

### INFO

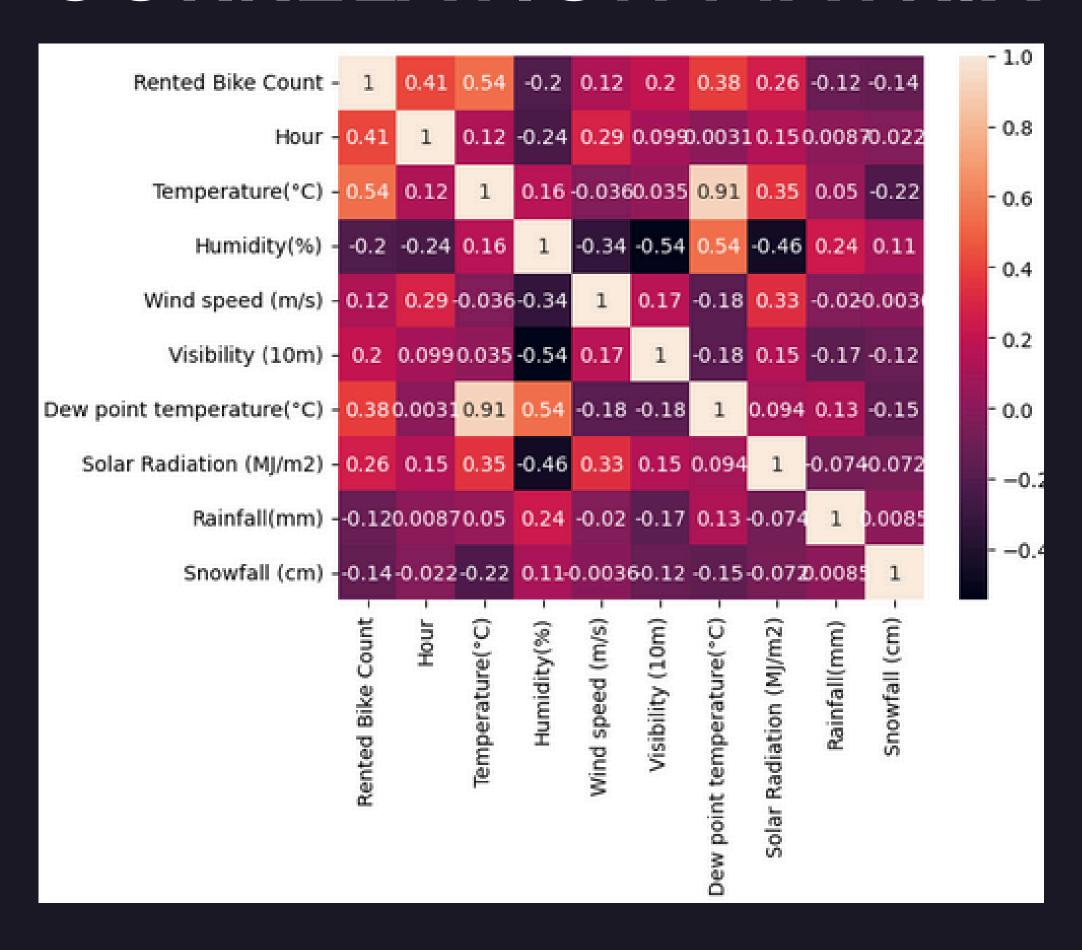
#### 8760 Records

#### 14 Columns

	Date	Rented Bike Count	Hour	Temperature(°C)	Humidity(%)	Wind speed (m/s)	Visibility (10m)	Dew point temperature(°C)	Solar Radiation (MJ/m2)	Rainfall(mm)	Snowfall (cm)	Seasons	Holiday	Functioning Day
0	01/12/2017	254	0	-5.2	37	2.2	2000	-17.6	0.0	0.0	0.0	Winter	No Holiday	Yes
1	01/12/2017	204	1	-5.5	38	8.0	2000	-17.6	0.0	0.0	0.0	Winter	No Holiday	Yes
2	01/12/2017	173	2	-6.0	39	1.0	2000	-17.7	0.0	0.0	0.0	Winter	No Holiday	Yes
3	01/12/2017	107	3	-6.2	40	0.9	2000	-17.6	0.0	0.0	0.0	Winter	No Holiday	Yes
4	01/12/2017	78	4	-6.0	36	2.3	2000	-18.6	0.0	0.0	0.0	Winter	No Holiday	Yes

