    |
| 1st | 7     | SniperEye       | 00:00:41  | 4          | 14    |
| 1st | 9     | ProThepirate    | 00:01:07  | 3          | 13    |
| 1st | 10    | dazdemon        | 00:00:59  | 3          | 8     |
| 2nd | 10    | Warddarry       | 00:00:41  | 2          | 12    |
| 2nd | 3     | Astia           | 00:00:47  | 2          | 10    |
| 2nd | 9     | Antidote4Losing | 00:00:41  | 2          | 9     |
| 2nd | 2     | Black Knight    | 00:00:48  | 3          | 13    |

$P(p_i \approx p_j | \mu_i = \mu_j, \sigma_i^2 + \sigma_j^2)$   
 $P(p_i \approx p_j | \mu_i - \mu_j = 0, \sigma_i^2 + \sigma_j^2 = 0)$

|   |    |                 |
|---|----|-----------------|
| 1 | 27 | SEWICSYDE OWNS  |
| 2 | 26 | FATAL REVENGE   |
| 3 | 25 | Paranoia 1      |
| 4 | 25 | Paulk           |
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| 7 | 24 | brian 2007      |
|   | 24 | SEXY MOZES      |
|   | 24 | droplates       |
|   | 24 | jaCKdaSaMuRai   |
|   | 24 | Il Me Il        |
|   | 24 | iamNightMare    |
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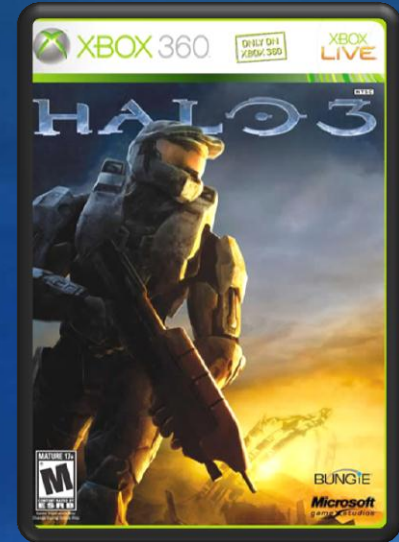
# Xbox 360 & Halo 3

- **Xbox 360 Live**

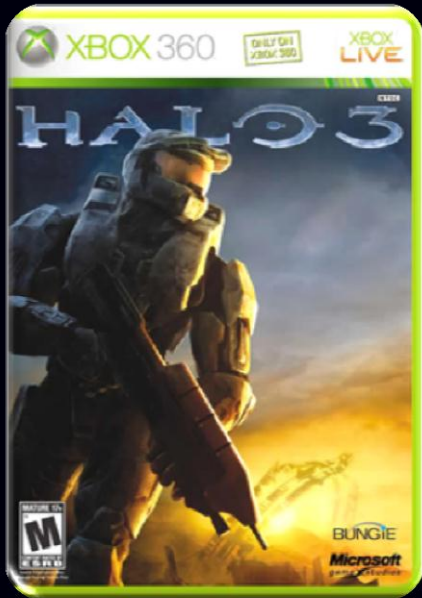
- Launched in September 2005
- Every game uses TrueSkill™ to match players
- > 35 million players
- > 4 million matches per day
- > 2 billion hours of gameplay / month

- **Halo 3**

- Launched on 25<sup>th</sup> September 2007
- Largest entertainment launch in history
- > 200,000 player concurrently (peak: 1,000,000)



