

# Search Pathology in the Game of Go

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Januar 31, 2006

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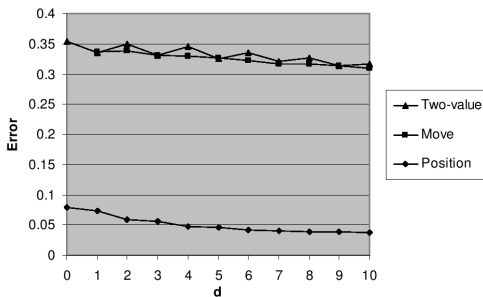
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# Search Pathology

- ▶ Minimax search in game trees can **degrade** quality of the backed-up heuristic evaluation (Nau, 1979; Beal, 1980)
- ▶ Usually **not observed** in real game-playing programs
- ▶ Absence was explained by
  - ▶ **Similarity** of values in sibling nodes (Beal, 1982; Bratko, 1982)
  - ▶ Early **terminal** positions (Pearl, 1984)
  - ▶ **Optimistic** evaluation function (Schrüfer, 1986)

Recent paper (Luštrek, Bratko, Gams, 2005):



- ▶ Disappears in model with **real-valued** evaluation function
- ▶ Win/loss error shows an **odd-even** effect

An **odd-even pathology** was also observed by (Nau, 1986) using minimax for a certain class of games

- ▶ Studies use **generated** noisy evaluation function applied to **small artificial** game trees.
- ▶ It can be expected that some games are more **susceptible** to pathology than others
- ▶ **Experimental data** by real game-playing programs rare

# Computer Go

- ▶ Misjudging status of a group introduces **error**, which can both over- and underestimate real value by a **large amount**
- ▶ Many Go programs do **not** use global search at all
- ▶ Go game trees have a nearly **uniform** branching factor

# NeuroGo

- ▶ Neural network based
- ▶ Learns by self-play
- ▶ Uses TD-learning for prediction of local connectivity and local reward
- ▶ Tournament version uses an extended set of input features
- ▶ Nega-max implementation of alpha-beta search  
Standard move ordering techniques; iterative deepening
- ▶ All legal moves are generated, apart from safe territory (threshold)

# Goal

- ▶ Examine *statistical properties* of NeuroGo's evaluation function
- ▶ Study *performance* depending on *search depth*

Error of evaluation function is unknown,  
but higher error will cause fewer wins



# Game Playing

- ▶ Opponent programs supporting [Go Text Protocol](#)
- ▶ [9×9](#) board
- ▶ [Komi 6.5](#)
- ▶ 21 balanced four-move [opening positions](#)
- ▶ [200 games](#) against each opponent / for each depth 1–4
- ▶ [Alternate](#) color every 2nd game
- ▶ Exclude [duplicates](#)
- ▶ Use GNU Go for [scoring](#)
- ▶ Athlon XP 2800+/512 MB ( $\approx 300$  evaluations/sec)

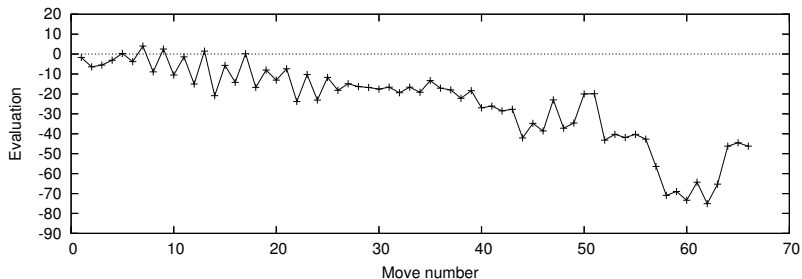
# Opponent Programs

- ▶ **Aya** 5.53, uses selective global search on  $9\times 9$
- ▶ **Crazy Stone** 0001-19, Monte Carlo
- ▶ **Explorer**, 6.6.x/Nov 3 2005, no global search
- ▶ **GNU Go** 3.6, no global search
  
- ▶ **NeuroGo** (depth 1)

# Statistical Properties

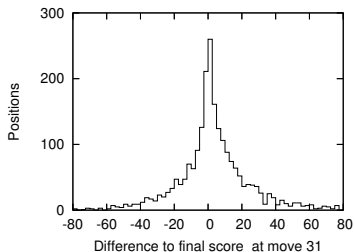
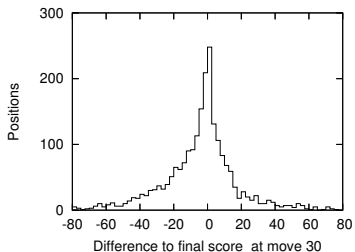
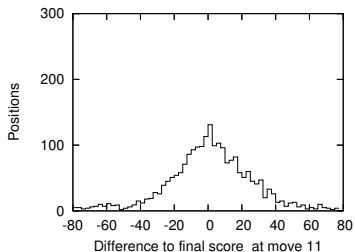
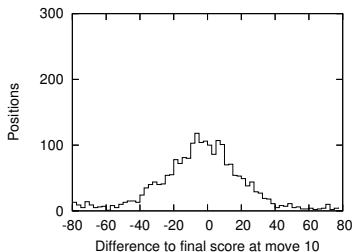
- ▶ All positions in games
- ▶ Black counts positive
- ▶ Black odd / White even move numbers
- ▶ NeuroGo automatically adds komi

