

Raindrops  
*don't*  
keep falling on  
my head

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# Agenda

1. Background
2. Data Source
3. Tools
4. Findings
5. Conclusion

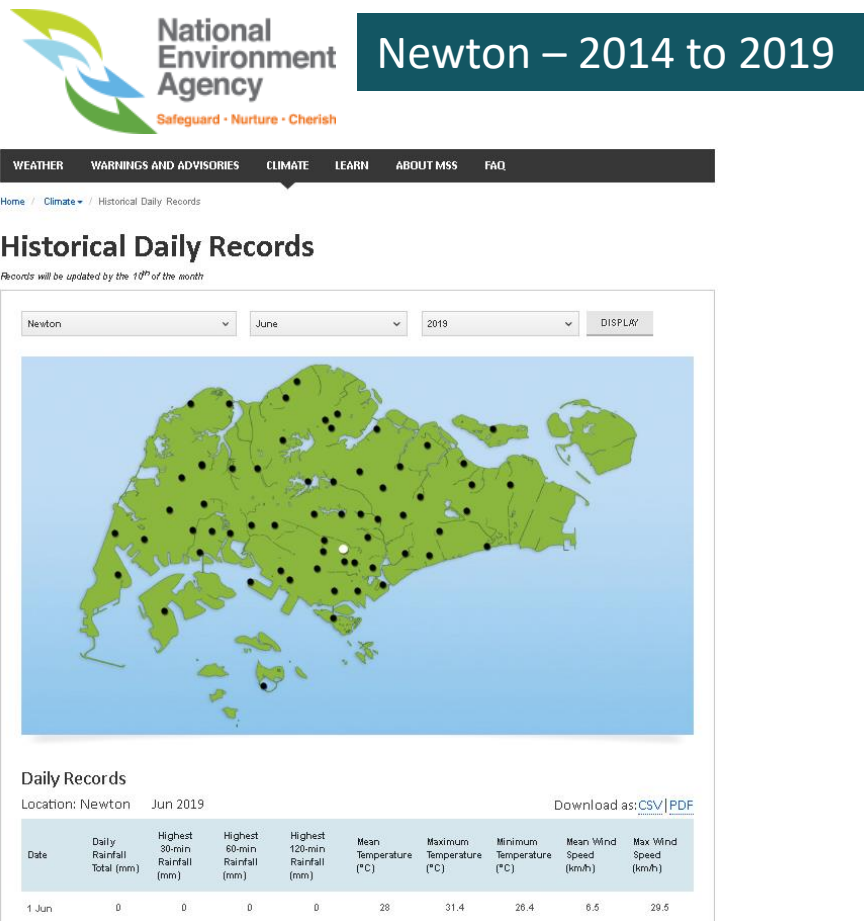
# Background

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When should I bring an umbrella to fetch my kid at Newton?

How much transport allowance should I budget for grab rides for my kid?

# Data Source



# Tools

1. Web scrapping
  - BeautifulSoup4
  - Selenium
2. EDA and Data Cleaning
  - Numpy
  - Pandas
3. Model Building and Selection
  - Scikit
  - Scipy.stats
  - Statsmodel
  - Seaborn and Matplotlib

