Prediction of Tennis Matches

IEOR 242 - UC Berkeley

Introduction









US Open Tennis 2019: Updated Men's Bracket and Final Predictions

RORY MARSDEN

2020 Australian Open women's odds, picks, predictions: Tennis expert says Serena Williams poised for upset

Gavin Mair is up over \$22,000 on his women's tennis tournament picks over the past three years.

CBS Sports Staff Jan 19, 2020 at 11:27 am ET • 2 min read

Machine learning predicts World Cup winner

Researchers have predicted the outcome after simulating the entire soccer tournament 100,000 times.

by Emerging Technology from the arXiv

A MACHINE LEARNING ANALYSIS OF THE NFL: PREDICTING NEW PLAYOFF CONTENDERS

🗎 SEPTEMBER 5, 2018 🖀 BY HARVARDSPORTS 🗿 4 MIN READ 😡 1 COMMENT

By Matty Cheng

US OPEN (TENNIS)





US Open Tennis 2019: Bracket Predictions, Odds for Top Players, **Prize-Money Info**

GILL CLARK >

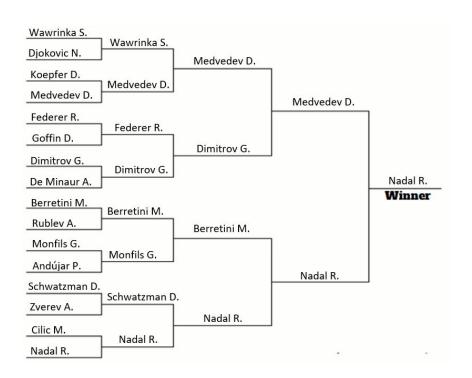


World Cup: Al predicts that Germany will win

June 12, 2018

Our Goal: Predict 2019 US Open Winner





Approach

Data Mining

Dataset from tennis-data.co.uk/ - Data manipulation with Python

R Models

Baseline - Logistic Regression - CART - Random Forest

Python Models

Principal Component Analysis - Neural Networks

Final Model

Model with highest accuracy selected - Simulation

Data Mining

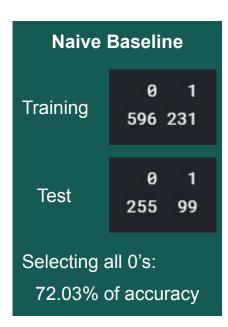
1. Raw Data Pulled From tennis-data.co.uk/

Tournament	Date	Surface	Round	Best of	Winner	Loser	WRank	LRank		W1	L1	W2	L2	W3	L3	
US Open	8/29/05	Hard	1st Round		5 Massu N.	Gambill J.M.	58		151		7	6	6	2	6	3
US Open	8/29/05	Hard	1st Round		5 Baker B.	Gaudio G.	195		9		7	6	6	2	6	4
US Open	8/29/05	Hard	1st Round		5 Ljubicic I.	Summerer T	19	-	198		6	3	7	6	6	3
US Open	8/29/05	Hard	1st Round		5 Malisse X.	Hernych J.	54		81		6	2	7	5	6	2
US Open	8/29/05	Hard	1st Round		5 Berdych T.	Kohlschreibe	34		79		7	6	7	6	6	4
US Open	8/29/05	Hard	1st Round		5 Nadal R.	Reynolds B.	2		132		6	3	6	3	6	4
US Open	8/29/05	Hard	1st Round		5 Novak J.	Norman D.	25		112		6	4	6	4	6	2

2. Feature Generation Performed (21 Variables) to Predict Output Variable (Win/Loss)

```
'diff_rank', 'diff_match_win_percent', 'diff_games_win_percent', 'diff_5_set_match_win_percent', 'diff_close_sets_percent', 'diff_match_win_percent_hard', 'diff_games_win_percent_hard', 'diff_5_set_match_win_percent_52' 'diff_games_win_percent_52', 'diff_5_set_match_win_percent_52', 'diff_close_sets_percent_52', 'diff_match_win_percent_hard_60', 'diff_games_win_percent_hard_60', 'diff_5_set_match_win_percent_hard_60', 'diff_close_sets_percent_hard_60', 'diff_games_win_percent_hard_60', 'diff_games_win_percent_hard_hh', 'diff_games_win_percent_hard_hh', 'diff_games_win_percent_hard_hh', 'diff_games_win_percent_hard_hh']
```

