

# Chess Research and Public Interests

## New Faculty Academy 2026

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## First, some chess and education background on me

- Grew up in Paramus, NJ.
- Youngest USCF Master since Bobby Fischer, 1973.
- US Junior Champion (first equal), 1977.
- FIDE International Master title, 1981. (I am not a Grandmaster),
- BA (not BS) in Mathematics, Princeton University, 1981.
- Marshall Scholar to Oxford (Merton College), 1981–1984.
- Junior Fellow of Merton. 1984–1986 and 1988, interleaved with—
- Cornell MSI postdoc (under [Anil Nerode](#) and Juris Hartmanis), 1986–1987 and 1988–1989.
- Assistant Professor at UB, Department of Computer Science (now CSE) from August 1989; tenure in 1995.
- Tenure and great colleagues conferred blessings of Freedom...

## Research Background

- Combinatorial Mathematics, then Computational Complexity.
- Central problem: **P Versus NP**.
- Expert on logical status: is it **undecidable** a-la Kurt Gödel? How to surmount **barriers**?
- In 1997, I thought I had an inside track on solving (the “algebraic version” of) it.
- Quantum Computing neighbors complexity. **Shor’s Theorem**.
- Co-wrote **textbook** with Richard Lipton.
- Also co-wrote his weblog **Gödel’s Lost Letter and P=NP**, 2009–2024.
- All the while, I resisted entreaties to do computer chess.
- Then came the Topalov-Kramnik 2006 World Championship **Scandal**.

# My Reactions

- ① (Try to) Reproduce the **Accusation**.
- ② Observe Facts About Chess Programs:
  - As the search progresses, they “change their minds” about which move is best.
  - List top move at each depth of search and whether it “matches” the move that was played (MM).
  - (Un)fair to count a “coincidence” if the played move matches at any depth?
  - Can set chess program to give values of multiple possible moves.
- ③ Simple Principle: The wider the value gap between the best move and any other move, more likely a strong human player will find it.
- ④ Judge a move “forced” if the value loss of the next-best move (“delta”) is catastrophic.
- ⑤ Also cases where multiple moves have Equal-optimal Value (EV).
- ⑥ **Tally** MM and EV cases and deltas over the games.
- ⑦ Find **people** to **talk** to...and **compare** results.

# The Fortune and Blessings of Simplicity

- How to build a *quantitative* model out of the simple principle?
- Simple elements **Strategy** and **Tactics** take us far.
- **Depth of Thinking** should be next.
- Do weaker players **prefer** weaker moves?
- Or are they more easily **distracted**?
- How shall we handle the element of **Difficulty**?
- **Recognition** “Versus” **Thinking**.
  - See the 2007 National Geographic documentary “**My Brilliant Brain**” with Susan Polgar (**crux here**).
  - We will try to glean comparable insight from numerical analytics.

# A Predictive Analytic Model

Means that the model:

- Addresses a series of events or decisions, each with possible outcomes  $m_1, m_2, \dots, m_j, \dots$
- Assigns to each  $m_j$  a probability  $p_j$ .
- Projects risk/reward quantities associated to the outcomes.
- Also assigns *confidence intervals* for  $p_j$  and those quantities.

In a **utility-based** model, each  $m_i$  has a utility or cost  $u_i$ .

Main risk/reward quantity then becomes  $E = \sum_i p_i u_i$ .

- **Insurance:**  $m_i$  are risk factors; costs  $u_i$  need not influence  $p_i$ .
- **Chess:**  $m_i$  are legal moves;  $u_i$  are engine values and influence  $p_i$ .
- **Multiple-choice tests:**  $m_i$  are possible answers to a test question,  $u_i = \text{gain/loss for right/wrong answer}$ .

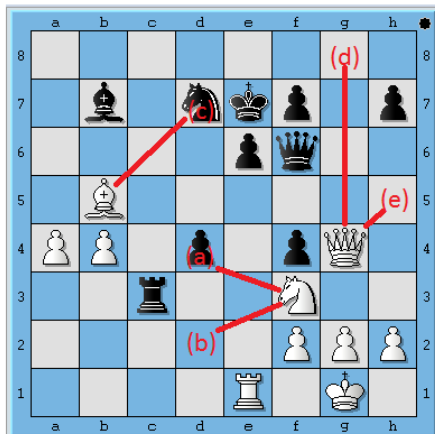
# Chess and Tests—With Partial Credits (Or LLMs?)

