

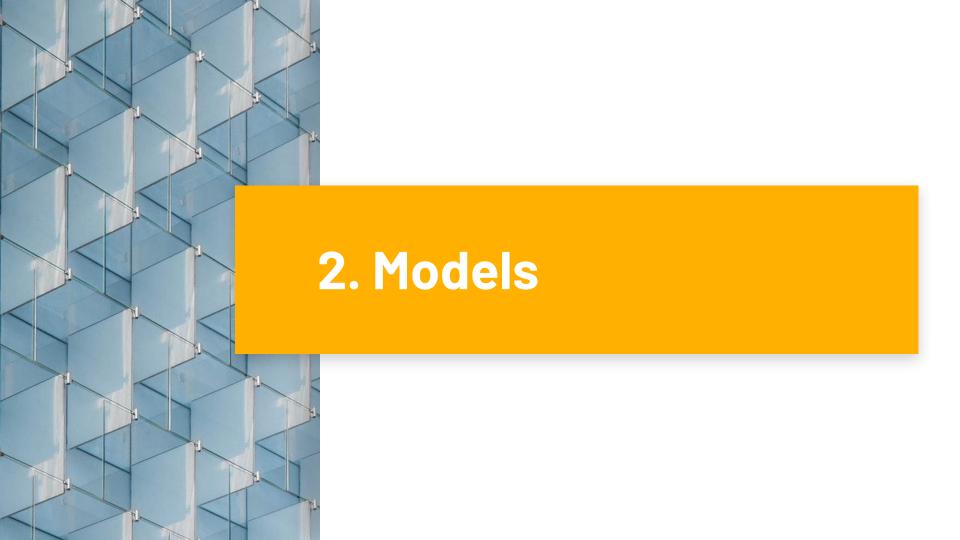
What We've Learned

- Under-confidence VS over-confidence
- Share information and perspectives helps make dragonfly eye work
- Keep bias away



Super Bowl LIII Prediction Dataset

Time	2013-2018				
Independent variables	Win or Lose				
	Points scored				
22 Dependent variables	Yards gained by passing/rushing(offense & defense)				
	Team turnovers lost(offense) / turnovers gained(defense)				
	Expected points				
	Weather condition, stadium type				
	Las Vegas betting odds				





To Predict Who Will Win:

- Logistic Regression
- Decision Tree



To Predict the Exact Score:

- Linear Regression
- Poisson Regression
- SARIMA



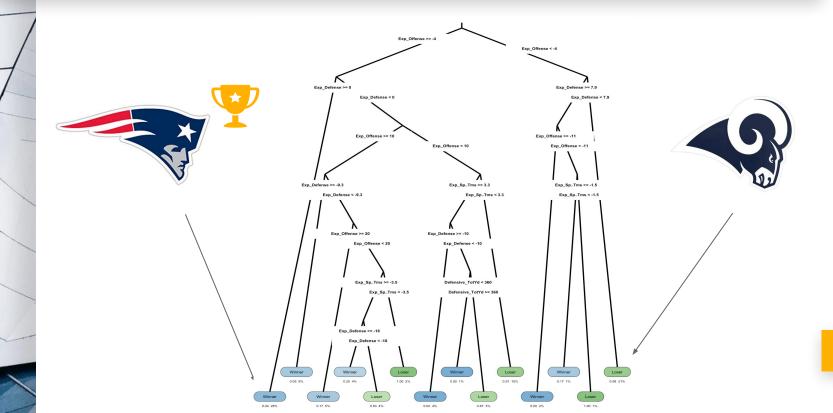
Logistic Regression

```
Coefficients:
                Estimate Std. Error z value
                                                Pr(>|z|)
(Intercept)
               -5.483943
                          3.111608 -1.762
                                                0.07800 .
Offensive_1stD -0.027344
                          0.082027 -0.333
                                                0.73886
Offensive_PassY 0.002299
                          0.005574
                                     0.412
                                                0.67998
Offensive_RushY 0.003231
                          0.007571
                                     0.427
                                                0.66959
                                                0.00213 **
               -1.090652
                          0.355081 -3.072
Offensive TO
                                                             > predict(fit5, newdata = NFL_rams, type = "response")
                          0.097577 -0.144
                                                0.88585
Defensive 1stD -0.014008
Defensive PassY 0.020888
                          0.007038
                                    2.968
                                                0.00300 **
Defensive_RushY 0.014649
                          0.008345
                                                0.07919
                                    1.755
                                                             0.9473962
Defensive_TO
                0.704976
                          0.296384
                                    2.379
                                                0.01738 *
                                                              > predict(fit5, newdata = NFL_patriots, type = "response")
Exp_Offense
                0.303312
                          0.061266
                                     4.951 0.00000073928 ***
Exp_Defense
                0.480864
                          0.083725
                                    5.743 0.00000000928 ***
                                                             0.9156235
                          0.518060
                                                0.48729
stadium type
               -0.359854
                                    -0.695
Favorite_or_not 1.130869
                          0.620120
                                    1.824
                                                0.06821 .
Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 422.43 on 315 degrees of freedom
Residual deviance: 114.68 on 303 degrees of freedom
  (397 observations deleted due to missingness)
```