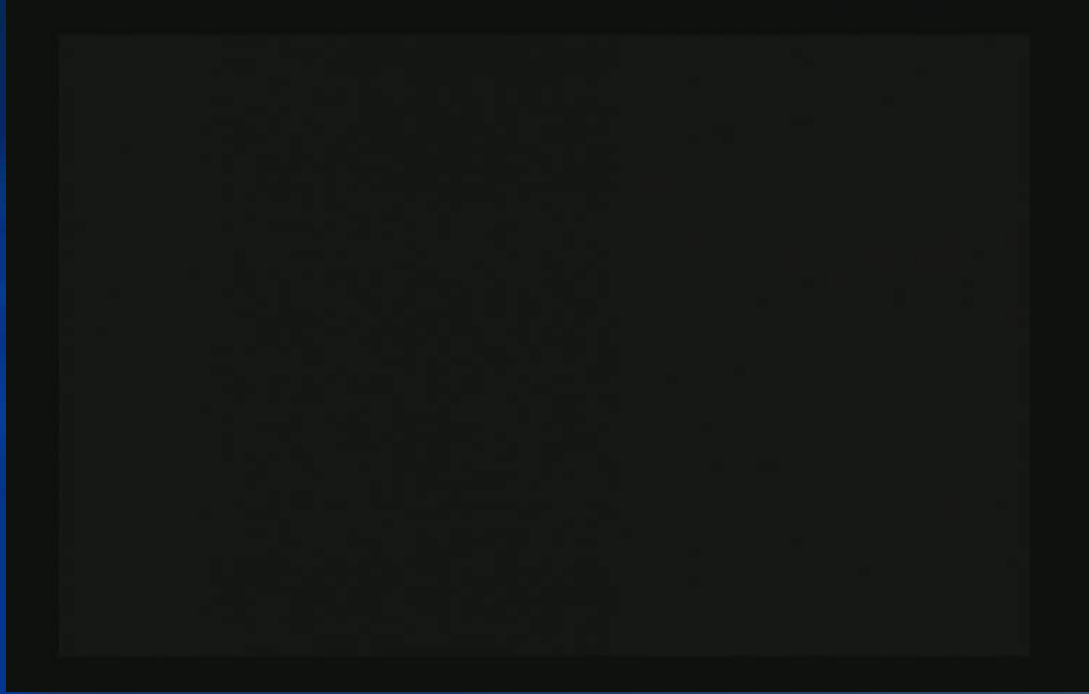
# Demo: Halo 3



## Halo 3 Game

- Matchmaking
- Skill Stats
- Tight Matches

# Halo 3 in Action



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# Learning to Play Go





# The Game of Go

- Started about 4000 years ago in ancient China.
- About 60 million players worldwide.
- 2 Players: Black and White.
- Board: 19×19 grid.
- Rules:
  - Turn: stone placed on vertex.
  - Capture.
- Aim: Gather territory



# Computer Go

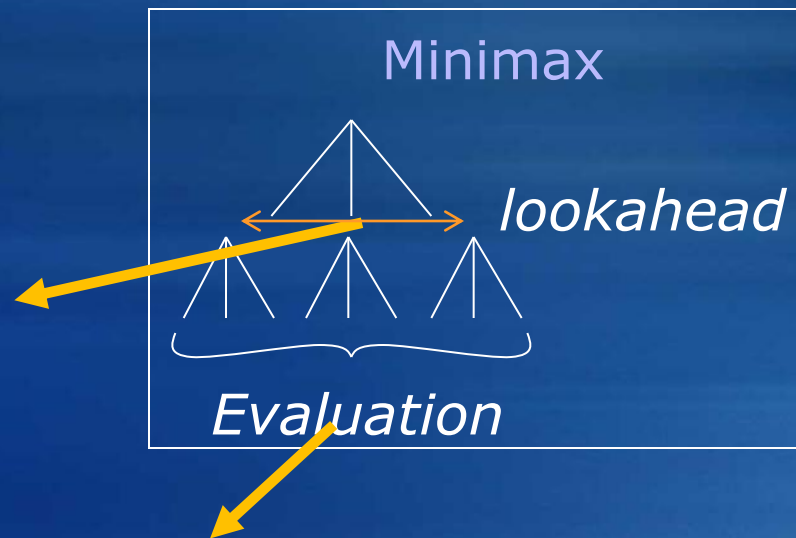
- 5th November 1997:  
Gary Kasparov beaten by Deep Blue.



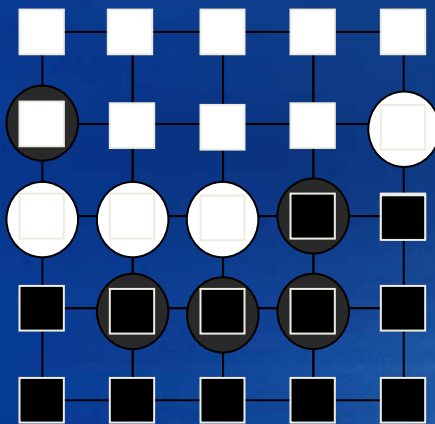
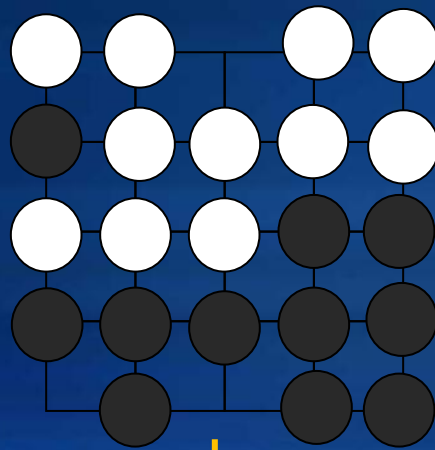
- Best Go programs cannot beat amateurs.