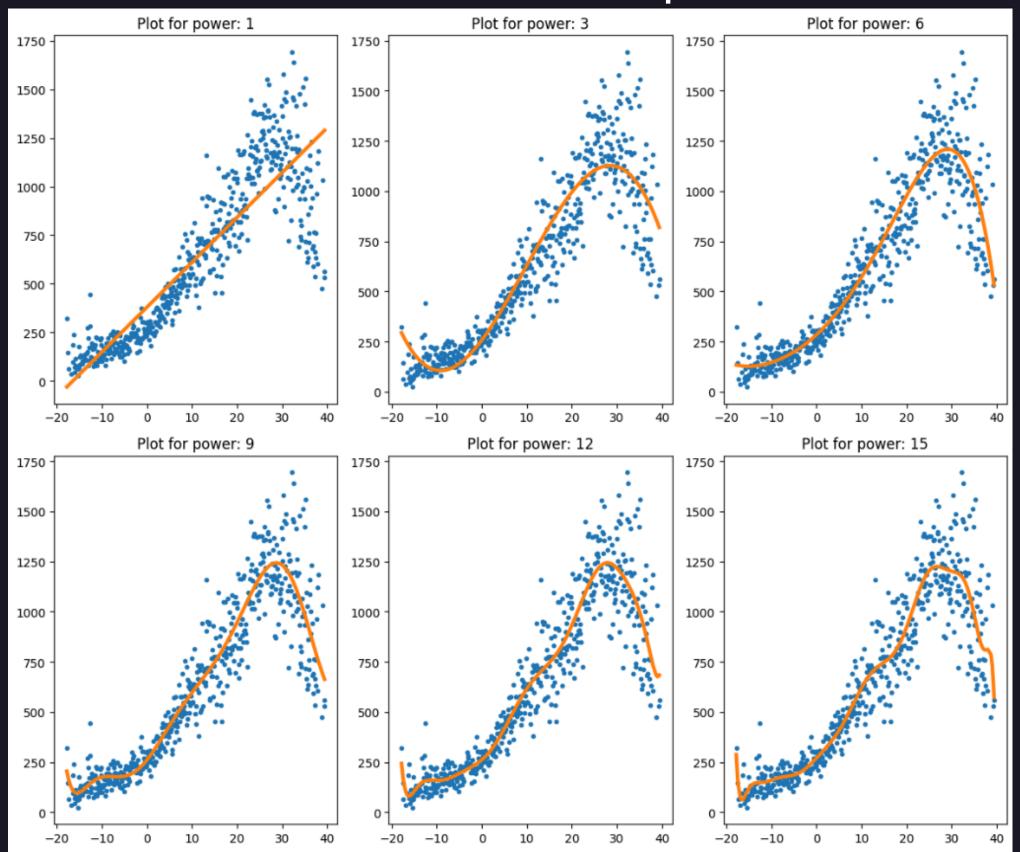
<clas< th=""><th colspan="8"><class 'pandas.core.frame.dataframe'=""></class></th></clas<>	<class 'pandas.core.frame.dataframe'=""></class>										
Range	RangeIndex: 8760 entries, 0 to 8759										
Data	Data columns (total 14 columns):										
#	Column	Non-Null Count	Dtype								
0	Date	8760 non-null	object								
1	Rented Bike Count	8760 non-null	int64								
2	Hour	8760 non-null	int64								
3	Temperature(°C)	8760 non-null	float64								
4	Humidity(%)	8760 non-null	int64								
5	Wind speed (m/s)	8760 non-null	float64								
6	Visibility (10m)	8760 non-null	int64								
7	Dew point temperature(°C)	8760 non-null	float64								
8	Solar Radiation (MJ/m2)	8760 non-null	float64								
9	Rainfall(mm)	8760 non-null	float64								
10	Snowfall (cm)	8760 non-null	float64								
11	Seasons	8760 non-null	object								
12	Holiday	8760 non-null	object								
13	Functioning Day	8760 non-null	object								
dtypes: float64(6), int64(4), object(4)											
memo	memory usage: 958.2+ KB										

