

Predict Player Performance for an Optimal Fantasy Game Lineup

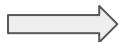
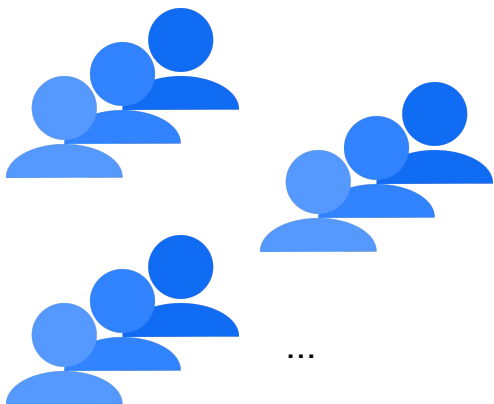
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Introduction



THE GAME INSIDE THE GAME.

Create an 8-player lineup!



Constraints like
Salary cap,
Team, Position

Before match day

player 1:		52	Performance score calculated by a given equation.
player 2:		14	
player 3:		33	
...			

lineup 1:	X8	300
lineup 2:	X8	296
lineup 3:	X8	307
...		...

After match day

Methods

Predict player score.

- Collect historical player statistics
- Data transformation & feature engineering
- Modeling

Form lineup given
constraints.

- Generate Lineup by linear programming
- Diversity Constraints:

Predict player score.

Collect historical player statistics: stats.nba.com

#	PLAYER	GP	MIN	PTS	FGM	FGA	FG%
1	James Harden	78	36.8	36.1	10.8	24.5	44.2
2	Paul George	77	36.9	28.0	9.2	21.0	43.8
3	Giannis Antetokounmpo	72	32.8	27.7	10.0	17.3	57.8
4	Joel Embiid	64	33.7	27.5	9.1	18.7	48.4
5	Stephen Curry	69	33.8	27.3	9.2	19.4	47.2

Data transformation & feature engineering

- Time series features, e.g. “previous game points”, “2nd previous game points” etc.
- Encode categorical feature team: winning rate, average points

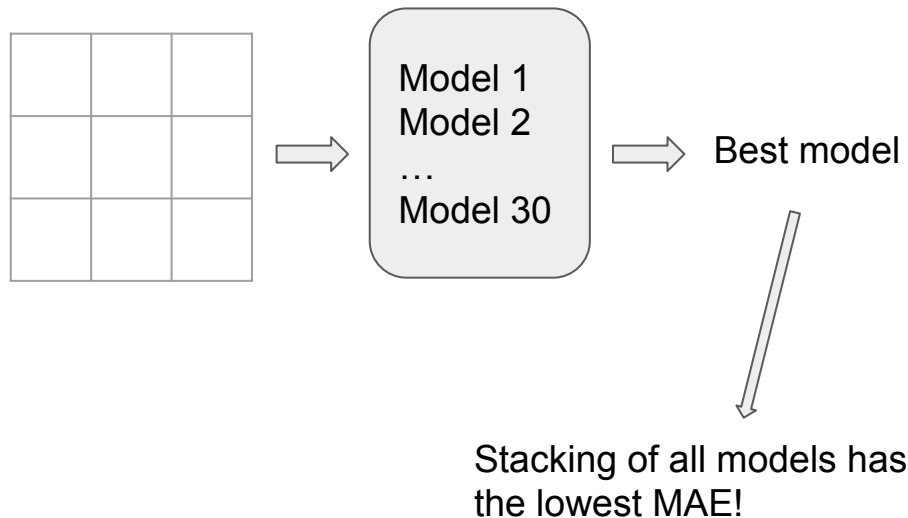
Predict player score.

H2OAutoML(max_models = 30, max_runtime_secs=150, seed = 1)

Modeling: H2O autoML

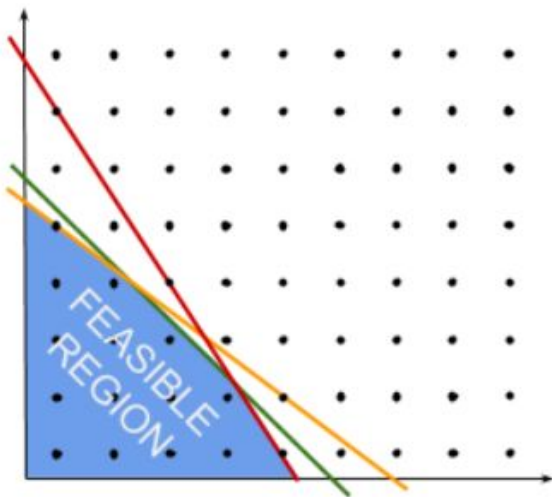
- Automating training & tuning of hyper-parameters
- Often has good performance in practical.
- Returns a leaderboard with the best model.

	model_id	mean_residual_deviance	rmse	mse	mae
	StackedEnsemble_AllModels_AutoML_20191029_185211	97.465	9.87244	97.465	7.79825
	GLM_grid_1_AutoML_20191029_185211_model_1	97.76	9.88736	97.76	7.80035
	StackedEnsemble_BestOfFamily_AutoML_20191029_185211	97.8783	9.89335	97.8783	7.81343
	XGBoost_grid_1_AutoML_20191029_185211_model_2	99.8868	9.99434	99.8868	7.88746
	GBM_5_AutoML_20191029_185211	100.385	10.0192	100.385	7.9257



Form lineup given
constraints.