# Computer Go

- Minimax search defeated.
- **High Branching Factor.**
  - Go: ~200
  - Chess: ~35
- **Complex Position Evaluation.**
  - Stone's value derived from configuration of surrounding stones.

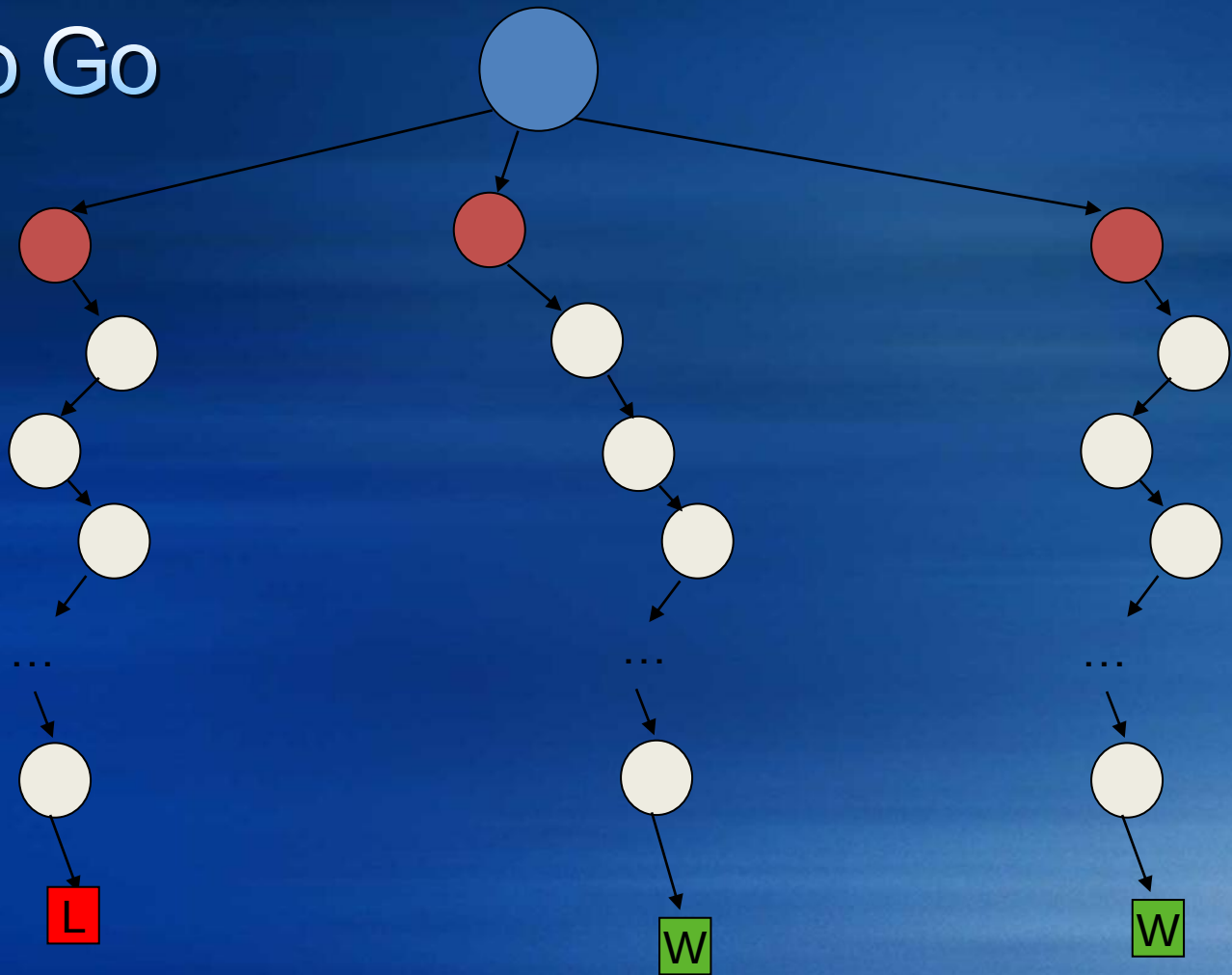


# Monte Carlo Go



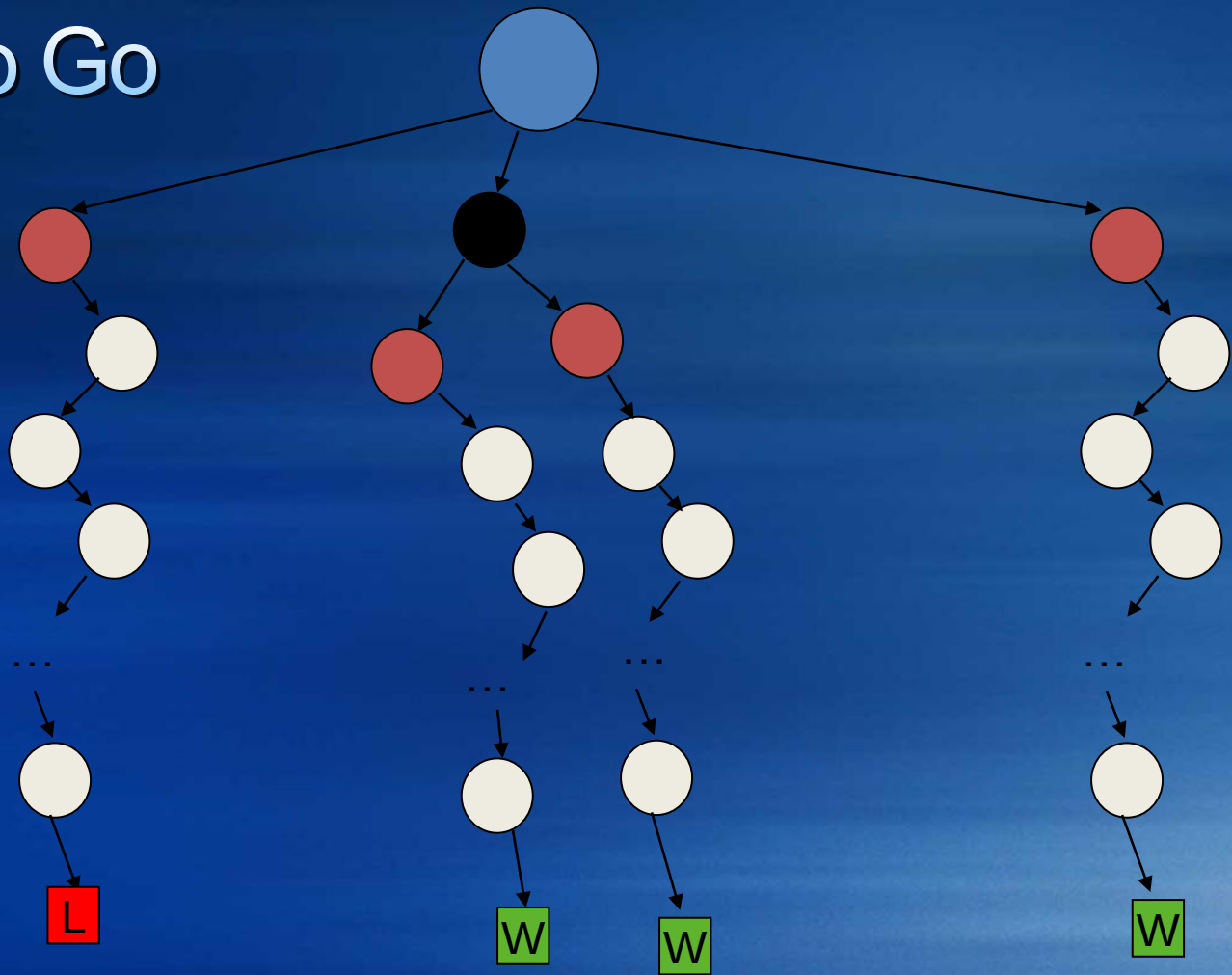