### CORRELATION MATRIX



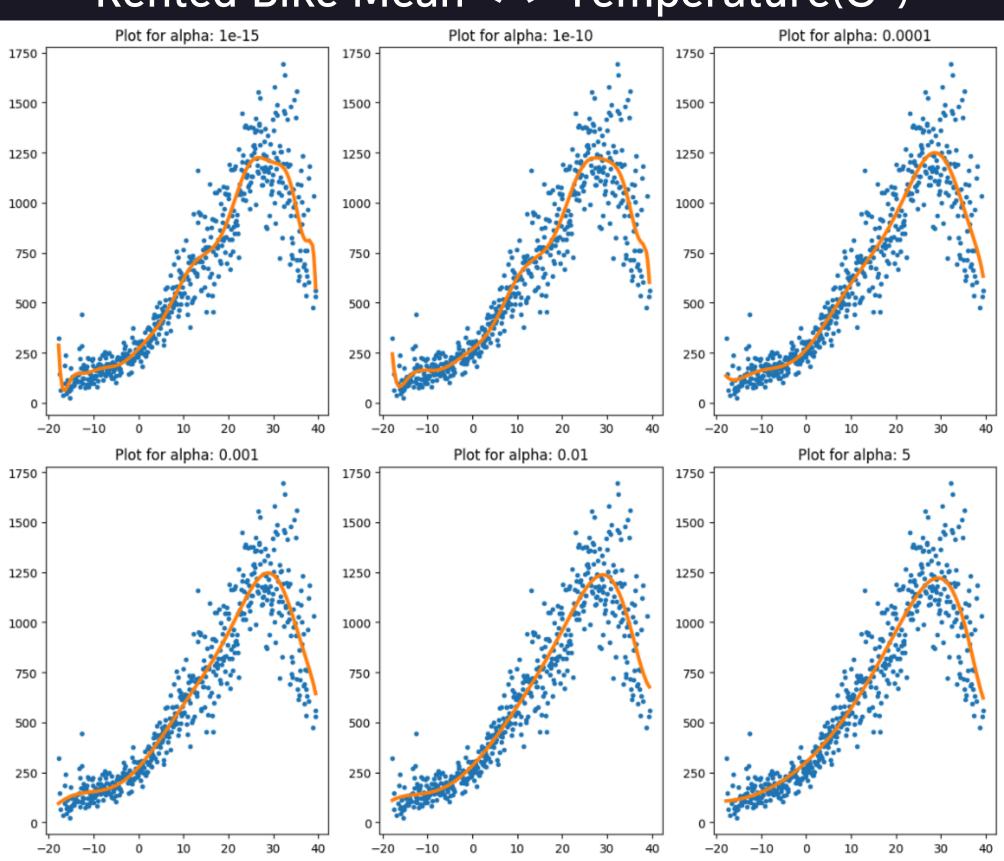
### LINEAR REGRESSION

Rented Bike Mean <-> Temperature(C°)