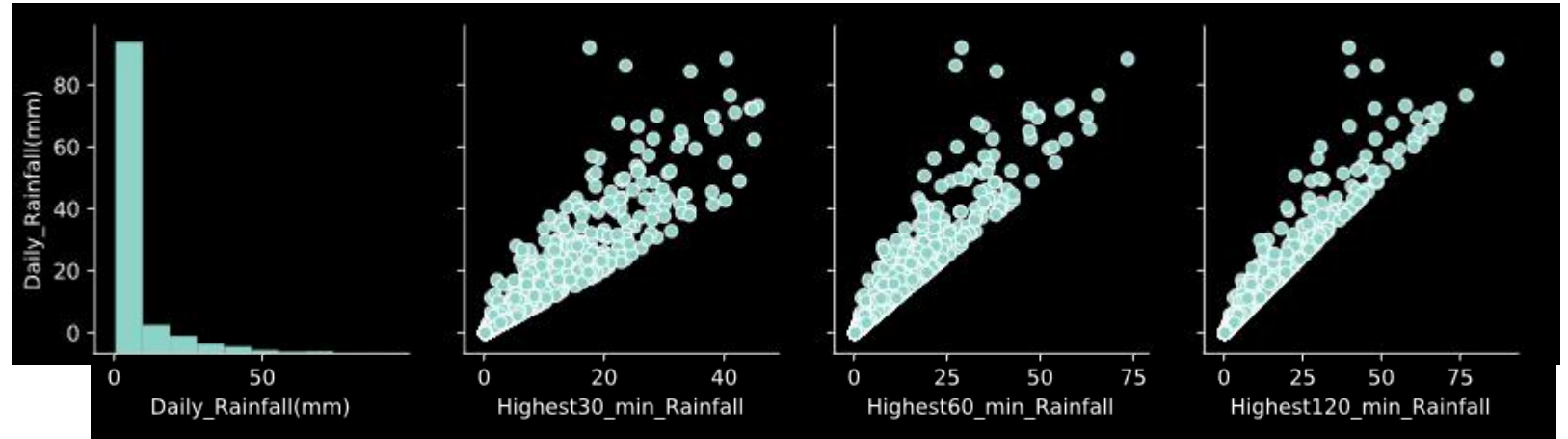


From 3600 data rows -> 1800 rows

# Findings – Remove “noisy” parameters

## PARAMETERS

1. Date
2. Daily Rainfall Total (mm)
3. Highest 30-min Rainfall (mm)
4. Highest 60-min Rainfall (mm)
5. Highest 120-min Rainfall (mm)
6. Mean Temperature (°C)
7. Maximum Temperature (°C)
8. Minimum Temperature (°C)
9. Mean Wind Speed (km/h)
10. Max Wind Speed (km/h)