Territory Hypothesis

# Monte Carlo Go

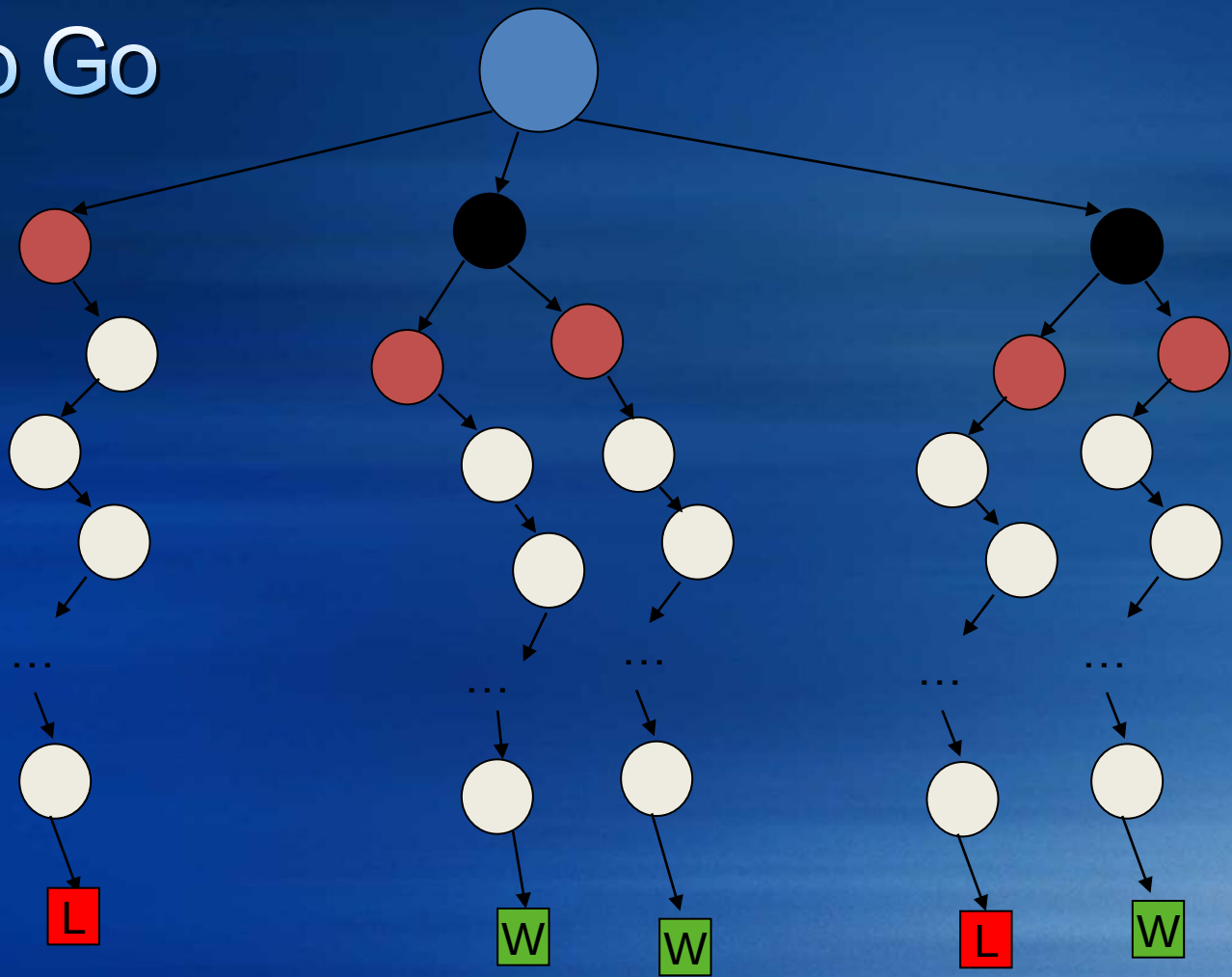




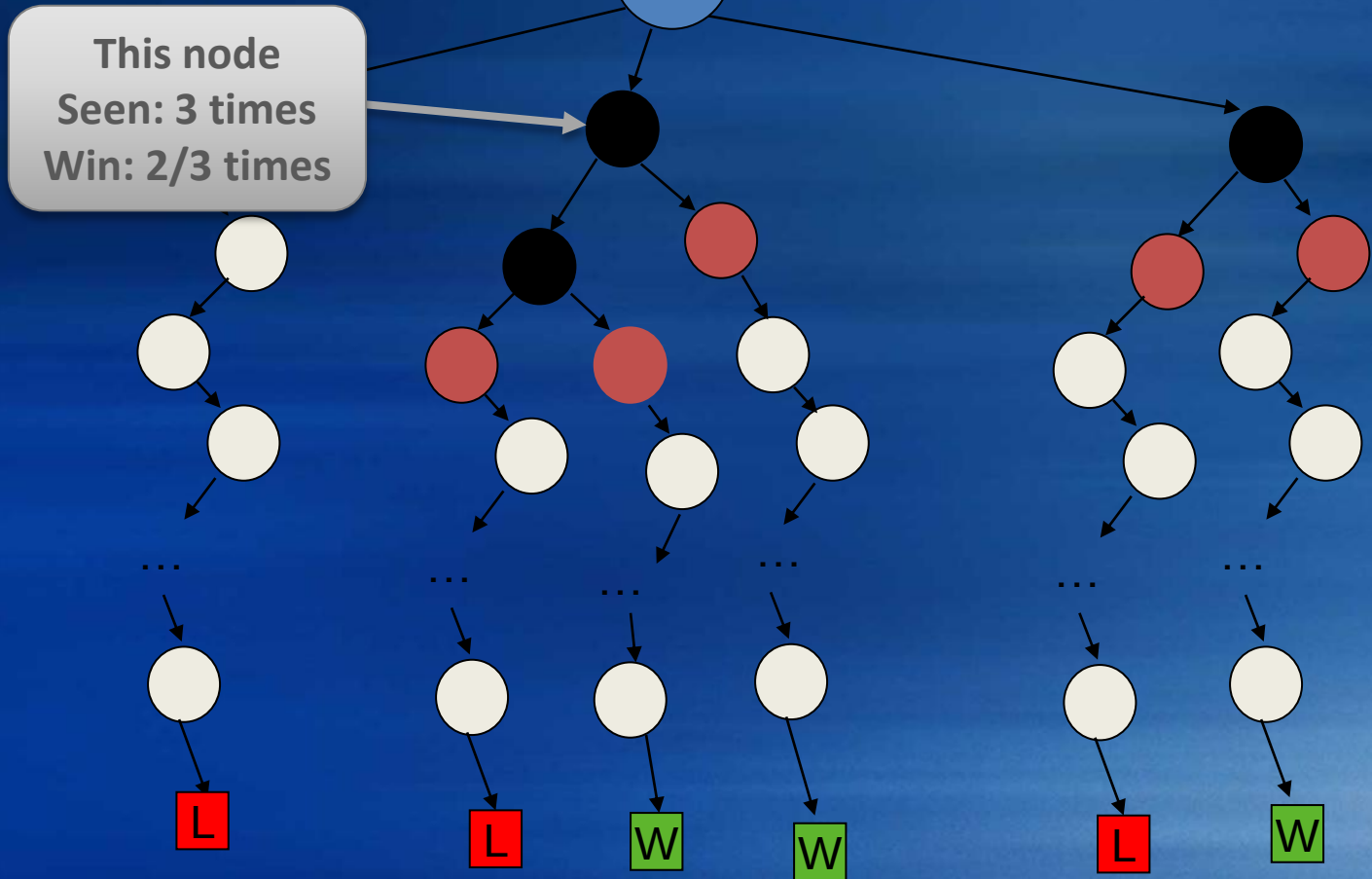