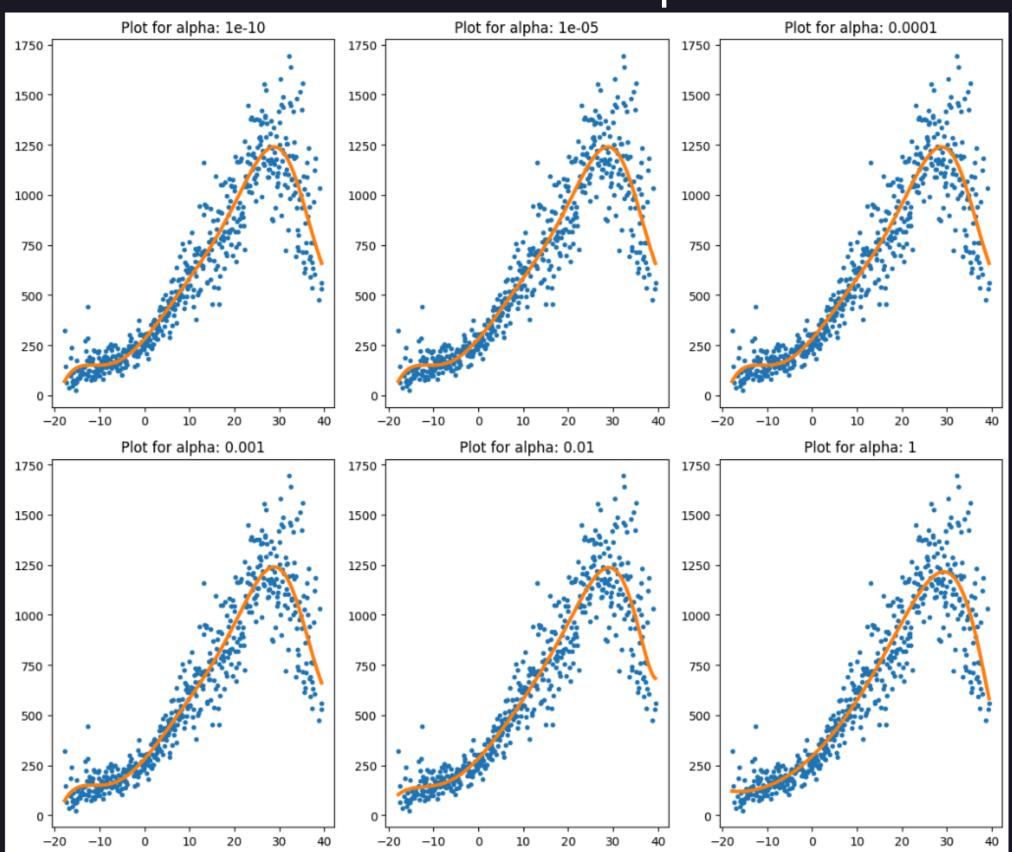
### RIDGE REGRESSION

Rented Bike Mean <-> Temperature(C°)



### LASSO REGRESSION

### Rented Bike Mean <-> Temperature(C°)



### REGRESSION RESULTS

#### Rented Bike Mean <-> Temperature(C°)

#### Linear Regression

	rss	rmse	r2_score	intercept
model_pow_1	2.1166e+07	196.89	0.77497	632.29
model_pow_2	2.0669e+07	194.56	0.78025	632.29
model_pow_3	1.2306e+07	150.13	0.86917	632.29
model_pow_4	1.0579e+07	139.2	0.88752	632.29
model_pow_5	1.0476e+07	138.52	0.88862	632.29
model_pow_6	1.0476e+07	138.52	0.88862	632.29
model_pow_7	1.0331e+07	137.55	0.89016	632.29
model_pow_8	1.0159e+07	136.41	0.89199	632.29
model_pow_9	1.0104e+07	136.04	0.89257	632.29
model_pow_10	1.0086e+07	135.91	0.89277	632.29
model_pow_11	1.0085e+07	135.91	0.89277	632.29
model_pow_12	9.9987e+06	135.32	0.8937	632.29
model_pow_13	9.9904e+06	135.27	0.89378	632.29
model_pow_14	9.9428e+06	134.95	0.89429	632.29
model_pow_15	9.836e+06	134.22	0.89543	632.29

#### Ridge Regression

	rss	rmse	r2_score	intercept
alpha_1e-15	9.836e+06	134.22	0.89543	632.29
alpha_1e-10	9.8713e+06	134.46	0.89505	632.29
alpha_1e-08	9.9589e+06	135.05	0.89412	632.29
alpha_0.0001	1.0111e+07	136.08	0.8925	632.29
alpha_0.001	1.0171e+07	136.48	0.89186	632.29
alpha_0.01	1.0243e+07	136.97	0.8911	632.29
alpha_1	1.0331e+07	137.56	0.89016	632.29
alpha_5	1.0404e+07	138.04	0.88939	632.29
alpha_10	1.055e+07	139.01	0.88783	632.29
alpha_20	1.0928e+07	141.47	0.88381	632.29