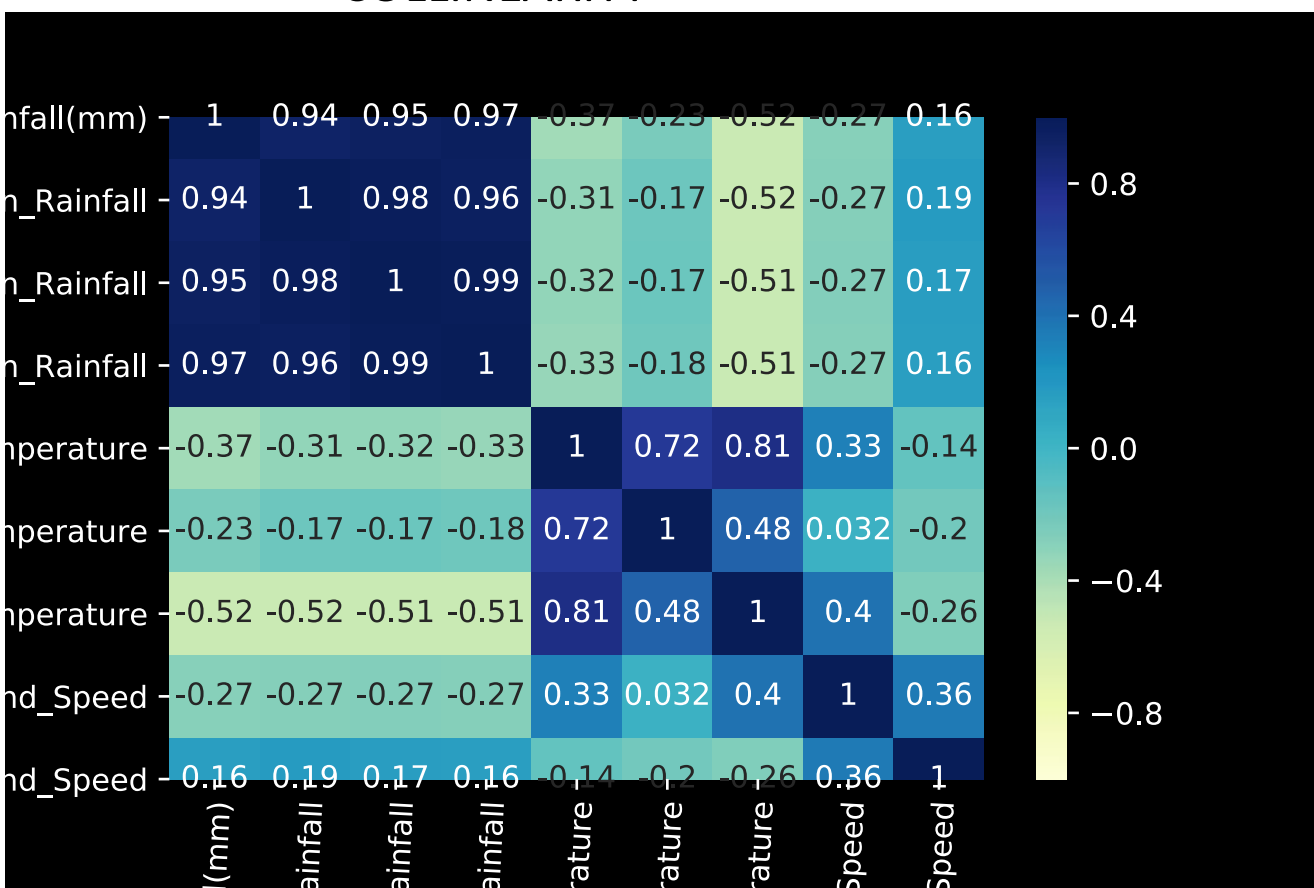


# Findings

## PARAMETERS

1. Date
2. Daily Rainfall Total (mm)
3. Highest 30-min Rainfall (mm)
4. Highest 60-min Rainfall (mm)
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9. Mean Wind Speed (km/h)
10. Max Wind Speed (km/h)

## COLLINEARITY



# Findings – Model 1

## PARAMETERS

1. Date
2. Daily Rainfall Total (mm)
3. Highest 30-min Rainfall (mm)
4. Highest 60-min Rainfall (mm)
5. Highest 120-min Rainfall (mm)
6. Mean Temperature (°C)
7. Maximum Temperature (°C)
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9. Mean Wind Speed (km/h)
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### OLS Regression Results

