

# Project Luther: Scraping and Modeling

Major League Soccer Data

#### **Overview**

■ Introduction

- Scraping the Data
- Modeling the Data
- Challenges
- Questions

#### Introduction

#### **Objective:**

To explain how to use the python data stack to scrape and model Major League Soccer (MLS) goalkeeper statistics in order to predict salary.

#### **Tools Used:**







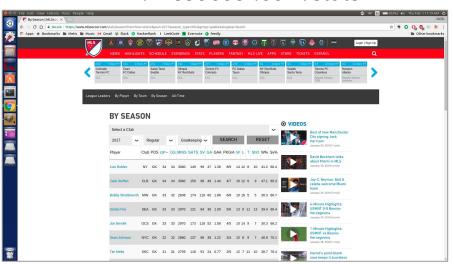




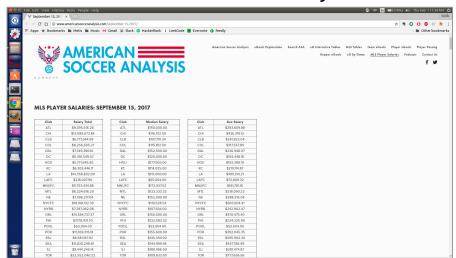
# **Scraping the Data**

#### **Data Sources**

www.mlssoccer.com/stats

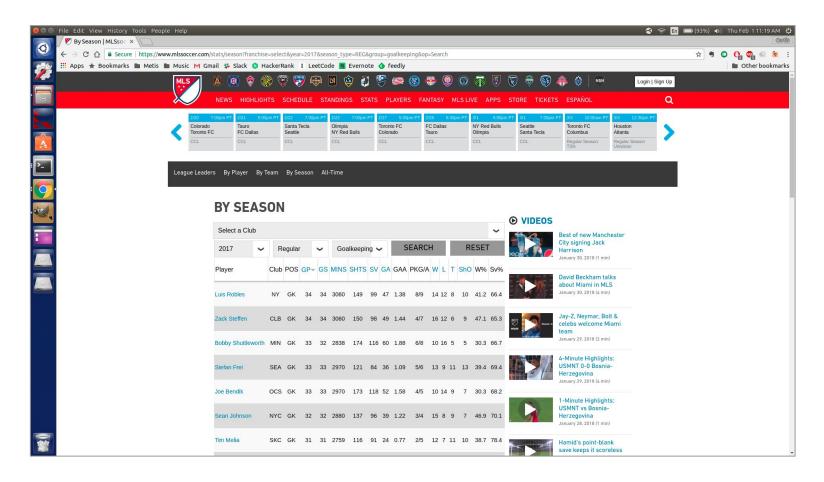


www.americansocceranalysis.com

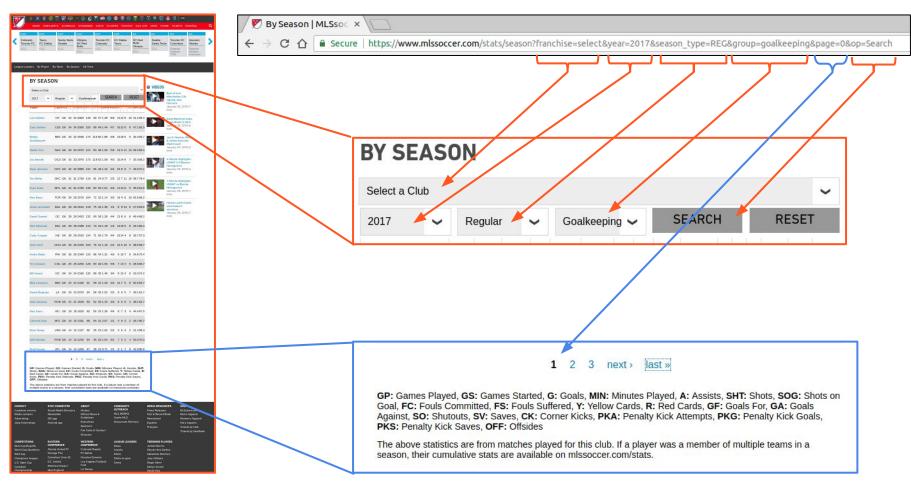


Player statistics and salary information for the years 2007 through 2017.

