

1. Core Strategy Philosophy

Our bot makes decisions based on Expected Value (EV) maximization using Monte Carlo simulation to estimate win probabilities. Every action is chosen to maximize long-term profit by combining:

2. Expected Value: The Decision Framework

Early in the tournament (no data yet), we assume balanced play: 33% each action.

EV(FOLD) : -1

EV(CALL):

1. Opponent folds (probability = fold_rate): $\text{fold_rate} \times 2$
 2. Opponent calls (probability = call_rate): $\text{call_rate} \times (\text{prob_win} \times 2 + \text{prob_tie} \times 0 + \text{prob_lose} \times (-2))$
 3. Opponent raises (probability = raise_rate): $\text{raise_rate} \times (\text{prob_win} \times 2 + \text{prob_lose} \times (-3))$
- Total: $\text{EV(CALL)} = \text{fold_rate} \times 2 + \text{call_rate} \times (\text{prob_win} \times 2 - \text{prob_lose} \times 2) + \text{raise_rate} \times (\text{prob_win} \times 2 - \text{prob_lose} \times 3)$

EV(RAISE):

1. Opponent folds (probability = fold_rate): $\text{fold_rate} \times 3$
 2. Opponent calls (probability = call_rate): $\text{call_rate} \times (\text{prob_win} \times 3 + \text{prob_lose} \times (-2))$
 3. Opponent raises back (probability = raise_rate): $\text{raise_rate} \times (\text{prob_win} \times 3 + \text{prob_lose} \times (-3))$
- Total: $\text{EV(RAISE)} = \text{fold_rate} \times 3 + \text{call_rate} \times (\text{prob_win} \times 3 - \text{prob_lose} \times 2) + \text{raise_rate} \times (\text{prob_win} \times 3 - \text{prob_lose} \times 3)$

3. Decision Logic by Hand Strength

Category 4-5: Premium Hands (Trips, Straight Flush)

- If $\text{EV(RAISE)} \geq \text{EV(CALL)}$ OR $\text{prob_win} > 85\%$: **RAISE**
- Otherwise: **CALL**

Category 3: Strong Hands (Straight)

- If $\text{EV(RAISE)} > \max(\text{EV_CALL}, \text{EV_FOLD})$: **RAISE**
- Else if $\text{EV(CALL)} > \text{EV(FOLD)}$: **CALL**
- Otherwise: **FOLD**

Category 2: Medium Hands (Flush)

- If $\text{prob_win} > 60\%$ AND $\text{EV(RAISE)} > \text{EV(CALL)}$: **RAISE**
- Else if $\text{prob_win} > 45\%$ OR $\text{EV(CALL)} > 0$: **CALL**
- Otherwise: **FOLD**

Category 1: Weak Hands (Pair)

Against aggressive opponents (raise_rate > 50%) with prob_win < 50%:

- If $\text{EV(CALL)} > 0$: **CALL**
- Otherwise: **FOLD**

Category 0: High Card Only

- If $\text{prob_win} > 60\%$ AND $\text{fold_rate} > 40\%$: **CALL**
- Else if $\text{EV(CALL)} > 0.3$: **CALL**
- Otherwise: **FOLD**

4. Why This Strategy Beats Different Opponent Types

Against Aggressive Opponents (High raise_rate)

- EV(CALL) and EV(RAISE) both decrease because:
 - $\text{Raise_rate} \times (\text{prob_win} \times 2 - \text{prob_lose} \times 3)$ term becomes very negative
 - We're risking -3 points frequently
- Weak hands (categories 0-2) → EV(CALL) often becomes negative → we fold more
- Strong hands (categories 3-5) → EV still positive → we continue but cautiously

Result: We avoid big losses on weak hands while still extracting value from strong hands.

Against Passive Opponents (High fold_rate)

- EV(RAISE) increases dramatically because:
 - $\text{fold_rate} \times 3$ is large (free +3 points)
- Medium hands (categories 2-3) → EV(RAISE) often > EV(CALL) → we raise more
- Even weak hands → EV(CALL) becomes positive → we call more often

Result: We steal pots aggressively when opponent gives up easily.

Against Balanced Opponents

- $\text{fold_rate} = \text{call_rate} = \text{raise_rate} \approx 0.33$
- EV calculations reflect true hand strength via prob_win
- Decisions become purely probability-driven

Result: We play Game Theory Optimal (GTO) poker.

Conclusion

Our strategy combines:

1. **Monte Carlo simulation** → Accurate win probabilities for specific hands
2. **Expected Value calculation** → Optimal action choice incorporating opponent behavior
3. **Hand-strength-based logic** → Risk management and value extraction

This approach makes the mathematically correct decision in every situation, adapting to opponent play style while maximizing long-term profit.