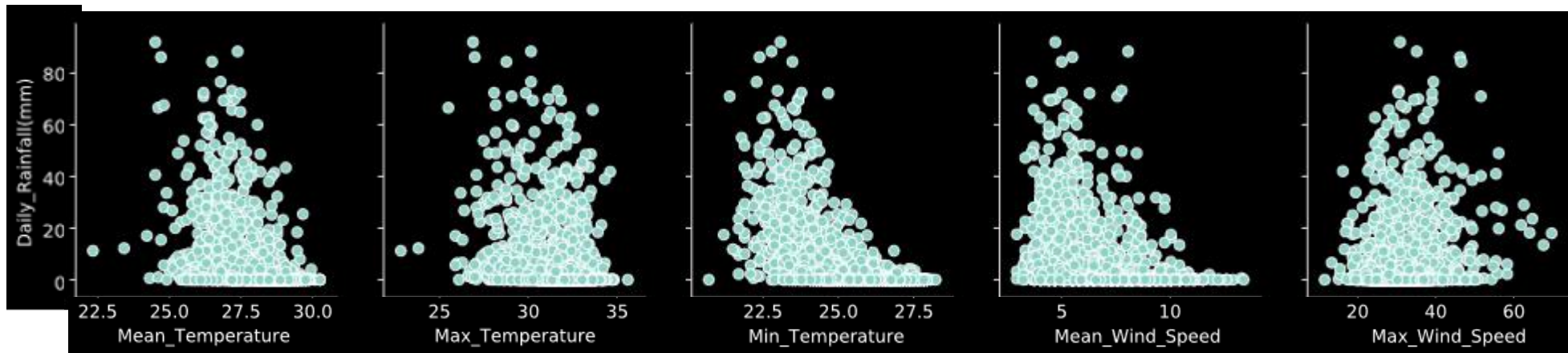
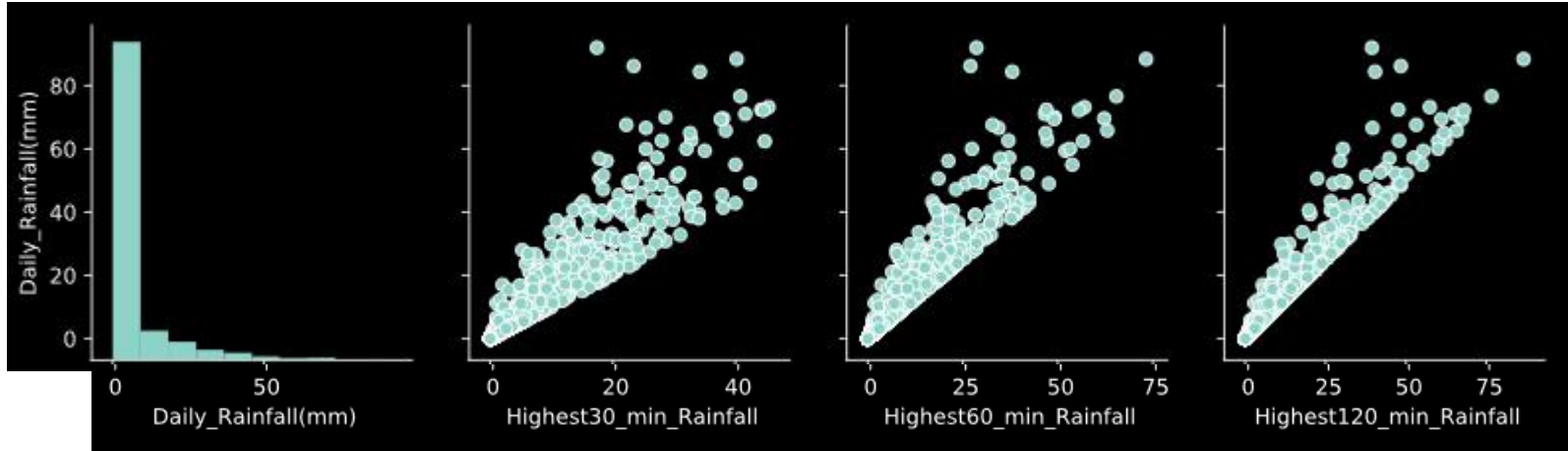
<b>Dep. Variable:</b>	Daily_Rainfall(mm)	<b>R-squared (uncentered):</b>	0.305
<b>Model:</b>	OLS	<b>Adj. R-squared (uncentered):</b>	0.304
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	265.1
<b>Date:</b>	Thu, 25 Jul 2019	<b>Prob (F-statistic):</b>	1.31e-142
<b>Time:</b>	17:09:32	<b>Log-Likelihood:</b>	-7005.6
<b>No. Observations:</b>	1814	<b>AIC:</b>	1.402e+04
<b>Df Residuals:</b>	1811	<b>BIC:</b>	1.403e+04
<b>Df Model:</b>	3		
<b>Covariance Type:</b>	nonrobust		

	<b>coef</b>	<b>std err</b>	<b>t</b>	<b>P&gt; t </b>	<b>[0.025</b>	<b>0.975]</b>
<b>Min_Temperature</b>	-0.0505	0.056	-0.90	0.368	-0.161	0.060
<b>Mean_Wind_Speed</b>	-2.2729	0.162	-14.03	0.000	-2.591	-1.955
<b>Max_Wind_Speed</b>	0.6659	0.038	17.46	0.000	0.591	0.741

<b>Omnibus:</b>	1136.439	<b>Durbin-Watson:</b>	1.816
<b>Prob(Omnibus):</b>	0.000	<b>Jarque-Bera (JB):</b>	11480.514
<b>Skew:</b>	2.867	<b>Prob(JB):</b>	0.00
<b>Kurtosis:</b>	13.909	<b>Cond. No.</b>	26.6



# Findings – Collinearity





# Findings – Model 2

## PARAMETERS

1. Date
2. Daily Rainfall Total (mm)
3. Highest 30-min Rainfall (mm)
4. Highest 60-min Rainfall (mm)
5. Highest 120-min Rainfall (mm)
6. Mean Temperature (°C)
7. Maximum Temperature (°C)
8. Minimum Temperature (°C)
9. Mean Wind Speed (km/h)
10. Max Wind Speed (km/h)