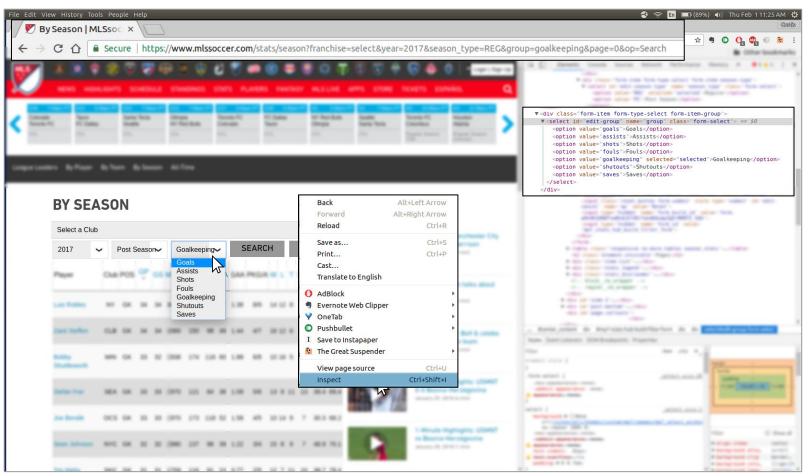
#### What does mlssoccer.com look like?



#### **Identifying URL Structure**

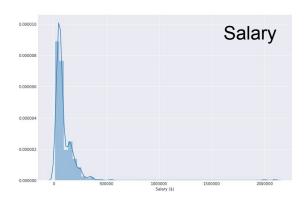


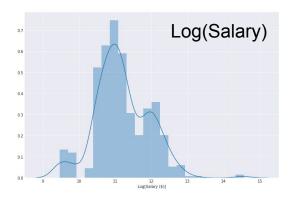
### **Locating Parameter Values**

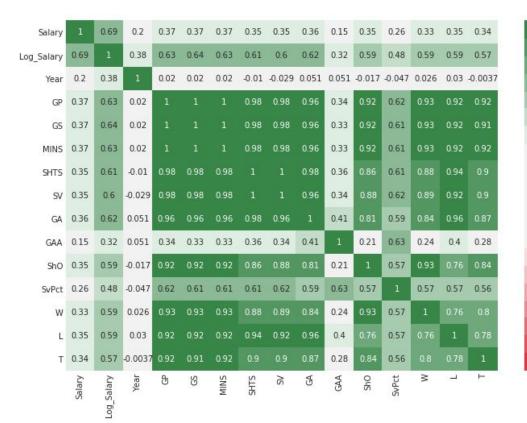


## **Modeling the Data**

#### Transforming & Exploring







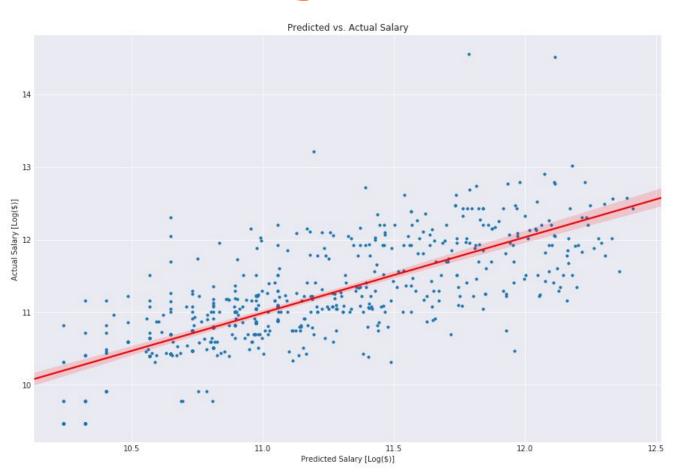
0.4

0.0

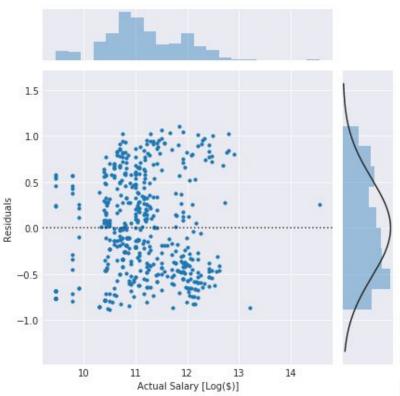
-0.4

-0.8

## Fitting a Model



#### **Evaluating the Fit**



	coef	std err	t	P> t	[0.025	0.975]		
Year	0.0847	0.008	10.414	0.000	0.069	0.101		
GP	-0.0055	0.084	-0.066	0.948	-0.170	0.159		
GS	0.4279	0.100	4.295	0.000	0.232	0.624		
MINS	-0.0055	0.002	-2.649	0.008	-0.010	-0.001		
SHTS	0.0108	0.017	0.655	0.513	-0.022	0.043 0.019		
SV	-0.0143	0.017	-0.831	0.406	-0.048			
GA	-0.0168	0.019	-0.862	0.389	-0.055	0.021		
GAA	0.0800	0.043	1.852	0.065	-0.005	0.165		
ShO	0.0268	0.025	1.056	0.292	-0.023	0.077		
SvPct	0.0025	0.001	1.917	0.056	-6.22e-05	0.005		
W	0.1032	0.111	0.931	0.352	-0.115	0.321		
L	0.1322	0.111	1.192	192 0.234 -0.086		0.350		
Т	0.1012	0.112	0.908	0.365	-0.118	0.320		
Intercept	-159.8163	16.372	-9.761	0.000	-192.005	-127.628		