The \_\_\_\_ of drug-resistant strains of bacteria and viruses has \_\_\_\_ researchers' hopes that permanent victories against many diseases have been achieved.

- (a) vigor . . . corroborated
- (b) feebleness . . . dashed
- (c) proliferation . . . blighted
- (d) destruction . . . disputed
- (e) disappearance . . . frustrated

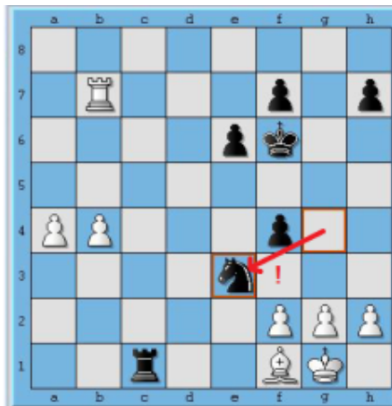
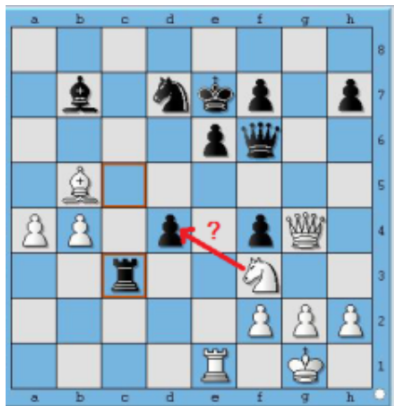
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Here (b,c) are **equal-optimal** choices, (a) is bad, but (d) and (e) are reasonable—worth part credit.

# Move Utilities Example (Kramnik-Anand, 2008)



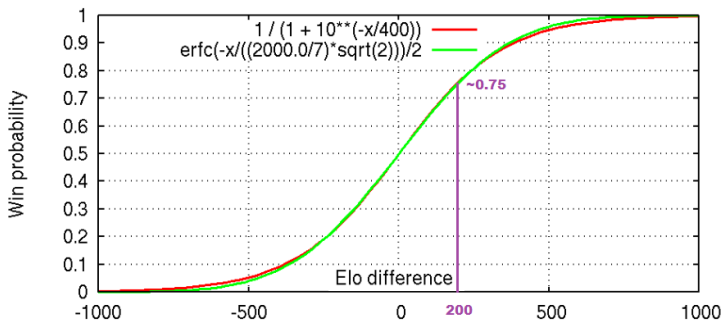
Depths...

Values by Stockfish 6

Move	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18	19
Nd2	103	093	087	093	027	028	000	000	056	-007	039	028	037	020	014	017	000	006	000
Bxd7	048	034	-033	-033	-013	-042	-039	-050	-025	-010	001	000	-009	-027	-018	000	000	000	000
Qg8	114	114	-037	-037	-014	-014	-022	-068	-008	-056	-042	-004	-032	000	-014	-025	-045	-045	-050
...			...		...				...			...		...				...	
Nxd4	-056	-056	-113	-071	-071	-145	-020	-006	077	052	066	040	050	051	-181	-181	-181	-213	-213

## Aptitude—Via Elo Grades ([calculator](#))

- Named for **Arpad Elo**, number  $R_P$  rates skill of player  $P$ .
- E.g. **1000** = bright beginner, **1600** = good club player, **2200** = master, **2800** = world championship caliber.
- Computer **engines** are far higher, e.g.: **Stockfish 16 = 3544**, **Torch 1.0 = 3531**, **Komodo Dragon 3.3 = 3529**.
- Expectation given by rating *difference* via this logistic curve:



## Main Parameters and Inputs

The (only!) player parameters trained against chess **Elo Ratings** are:

- $s$  for “**sensitivity**”—strategic judgment. *Like Anatoly Karpov.*
- $c$  for “**consistency**” in tactical minefields. *Like Mikhail Tal.*
- $h$  for “**heave**” or “**Nudge**”—obverse to depth of thinking.