## OLS Regression Results

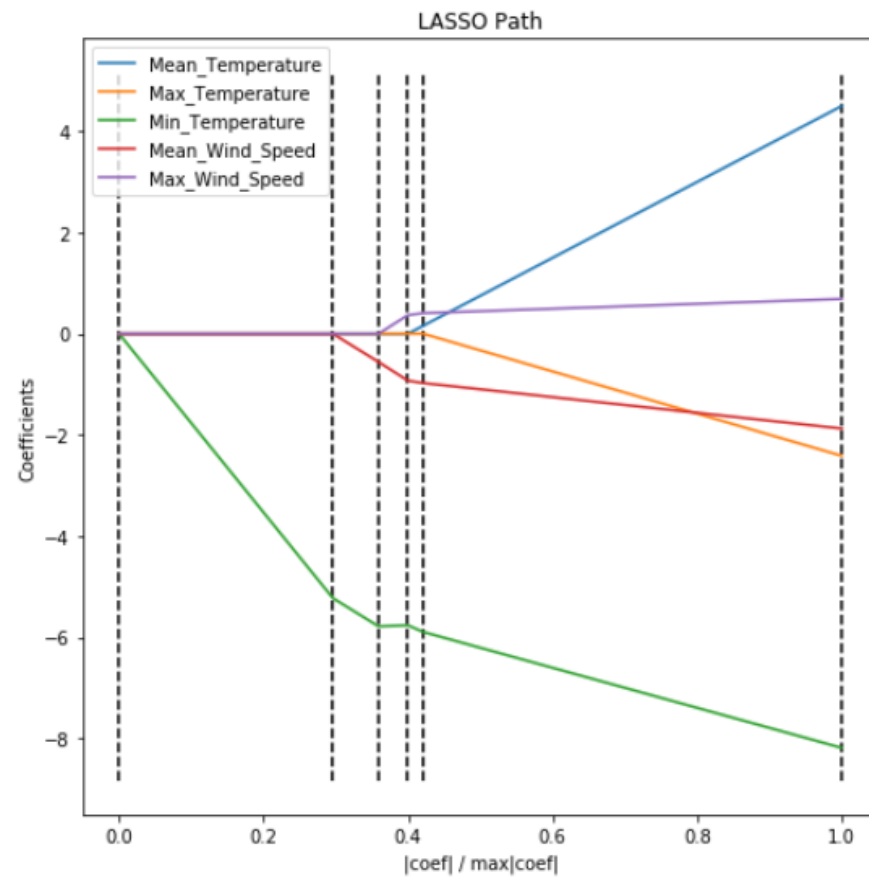
<b>Dep. Variable:</b>	Daily_Rainfall(mm)	<b>R-squared (uncentered):</b>	0.375
<b>Model:</b>	OLS	<b>Adj. R-squared (uncentered):</b>	0.374
<b>Method:</b>	Least Squares	<b>F-statistic:</b>	217.5
<b>Date:</b>	Thu, 25 Jul 2019	<b>Prob (F-statistic):</b>	5.76e-182
<b>Time:</b>	17:12:44	<b>Log-Likelihood:</b>	-6908.9
<b>No. Observations:</b>	1814	<b>AIC:</b>	1.383e+04
<b>Df Residuals:</b>	1809	<b>BIC:</b>	1.386e+04
<b>Df Model:</b>	5		
<b>Covariance Type:</b>	nonrobust		

	coef	std err	t	P> t	[0.025	0.975]
<b>Mean_Temperature</b>	5.5718	0.492	11.331	0.000	4.607	6.536
<b>Max_Temperature</b>	-0.7129	0.254	-2.812	0.005	-1.210	-0.216
<b>Min_Temperature</b>	-5.0204	0.352	-14.250	0.000	-5.711	-4.329
<b>Mean_Wind_Speed</b>	-1.5635	0.167	-9.337	0.000	-1.892	-1.235
<b>Max_Wind_Speed</b>	0.3478	0.042	8.188	0.000	0.264	0.431