Generate Lineup by linear programming



Goal:
$$\sum_{p=1}^N f_p x_{pl}$$

Subject To:

$$\sum_{j=1}^p c_j x_{ij} \leq 5000, \quad (\text{budget constraint})$$

$$\sum_{k=1}^p x_{ij} = 8, \quad (\text{lineup size constraint})$$

position constraints:

$$1 \leq \sum_{j \in C} x_{ij} \leq 2$$

$$2 \leq \sum_{j \in (PG, SG)} x_{ij} \leq 3$$

$$2 \leq \sum_{j \in (PF, SF)} x_{ij} \leq 3$$

team constraints:

$$t_{il} \leq \sum_{k \in T_l} x_{ik}, \quad l = 1, \dots, N_T$$

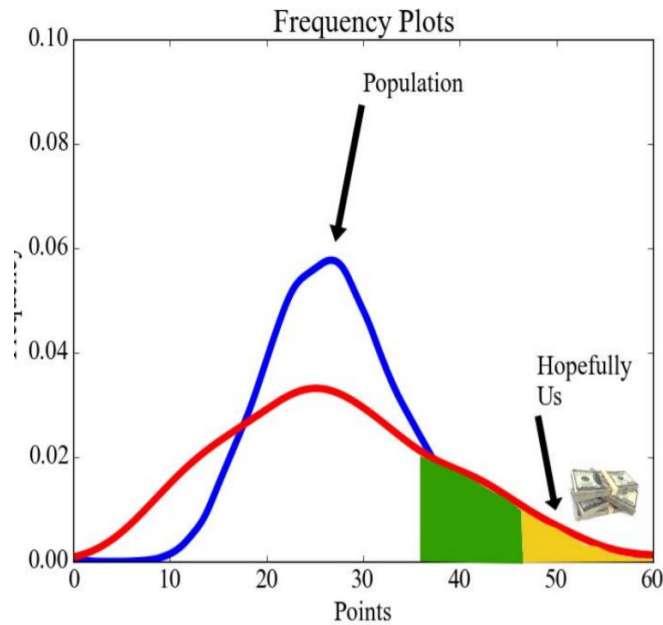
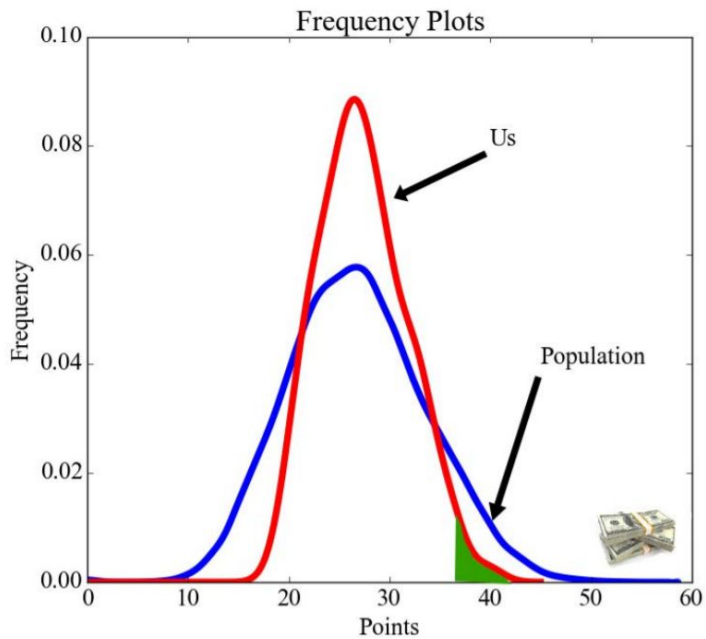
$$\sum_{l=1}^{N_T} t_{il} \geq 3$$

$$t_{il} \in \{0, 1\}, \quad l = 1, \dots, N_T$$

Form lineup given
constraints.

Diversity Constraints:
Make sure lineup i has only a few overlap players from
lineup 1 to i-1

$$\sum_{p=1}^N x_{pk}^* x_{pl} \leq \gamma, k = 1, \dots, l - 1$$



Evaluation



Sell lineup strategy (\$30 monthly subs fee 😞 expensive!)

draftkings_NBA_2019-02-01_players.csv
draftkings_NBA_2019-02-02_players.csv
draftkings_NBA_2019-02-03_players.csv
draftkings_NBA_2019-02-04_players.csv
draftkings_NBA_2019-02-05_players.csv
draftkings_NBA_2019-02-06_players.csv
draftkings_NBA_2019-02-07_players.csv
draftkings_NBA_2019-02-08_players.csv
draftkings_NBA_2019-02-09_players.csv
draftkings_NBA_2019-02-10_players.csv
draftkings_NBA_2019-02-11_players.csv

VS. Our prediction \Longrightarrow Avg. MAE:

FC's	Ours
7.80	7.67

Feb. 1 - Mar. 31, 2019

Evaluation

- Simulation on Feb.01
- 4 of our 49 lineups won

Conclusion

- A capable machine learning solution for selecting lineups
- Acceptable performance as a prototype
- Need to improve ROI in the future