## Example Game (NeuroGo vs Aya)



- ▶ Stable periods
- ▶ Slow changes
- ▶ Rapid changes
- ▶ Oscillations

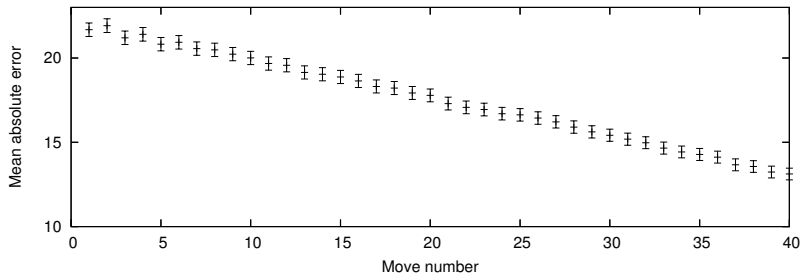
## Difference to final score



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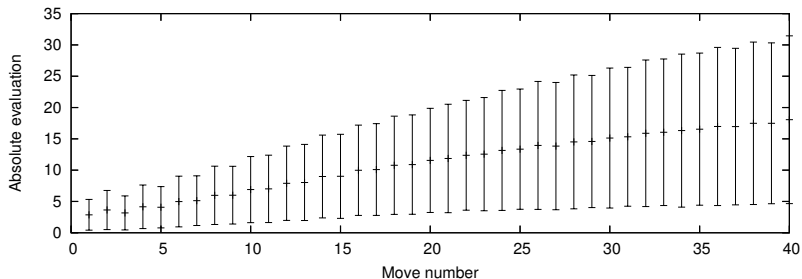
- ▶ Approximation of the **error** in evaluation
- ▶ **Gaussian** only early in game
- ▶ **Peak** around zero increases with move number
- ▶ Large **tails**
- ▶ **Similar**, but slight asymmetry for **Black/White** to move

## Mean absolute difference to final score



- ▶ Decreases nearly **linearly**
- ▶ Apart from very early positions, **no dependency** on color to move

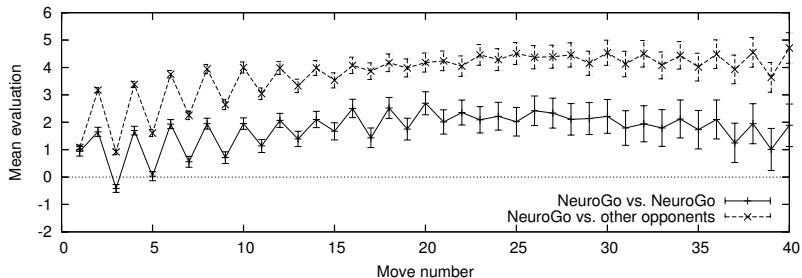
## Mean absolute evaluation (with deviation)



- ▶ **Increases** nearly **linearly** after move 5
- ▶ Smaller slope after move 20
- ▶ **Variance increasing** with move number

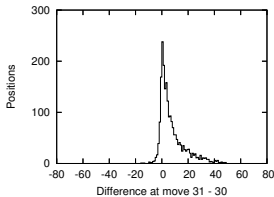
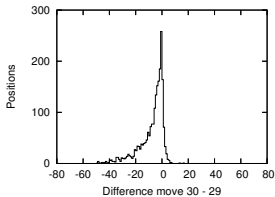
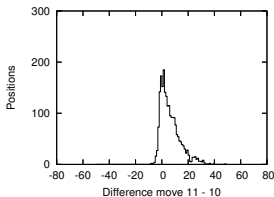
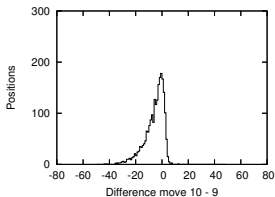


## Average evaluation for player to move



- ▶ On average **optimistic** for player to move (especially in middle game)
- ▶ Favours positions with **White** to move (komi over-compensates)

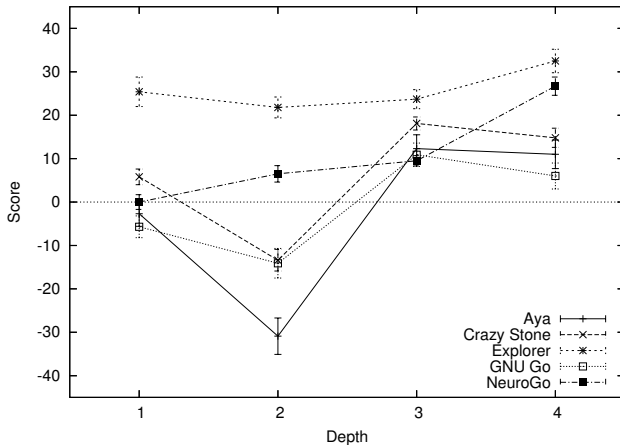
## Difference between subsequent moves



- ▶ Large number of positions with **stable** evaluation
- ▶ If **unstable**, greatly favours **player to move**

# Performance

## Average score



- ▶ Results vs Aya, Crazy Stone, GNU Go show a **pathology**
- ▶ Depth 2 worse than 1; 4 worse than 3
- ▶ **Not visible** in games against Explorer
- ▶ **Absent** in games against itself

# Conclusion

- ▶ **Evaluation function** of a medium-strength Go program was examined
- ▶ Evaluation was found **optimistic** for player to move, especially in **unstable positions**
- ▶ Difference to final score decreased **linearly** with move number
- ▶ Resembles **Gaussian** only for early positions
- ▶ **Search pathology** occurred
- ▶ Significant impact on **playing strength** at low depths
- ▶ Games vs fixed-depth version of same Go program are **not sufficient** to observe this effect