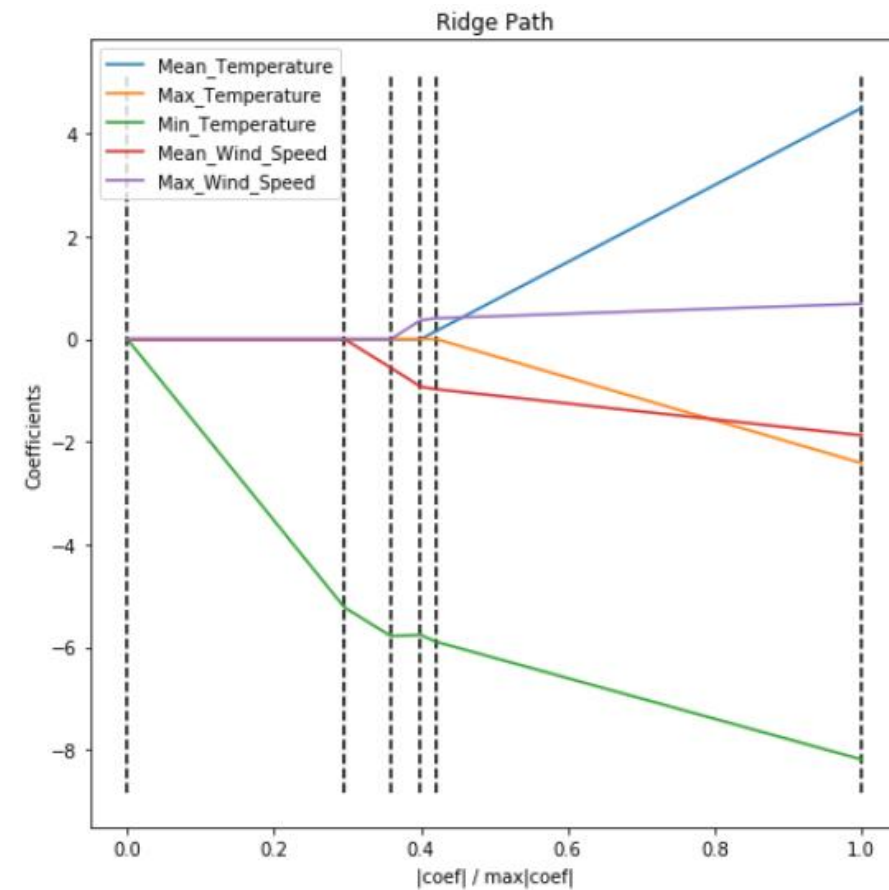
<b>Omnibus:</b>	1172.831	<b>Durbin-Watson:</b>	1.805
<b>Prob(Omnibus):</b>	0.000	<b>Jarque-Bera (JB):</b>	14136.383
<b>Skew:</b>	2.915	<b>Prob(JB):</b>	0.00
<b>Kurtosis:</b>	15.371	<b>Cond. No.</b>	142.

# Findings – LASSO vs Ridge Models

LASSO



RIDGE



# Findings – Model Selection

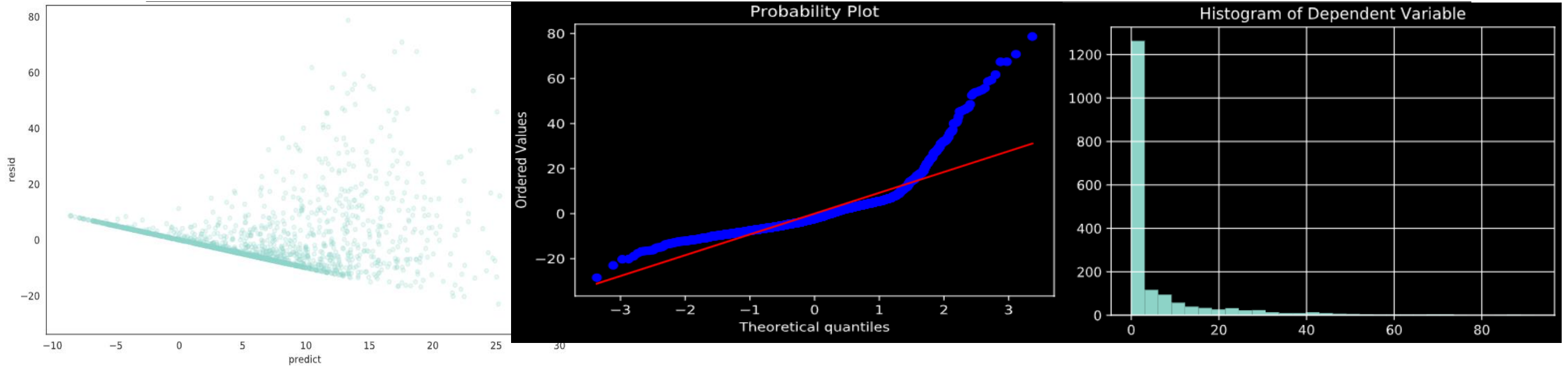
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APPROACH: TRAINING (60%) VS VALIDATION (20%) VS TEST (20%)