**Trained** on all available in-person classical games in 2010–2019 with both players near the same Elo marker 1025, 1050, . . . , 2775, 2800, 2825.

**Being retrained** on new FIDE range **1400** . . . 2825, **from 1/1/25 on**.

- Given an Elo rating  $R$ , “central slice” gives corresponding  $s_R, c_R, h_R$ .
- Only other input is the grid of move utilities  $u_{i,d}$  at various depths  $d$  of search, further **scaled** to make (perceived) values  $v_i$  (and  $\rho_i$ ).
- Then  $\delta_i = v_1 - v_i$  is difference to best move.
- Other than these, **my model knows nothing about chess**.

# One Wonky Slide: Log-Linear Versus Loglog-Linear

The generic **log-linear** model puts

$$\log\left(\frac{1}{p_i}\right) = \alpha + \beta u_i, \quad \text{or equivalently,} \quad \log\left(\frac{1}{p_i}\right) - \log\left(\frac{1}{p_1}\right) = \beta \delta_i$$

- Solved by **softmax** giving  $p_i = p_1 \cdot \exp(-\beta u_i)$ .
- Each  $p_i$  is represented as a **multiple** of the top probability  $p_1$ .
- Ubiquitous in AI—but **does not work for chess**.

The **loglog-linear** model puts  $\log \log\left(\frac{1}{p_i}\right) - \log \log\left(\frac{1}{p_1}\right) = \beta \delta_i$ , i.e.:

$$\frac{\log(1/p_i)}{\log(1/p_1)} = \exp(\beta \delta_i).$$

- Gives  $p_i = p_1^{\exp(\beta \delta_i)}$ .
- So  $p_i$  are represented as **powers** of the best-move probability  $p_1$ .
- In place of  $\beta \delta_i$ , I really have  $\left(\frac{\delta_i - h \rho_i}{s}\right)^c$ , with  $h$  tightly clamped.

## How The Model Operates

- Take  $s, c, h$  from a player's rating (or wider skill profile).
- Generate probability  $p_i$  for each legal move  $m_i$ .
- Paint  $m_i$  on a 1,000-sided die, **1,000** $p_i$  times.
- **Roll the die** to give confidence intervals that go with the  $p_i$ .
- (Correct after-the-fact for chess decisions not being independent.)

### Main Outputs:

- **Statistical z-scores** for various (*actual*–*projected*) quantities:
  - **T1-match**: Agreement with the move listed first by the computer.
  - **EV-match**: Includes moves of equal-optimal value not listed first.
  - **ASD**: Average *scaled* difference in value from inferior moves.
- An **Intrinsic Performance Rating (IPR)** for the set of games.

Fit  $s, c, h$  by making **T1, EV, ASD** be **unbiased estimators** on the training sets, which are stratified by Elo ratings.

## Karpov & Tal at Montreal “Tourney of Stars” 1979

- Tied for first with 12/18 in star-studded double round-robin.
- Karpov was rated **2705**, Tal only **2615**.
- Karpov (per Stockfish 11):  $s = 0.016$ ,  $c = 0.307$ .
- Tal (per Stockfish 11):  $s = 0.026$ ,  $c = 0.365$ .
- Lower  $s$  is better—so Karpov was more “Karpovian.”
- Higher  $c$  is better—so my model with Tal’s parameters would make fewer large mistakes.

Are these grainy parameters enough to mimic human tendencies?

- IPRs: Karpov **2625 +- 155**, Tal **2730 +- 185**.
- Whole tourney IPR is (only!) **2575 +- 50** ( $s = 0.041$ ,  $c = 0.385$ ).
- Average Elo of players, **2621**, is within error bars. Surprise is that the IPR is not near 2700s range. Today’s elite regularly hit 2800+.

## Simplicity and Public Outreach

- Originally I intended to use *distributional distance measures*, of which **fidelity** is one.
- I realized that results would be difficult to **explain**.
- Hence I mapped everything to rolling dice, with math known since 1800.
- Other simple matters: **Should Metrics Be Linear?**
- **“Pandemic Lag”** in updating ratings.
- Are their players who are majorly better, relative to their peers at slow chess, in fast chess?
- [Segue to public-outreach cases and demos]
- Q & A — And Thanks.