# Monte Carlo Go



# Monte Carlo Go



# Monte Carlo Go

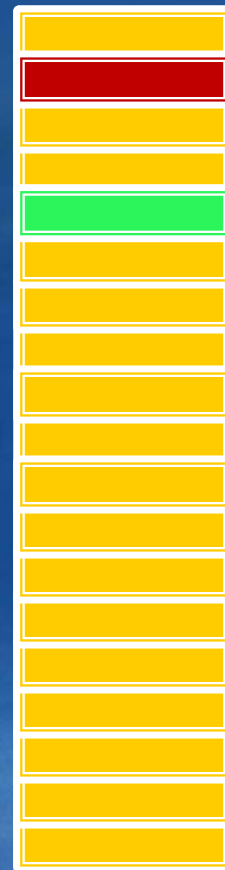
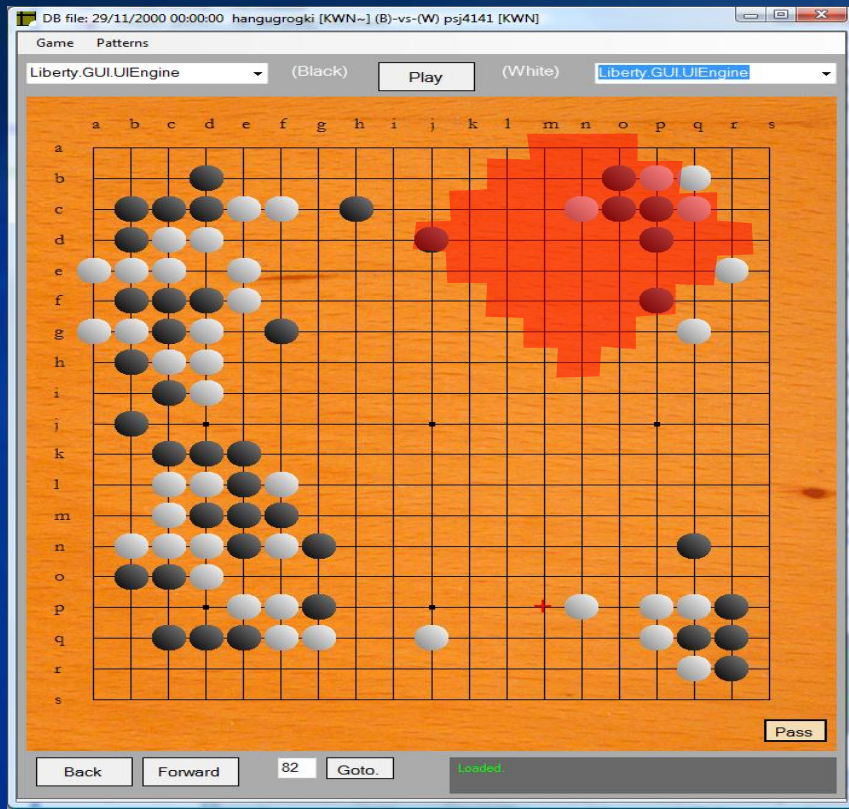