	Validation(20%)	Test(20%)					Mean		
Linear Regression	0.356	0.32781315	0.27783897	0.20188714	0.28675351	0.24299511	0.267	+-	0.042
Ridge	0.359	0.32499962	0.27858819	0.20356542	0.28633999	0.24957201	0.269	+-	0.04
LASSO	0.345	0.30278124	0.24636422	0.21276213	0.25359426	0.2741038	0.258	+-	0.03
Polynomial deg=2	0.379	0.42478139	0.40851773	0.03682591	0.2427548	0.31004752	0.285	+-	0.141

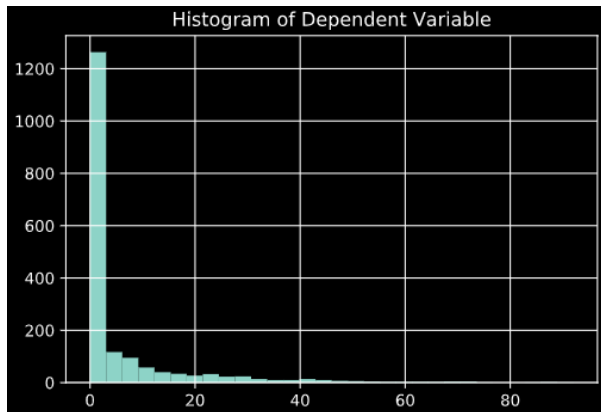
**Select Model: Linear Regression**

# Findings – Linear Regression Model Evaluation

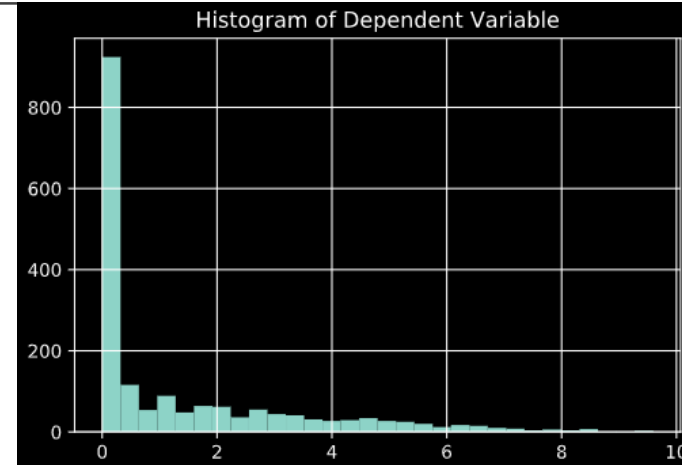
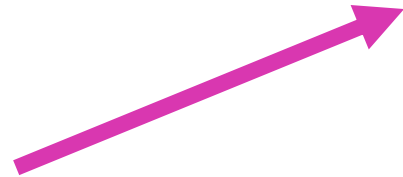