Omnibus: 36.689

Skew:

Kurtosis:

0.000

4.776

Prob(Omnibus):

**Durbin-Watson:** 

Jarque-Bera (JB):

1.923

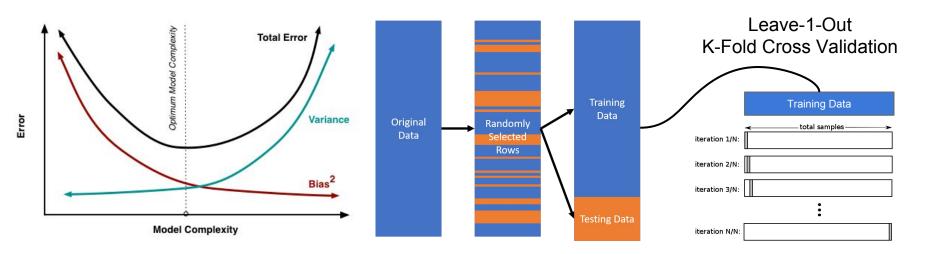
71.917

Prob(JB): 2.42e-16

Cond. No. 1.58e+06

Dep. Variable:	Log_Salary	R-squared:	0.590
Model:	OLS	Adj. R-squared:	0.576
Method:	Least Squares	F-statistic:	43.23
Date:	Thu, 01 Feb 2018	Prob (F-statistic):	1.94e-67
Time:	21:13:45	Log-Likelihood:	-268.82
No. Observations:	405	AIC:	565.6
Df Residuals:	391	BIC:	621.7
Df Model:	13		
Covariance Type:	nonrobust		

### **Generalizing the Fit**



Regularized Model Scores on Test Data									
RidgeCV	LassoCV	ElasticNetCV							
alpha = 0.1	alpha ~ 5.9e <sup>-4</sup>	alpha ~ 3.7e <sup>-4</sup> (L1 = 0.975)							
R <sup>2</sup> ~ 0.538, R <sup>2</sup> -adj ~ 0.497	R <sup>2</sup> ~ 0.544, R <sup>2</sup> -adj ~ 0.505	$R^2 \sim 0.546$ , $R^2$ -adj $\sim 0.506$							

### Challenges & Improvements

#### Issues:

- Salary is often fixed for several years at a time due to contract structure.
- Extreme outliers skew the distribution of salaries positively and the features we have cannot explain this variance.
- Many players bring intangible value to their teams that isn't captured in their stats.
- Data integrity issues (errors in the MLS database)

#### **Possible Solutions:**

- Add additional features to the dataset:
  - Contract signing year and any performance incentives.
  - Complete player history, not just their time in the MLS (i.e., international experience, time in other leagues)
- Data validation with other sources.

