

Can Mistral LLM win at Poker?

Mistral Fine-tuning Hackathon, June 2024

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Plan

- 1. Inspiration and context
- 2. The case of Poker
- 3. Our methodology
- 4. Results and performance
- 5. Key improvements areas





Recent experiences show that GPT 3.5 Turbo trained on standardized chess games achieves 1800 ELO

Inspiration & context

LLM fine-tuned on PGN chess games:

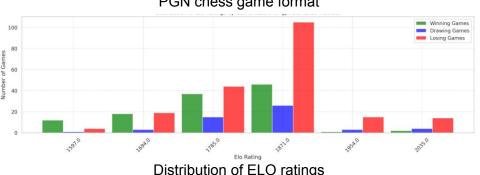
- Returns the most likely next move
- Adapts its style to the previous sequence of moves.
- Is able to play chess without knowing the rules, with few forbidden moves.
- Outperforms ChatGPT and reaches impressive 1800 ELO

```
[Event "F/S Return Match"]
[Site "Belgrade, Serbia JUG"]
[Date "1992.11.04"]
[Round "29"]
[White "Fischer, Robert J."]
[Black "Spassky, Boris V."]
[Result "1/2-1/2"]

1. e4 e5 2. Nf3 Nc6 3. Bb5 a6 4. Ba4 Nf6 5. 0-0 Be7 6. Re1 b5 7. Bb3 d6 8. c3 0-0 9. h3 Nb8 10. d4 Nbd7 11. c4 c6 12. cxb5 axb5 13. Nc3 Bb7 14. Bg5 b4 15.

Nb1 b6 16 Bb4 c5 17 dve5 Nve4 18 Bve7 Ove7 19 evd6 Of6 20 Nbd2 Nvd6 21

PGN chess game format
```



Finetuning an LLM the same way could be interesting for Poker specificities (missing information, bluffing, etc.)

The case of Poker

Poker specificities



Missing information:

Not all knowledge of the environment is available (hidden cards).



Bluffing and psychology:

Poker involves bluffing and not playing like a robot to remain unpredictable.



Non-deterministic environment:

Poker involves chance with card shuffling and dealing.



Multi player:

Poker games involve a dynamic number of players

Opportunities for LLMs

Games format:

Poker games can be structured in a PGN-like format, with contextual information and a sequence of actions.

Adaptable style of play:

LLMs may understand opponents' style of play and exploit advantageous situations.

A 3-steps pipeline for finetuning has been implemented with a backend and a frontend for online testing

Our methodology Mistral fine tune and inference Mistral server: Bluff fine tuned model sending data to Poker playing Calling fine mistral servers and start the fine tuning interface tuned model Backend (fast api) Fontend (streamlit) formatte calling raw cleaned d data mistral fine data data tuned model for api to perform converting structure converting the tests selected raw data hands format to Ilm testing format and mistral hands logs to results structure hand format api format downloading data from diverse sources

Finetuning led to MistralBluff, an LLM playing only legit moves, similar to the pros

Results and performance: open range UTG position

Professionals vs MistralBluff moves

						6 3 1	
	bets	calls	raises	allın	checks	folds	caps
bets	16	0	0	0	20	0	0
calls	0	38	5	0	0	59	0
raises	0	4	45	0	0	46	0
allin	0	0	0	2	1	6	0
checks	8	0	1	0	151	0	0
folds	0	19	15	0	0	3017	0
caps	0	0	0	0	0	1	0
nombre	de cou	p légal	: 3454 3	454			

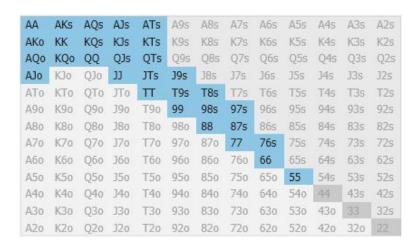
Professionals vs Mistral non fine tuned moves

	bets	calls	raises	allin	checks	folds	caps
bets	14	0	0	0	8	0	0
calls	0	16	1	0	0	0	0
raises	0	3	5	0	0	0	0
allin	0	1	0	0	1	0	0
checks	36	0	0	0	44	0	0
folds	7	87	172	0	48	1	0
caps	0	1	0	0	0	0	0
nombre	de cou	légal:	429 345	54			