**Evaluation:** The raw Model cant predict extreme target values

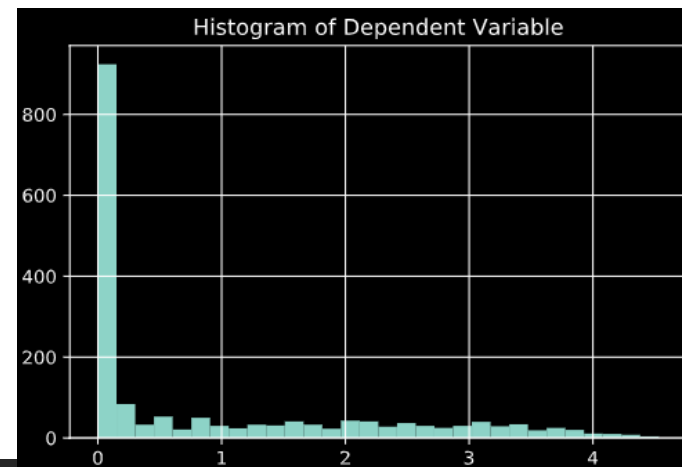
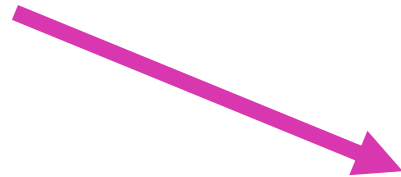
# Findings – Model Evaluation



Square root of  $y$



Log  $y$

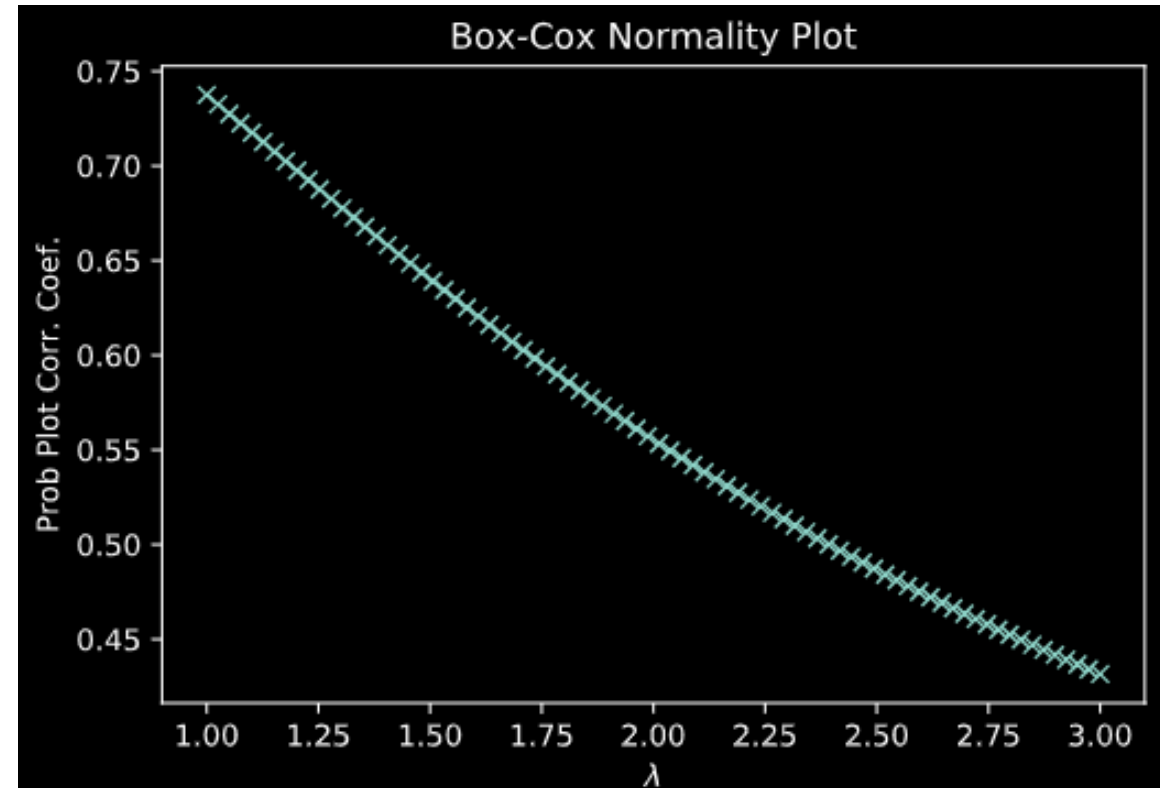


# Conclusion

Historical Temperature and Wind Speed alone are **not** good parameters to predict Rainfall at Newton.

Probable reasons:

1. Cloud movements/weather in surrounding areas.
2. There is a seasonal/time dimension that is not accounted.





# Thank you

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