R Models



Logistic Regression

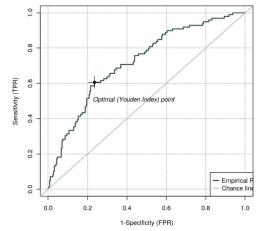
Features Used:

Percent Difference in Matches Won
Percent Difference Between "Close" Sets
Percent Difference in Matches Won in 52 wks

TRUE	FALSE	
109	146	0
70	29	1

Accuracy: 66.3%

FPR: 32.3% TPR: 56.5%



AUC: .691

R Models

CART

diff_q;_nes_win_percent_5 > 10.0391

0 diff_match_win_percent >= -0.107

diff_games_win_percent_h < 0.0159

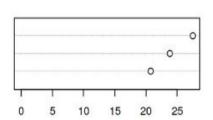
0 diff_match_win_percent_h >= 0.0629

Accuracy in the test set: 70%

Random Forest

Variable Importance:

diff_games_win_percent_52 diff_match_win_percent diff_match_win_percent_52



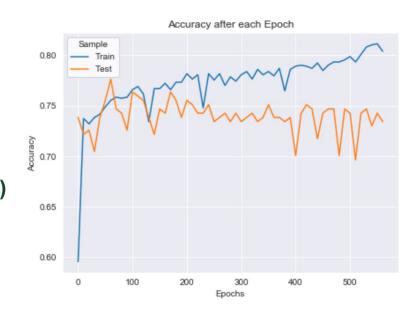
modelRF

Accuracy in the test set: 74%



Neural Network

- Built Using Keras Library in Python
 - Layer 1: 64 Units (Relu Activation)
 - Layer 2: 32 Units (Relu Activation)
 - Layer 3: 1 Unit (Sigmoid Optimizer)
- Train Accuracy: 0.758
- Test Accuracy: 0.776

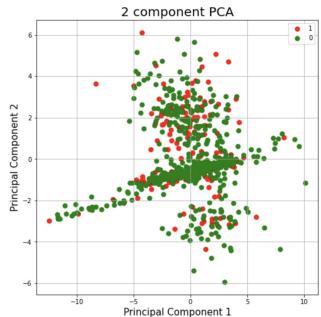


Python Models

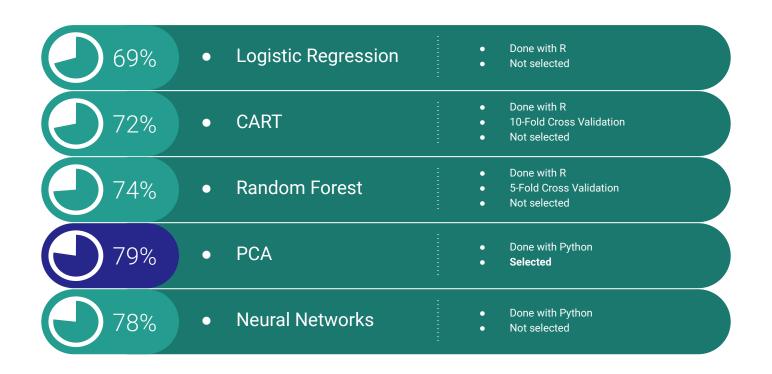
Principal Component Analysis

- Standardizing Features
- We keep 2 principal components
- Test Accuracy: 0.79

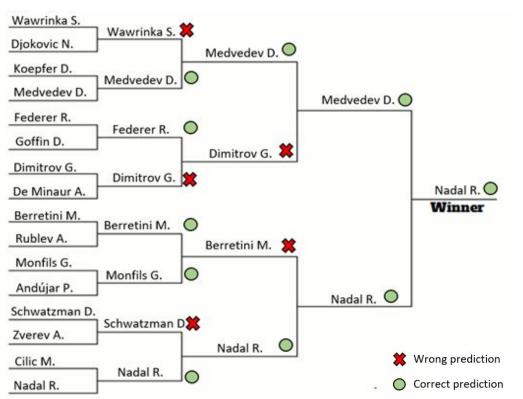
	Explained Variance percentage
First PC	64 %
Second PC	23 %



Comparison & Selection



Results (2019 US Open Model Prediction)



Next step:

US Open 2020

Thank You

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