# Questions?

# Backup

#### **Querying Example**

```
'<!DOCTYPE html>\n<!--[if IEMobile 7]><html class="iem7" lang="en" dir="ltr"><![endif]-->\n<!--[if lte IE 6]><ht</pre>
ml class="lt-ie9 lt-ie8 lt-ie7" lang="en" dir="ltr"><![endif]-->\n<!--[if (IE 7)&(!IEMobile)]><html class="lt-ie8 lt-ie8 lt-ie
9 lt-ie8" lang="en" dir="ltr"><![endif]-->\n<!--[if IE 8]><html class="lt-ie9" lang="en" dir="ltr"><![endif]-->
\n<!--[if IE 9]><html class="eq-ie9" lang="en" dir="ltr"><![endif]-->\n<!--[if (gte IE 9)|(gt IEMobile 7)]><!-->
<html class="no-ie" lang="en" dir="ltr"\n xmlns:og="http://ogp.me/ns#"\n xmlns:article="http://ogp.me/ns/artic</pre>
le#"\n xmlns:book="http://ogp.me/ns/book#"\n xmlns:profile="http://ogp.me/ns/profile#"\n xmlns:video="http://o
gp.me/ns/video#"><!--<![endif]-->\n\n<head profile="http://www.w3.org/1999/xhtml/vocab">\n\n <meta charset="utf-
8" />\n<link rel="apple-touch-icon" sizes="144x144" href="/sites/league/themes/league/img/apple-touch-icon-144x14
4-precomposed.png" />\n<link rel="apple-touch-icon" sizes="152x152" href="/sites/league/themes/league/img/apple-t
ouch-icon-152x152-precomposed.png" />\n<link rel="apple-touch-icon" sizes="120x120" href="/sites/league/themes/le
aque/img/apple-touch-icon-120x120-precomposed.png" />\nlink rel="shortcut icon" href="https://leaque-mp7static.m
lsdigital.net/favicon.ico?MbTlG7NMwJYaZIfo0mOlfyfDvF9eMba2" type="image/vnd.microsoft.icon" />\n<link rel="apple-
touch-icon" sizes="72x72" href="/sites/leaque/themes/leaque/img/apple-touch-icon-72x72-precomposed.png" />\nlink
rel="apple-touch-icon" sizes="60x60" href="/sites/league/themes/league/img/apple-touch-icon-60x60-precomposed.pn
g" />\nlink rel="apple-touch-icon" sizes="57x57" href="/sites/league/themes/league/img/apple-touch-icon.png" />
 \nnk rel="apple-touch-icon" sizes="76x76" href="/sites/league/themes/league/img/apple-touch-icon-76x76-precomp
osed.png" /\nlink rel="apple-touch-icon" sizes="114x114" href="/sites/league/themes/league/img/apple-touch-icon" sizes="114x114" href="/sites/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/themes/league/them
-114x114-precomposed.png" />\n<meta name="apple-itunes-app" content="app-id=397303467" />\n<script src="//cdns.gi gya.com/js/gigyaGAIntegration.js"></script \n<script \src="//cdns.gigya.com/js/socialize.js?apiKey=3 qXcJkloa6NFF9
zexvt85l9soAHM8lMBWhxcXyhpo3egangPp8bQONNH8vunw-rTE&lang=en">{lang: "en"}</script>\n<script>var sf startpt=
```

### **Parsing and Compiling**

```
1 # Parse HTML response
 2 soup = BeautifulSoup(response.text,'lxml')
 3 stats table = soup.find('table')
 5 # Extract salary data
 6 stat header = []
 7 stat data = []
 8 for row in stat table.findAll('tr'):
       row data = []
10
       # Get row type and check if header or data row
12
       row type = row.findChild().name
13
       if row type == 'th':
14
           # Extract header
15
           for h in stat table.findAll('th'):
16
               stat header.append(h.text)
17
       else:
18
           # Extract data
19
           for data in row.findAll('td'):
20
               row data.append(data.text)
21
           stat data.append(row data)
22
23 # Compile stat data into dataframe
24 stat df = pd.DataFrame(stat data, columns=stat header)
```

Player	Club	POS	GP	GS	MINS	SHTS	SV	GA	GAA	W	L	Т	ShO	Wpct	SvPct	Year	Season	PKG	PKA
Bouna Coundoul	COL	GK	30.0	30.0	2668.0	158.0	120.0	32.0	1.07	9.0	12.0	8.0	9.0	30.0	75.9	2007	REG	2.0	2.
Kevin Hartman	KC	GK	30.0	30.0	2700.0	159.0	110.0	45.0	1.50	11.0	12.0	7.0	5.0	36.7	69.2	2007	REG	6.0	7.
Matt Reis	NE	GK	30.0	30.0	2700.0	169.0	120.0	43.0	1.43	14.0	8.0	8.0	10.0	46.7	71.0	2007	REG	3.0	3.
Joe Cannon	LA	GK	29.0	29.0	2610.0	171.0	119.0	46.0	1.59	9.0	13.0	7.0	5.0	31.0	69.6	2007	REG	4.0	5.
Troy Perkins	DC	GK	29.0	29.0	2610.0	155.0	117.0	32.0	1.10	16.0	6.0	7.0	8.0	55.2	75.5	2007	REG	1.0	3.
Kendall McIntosh	POR	GK	0.0	0.0	0.0	0.0	0.0	0.0	0.00	0.0	0.0	0.0	0.0	0.0	0.0	2017	REG	0.0	0.
Josh Saunders	ORL	GK	0.0	0.0	0.0	0.0	0.0	0.0	0.00	0.0	0.0	0.0	0.0	0.0	0.0	2017	REG	0.0	0.
Eric Kronberg	MTL	GK	0.0	0.0	0.0	0.0	0.0	0.0	0.00	0.0	0.0	0.0	0.0	0.0	0.0	2017	REG	0.0	0.
Bryan Meredith	SEA	GK	0.0	0.0	0.0	0.0	0.0	0.0	0.00	0.0	0.0	0.0	0.0	0.0	0.0	2017	REG	0.0	0.
Ryan Meara	NY	GK	0.0	0.0	0.0	0.0	0.0	0.0	0.00	0.0	0.0	0.0	0.0	0.0	0.0	2017	REG	0.0	0.

## **Linking Datasets**