- MistralBluff did not play a single forbidden move (3454/3454) against 3025/3459 forbidden for Mistral AI non fine-tuned
- MistralBluff plays as professional for most of the moves not as Mistral AI non fine-tuned

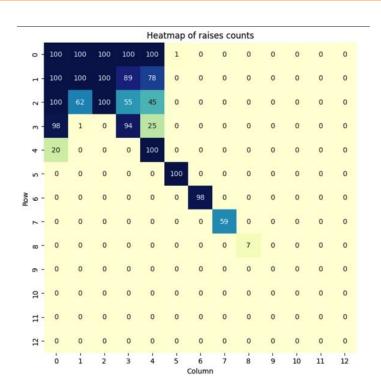
MistralBluff has almost the same opening ranges as professional players

Results and performance: opening range UTG position



Professional player UTG openraise range

- MistralBluff opening range at UTG position is very close to professional poker players
- It understood that suited hands are better than off-suited ones



MistralBluff UTG openraise range

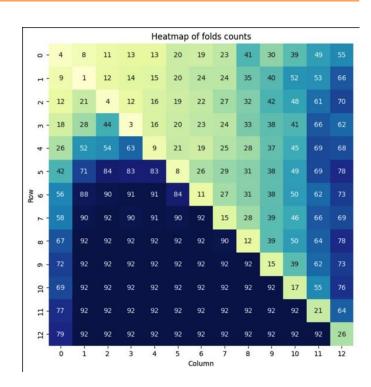
MistralBluff changes its opening range according to its position at the table, as professional players

Results and performance: open range BB position

	Α	K	Q	J	T	9	8	7	6	5	4	3	2	
Α	6.1	3.3	2.7	2.3	1.6	1.2	0.8	0.5	0.2	0.2	0.1	0EV	-0.1	
K	3.0	5.0	0.7	0.3	-0.1	-0.6	-0.9	-0.9	-0.9	-1.0	-1.1	-1.2	-1.2	
Q	2.4	0.2	4.3	-0.1	-0.3	-0.6	-0.8	-1.0	-1.0	-1.1	-1.2	-1.3	-1.3	18
J	2.0	-0.2	-0.6	3.8	-0.2	-0.5	-0.7	-0.9	-1.1	-1.2	-1.2	-1.4	-1.4	3-bet/Call
Т	1.5	-0.6	-0.8	-0.7	3.2	-0.4	-0.6	-0.8	-1.0	-1.3	-1.3	-1.4	-1.5	
9	0.8	-1.1	-1.2	-1.0	-0.9	2.7	-0.6	-0.8	-1.0	-1.2	-1.5	x	х	3-bet shove
8	0.4	-1.4	-1.4	-1.3	-1.2	-1.1	2.3	-0.7	-0.9	-1.1	x	x	x	
7	0.1	-1.4	-1.6	-1.5	-1.4	-1.3	-1.2	1.9	-0.8	-1.0	x	×	x	Call
6	-0.2	-1.5	-1.6	-1.7	-1.7	-1.5	-1.4	-1.3	1.5	-0.9	-1.1	x	х	
5	-0.3	-1.6	-1.7	-1.8	x	x	x	x	x	1.0	-1.0	x	x	Fold
4	-0.4	-1.6	-1.8	-1.8	x	x	x	х	x	х	0.7	х	х	
3	-0.5	-1.7	-1.9	-2.0	x	х	x	x	x	х	х	0.4	х	
2	-0.6	-1.8	-2.0	-2.0	x	x	x	х	x	x	x	х	0EV	

Professional player BB opening range

- **MistralBluff learned to play "loose"** when at Big Blind position
- MistralBluff opening range is consistent with professional players



MistralBluff BB opening range

Despite these promising results, there are still many areas for improvement

Key improvements areas

• **Sizing:** MistralBluff is bad at raise sizing, sizing could be improved in several ways: using bet sizes relative to the pot or blinds, using an agent, using external functions with simple rules.

• **Data:** One of the main difficulties was finding quality data. MistralBluff was trained on data from single professional player and representing 8M tokens. By having more quality hands, MistralBluff could see more game configurations.

• **Evaluation:** It is difficult to evaluate the level of a poker player, as there is no easily calculated ELO score. There are few metrics to evaluate post-flop play and we would have needed more time to test on dozens of online games.