**This node**  
**Seen: 3 times**  
**Win: 2/3 times**

# Learning From the Experts with TrueSkill



# Learning From the Experts with TrueSkill



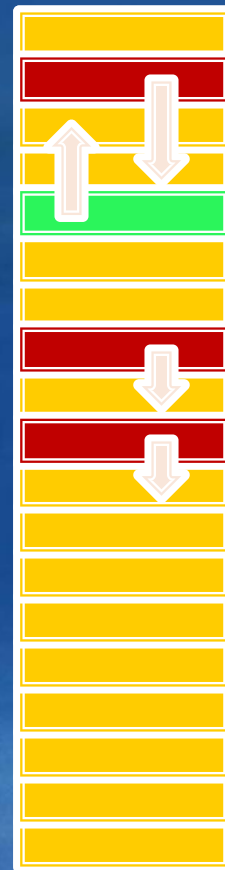


# Learning From the Experts with TrueSkill

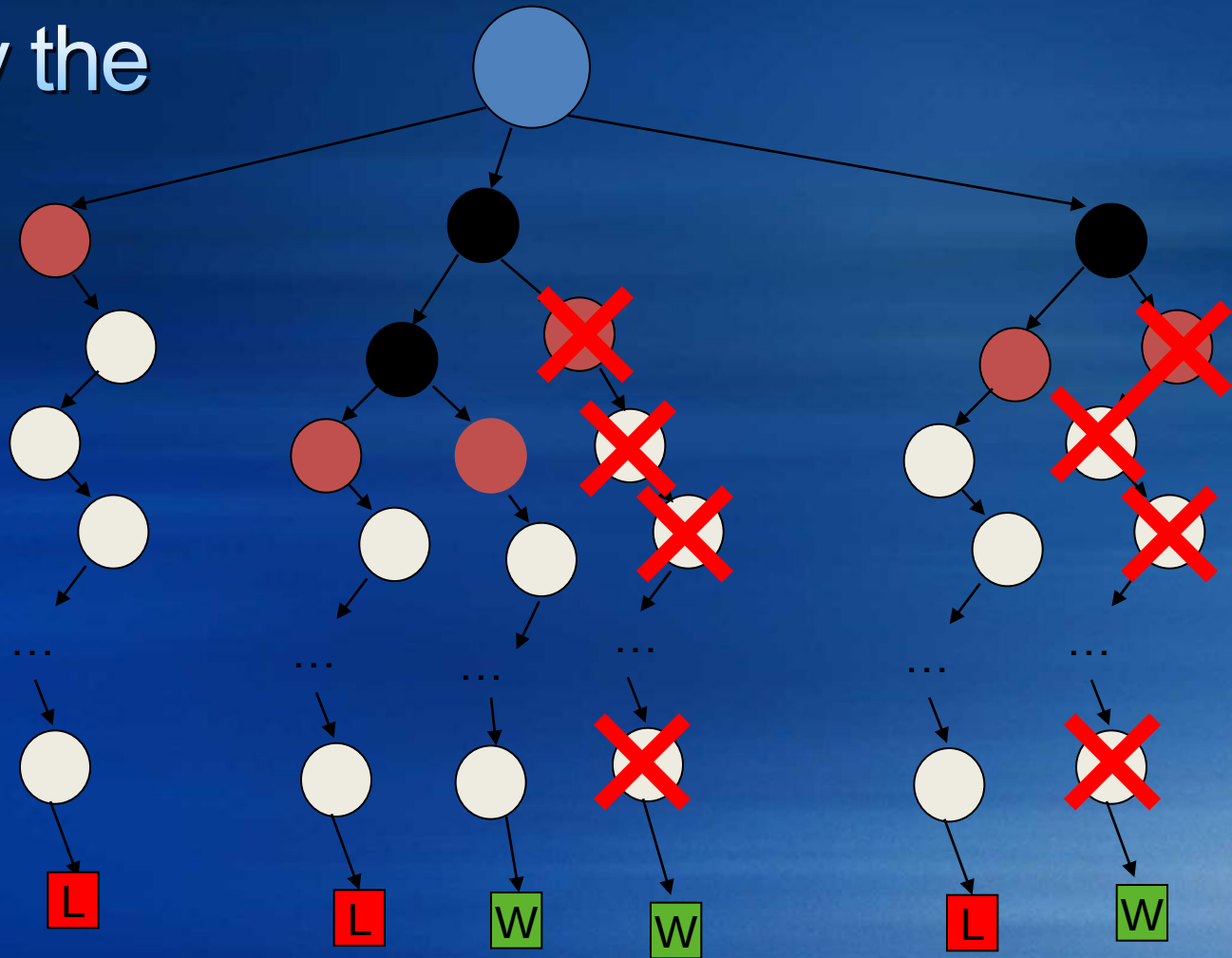




# Learning From the Experts with TrueSkill

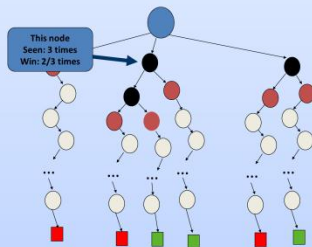
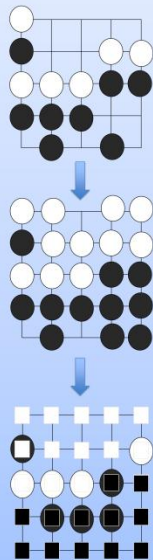


# Prune Away the Bad Moves



# Machine Learning Assisted Monte Carlo Go

## Monte Carlo Go



- Play random games ('rollouts').
- Each game gives a sample win/loss.
- Average estimates position value.
- Store game tree in memory.
- Bootstrap rollout policy.

## Pattern Ranking System



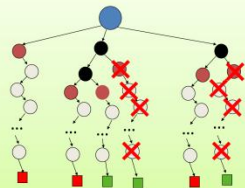
Pattern Table



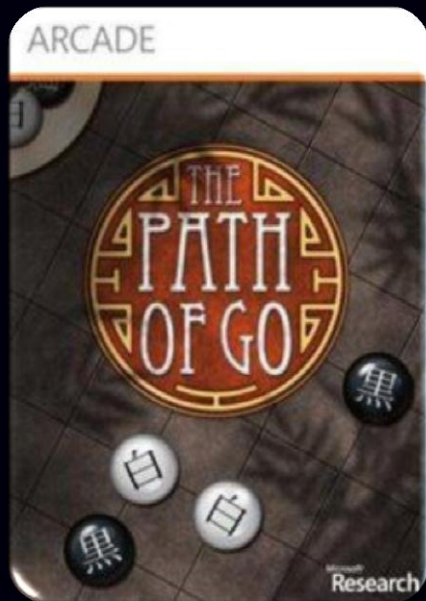
- Learn pattern rankings using TrueSkill.
- Moves chosen over other moves by experts are inferred to have higher value.
- Training Data: 200,000 Expert Go Games.

## Pattern Pruning

- Too many possible moves to evaluate.
- Pattern system estimates move quality.
- Prune bad moves from the game tree.



# Demo: The Path of Go



## The Path of Go

- MSRC Go AI (written in F#)
- TrueSkill Match Making
- XNA Game Studio

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# Conclusions

- Computer games can be used as test beds for research.
- Machine learning can be used to improve the user experience in computer games.
- Both research and applications are in their infancy and there are many open questions.
- XNA framework exists to plug in machine learning algorithms.
- For more question, please drop us a line

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