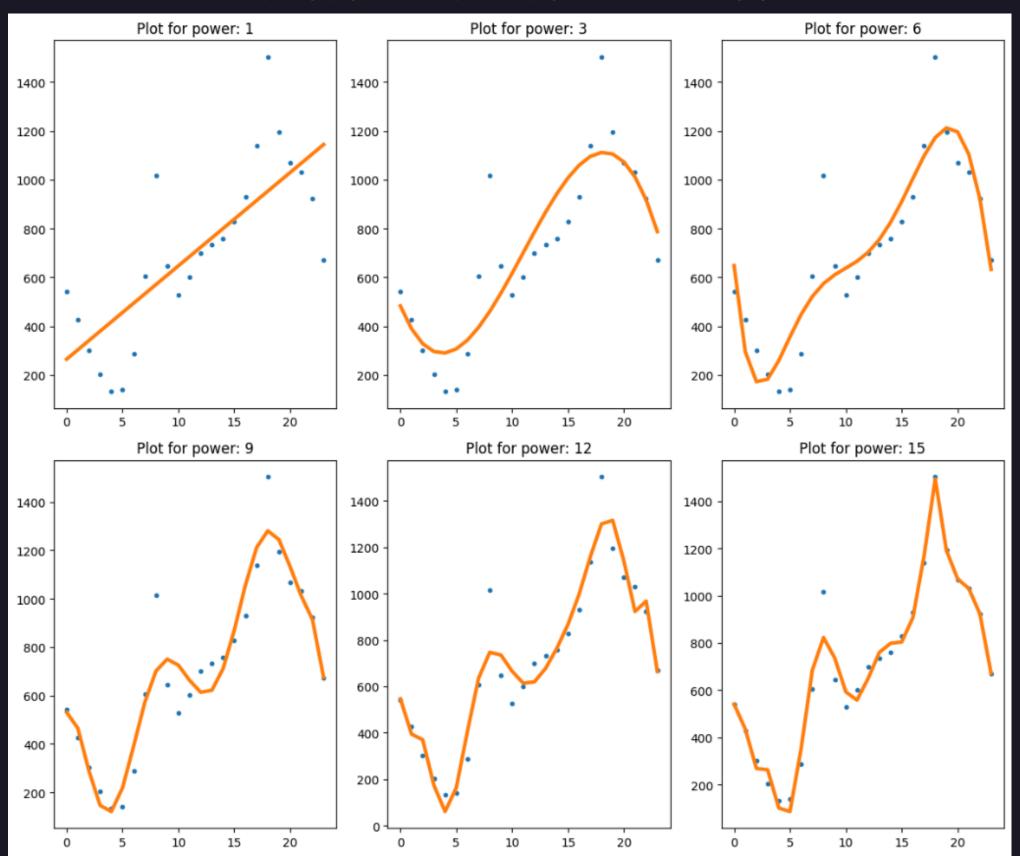
#### Lasso Regression

	rss	rmse	r2_score	intercept
alpha_1e-15	1.0221e+07	136.82	0.89134	632.29
alpha_1e-10	1.0221e+07	136.82	0.89134	632.29
alpha_1e-08	1.0221e+07	136.82	0.89134	632.29
alpha_1e-05	1.0221e+07	136.82	0.89134	632.29
alpha_0.0001	1.0221e+07	136.82	0.89133	632.29
alpha_0.001	1.0221e+07	136.82	0.89133	632.29
alpha_0.01	1.0248e+07	137	0.89104	632.29
alpha_1	1.0403e+07	138.03	0.88939	632.29
alpha_5	1.0965e+07	141.71	0.88342	632.29
alpha_10	1.1932e+07	147.83	0.87315	632.29

alpha\_1e-05 0
alpha\_0.0001 0
alpha\_0.001 1
alpha\_0.01 3
alpha\_1 10
alpha\_5 11
alpha\_10 12