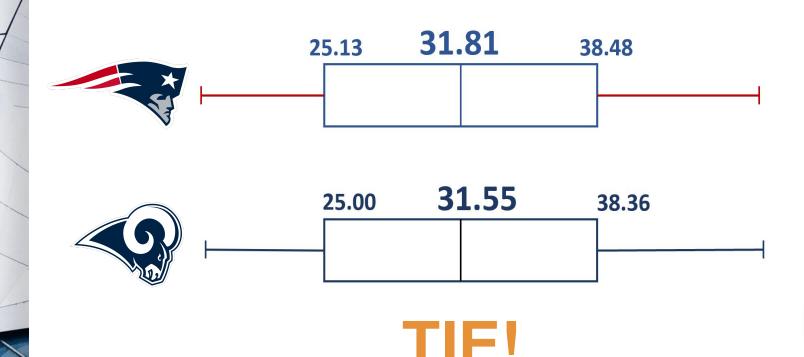
Number of Fisher Scoring iterations: 8

AIC: 140.68

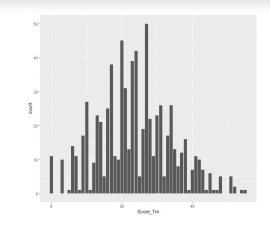
Decision Tree



Linear Regression * Vegas Spread Betting



Poisson Regression



```
Deviance Residuals:

Min 1Q Median 3Q Max
-4.4515 -0.8754 -0.0321 0.9112 4.0932
```

Coefficients:

```
Estimate Std. Error z value Pr(>|z|)
(Intercept)
              1.8224452 0.1187001 15.353 < 2e-16 ***
Offensive_PassY 0.0021406 0.0003197
                                  6.695 2.16e-11 ***
Offensive_RushY 0.0025864 0.0004032 6.414 1.42e-10 ***
Offensive TO
             -0.1064921 0.0192804 -5.523 3.33e-08 ***
Defensive_1stD 0.0101540 0.0061990
                                 1.638 0.10142
Defensive_PassY 0.0002940 0.0003463
                                 0.849 0.39587
Defensive_Rushy -0.0010887  0.0004623  -2.355  0.01854 *
Defensive_TO
              0.0682428 0.0155828
                                 4.379 1.19e-05 ***
stadium_type
              0.1225009 0.0388557 3.153 0.00162 **
              -0.0165678 0.0031499 -5.260 1.44e-07 ***
Spread
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

(Dispersion parameter for poisson family taken to be 1)



26 vs 32





Seasonal | Autoregressive | Integrated | Moving Average

[Period] [Prior Values] [Differencing] [Regression Error]



Super Bowl LIII Prediction Dataset

Team	Year -		Internal			External	
		Score	Offensive	Defensive	Temp	Weather	Bet
Patriots	2013 - 2018						
Rams	2013 - 2018						
Other Teams	2018						



• SARIMA (*p*, *d*, *q*, *P*, *D*, *Q*)

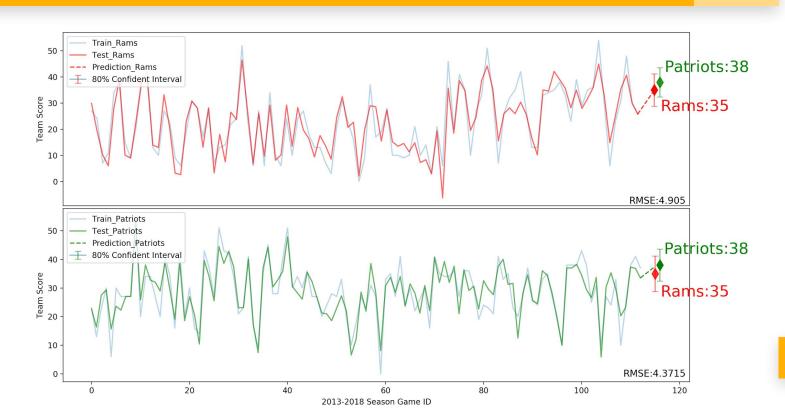
- p: the order of non-seasonal AR term
- d: the order of non-seasonal differencing
- g: the order of non-seasonal MA term
- P: the order of seasonal AR term
- D: the order of seasonal differencing
- \circ Q: the order of seasonal MA term



from pmdarima.arima import auto_arima

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Workflow

Dataset

Model

Predict

Score

- 2013-2018
 Match Data
- Weather
- Las VegasBetting Line
- 717 x 22

- LogisticRegression
- Decision Tree
- Linear Regression with Betting
- PoissonRegression
- SARIMA

- Patriots
- Rams
- Input = Avg. data of previous two games: Division
 - & Conf. Champ

- Weight Avg.
- 20% Linear
- 20% Poisson
- 60%-SARIMA

Summary

Logistic Regression	Decision Tree	Linear Regression with Betting	Poisson Regression	Seasonal ARIMA	
8 Sig. Var.	First node: Exp.Offensive	8 Sig. Var.	9 Sig. Var.	P(0,1,1,0,1,1) R(0,1,1,1,1,2)	
AUC = 0.979	ACC = 0.933	Adj. R square = 0.7895	Residual Deviance = 319.1 (d.f.=149)	Avg(RMSE) = 4.639	
		Weight: 20%	Weight: 20%	Weight: 60%	
Rams	Patriots	P(31): R(31)	P(26): R(32)	P(38) : R(35)	

P = Patriots, R = Rams













Reference

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[7]. Detailed NFL Play-by-Play Data 2009-2018,
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