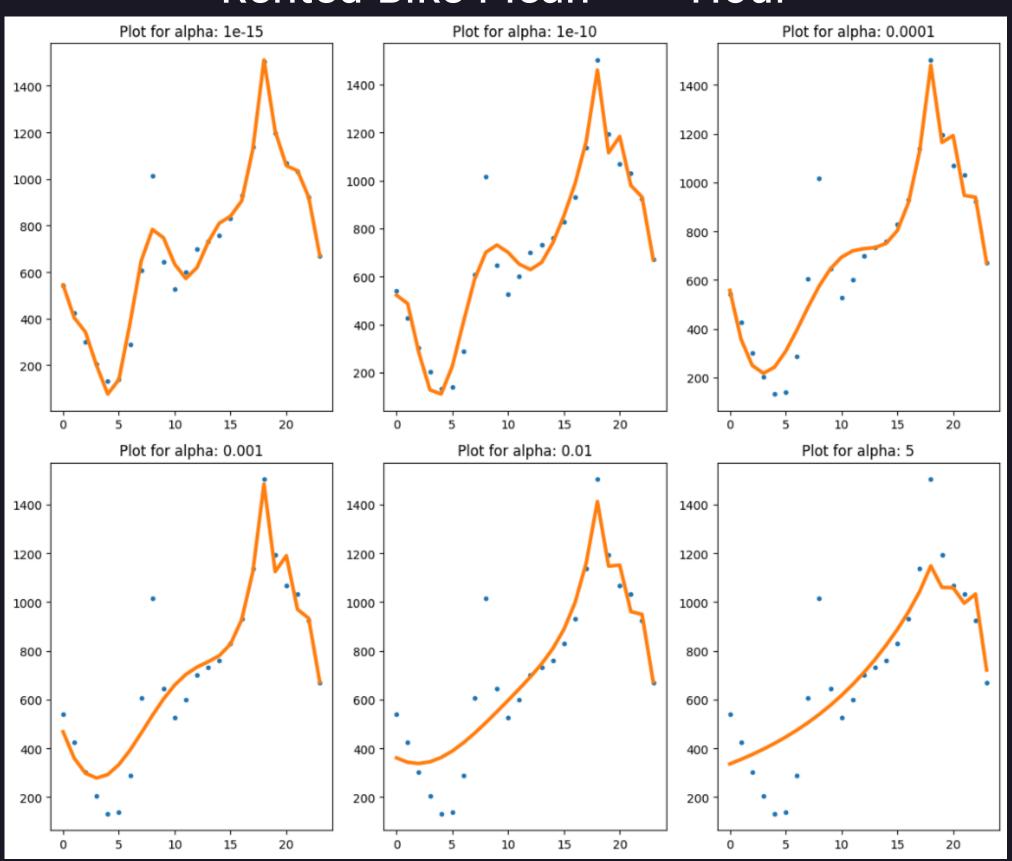
### LINEAR REGRESSION

#### Rented Bike Mean <-> Hour



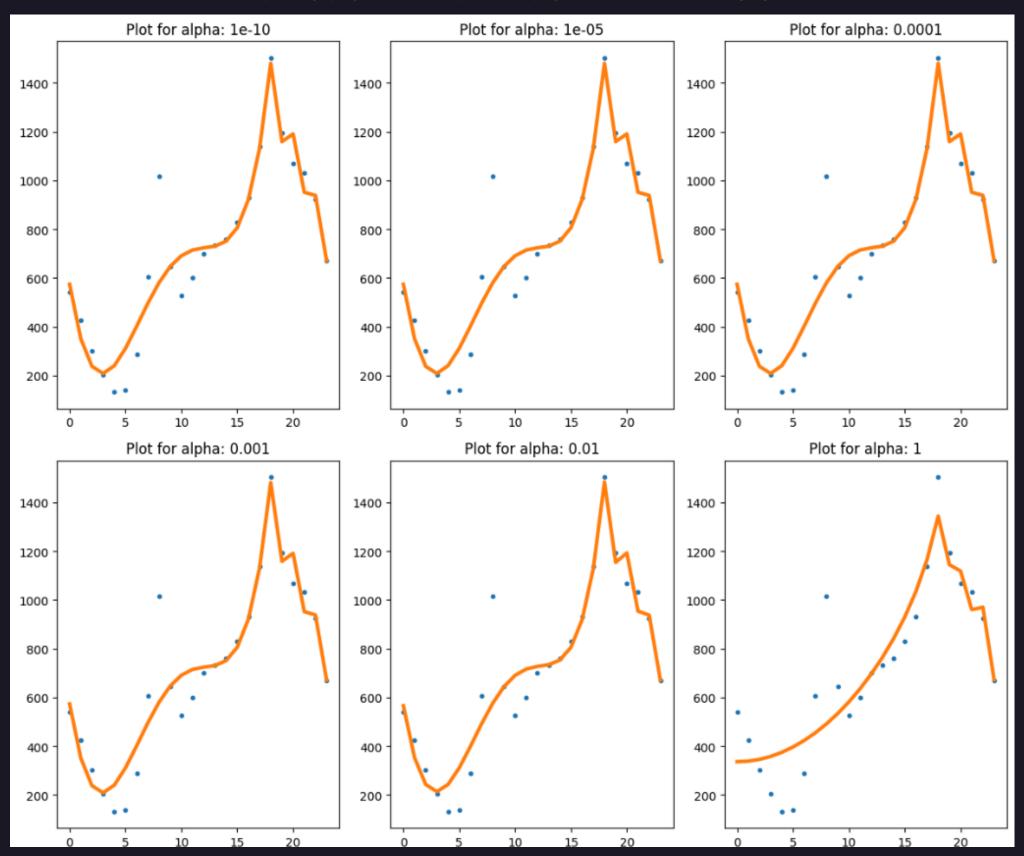
### RIDGE REGRESSION

#### Rented Bike Mean <-> Hour



## LASSO REGRESSION

#### Rented Bike Mean <-> Hour



### REGRESSION RESULTS

#### Rented Bike Mean <-> Hour

#### Linear Regression

	rss	rmse	r2_score	intercept
model_pow_1	1.2353e+06	226.87	0.57631	704.6
model_pow_2	1.2056e+06	224.13	0.58651	704.6
model_pow_3	7.3773e+05	175.32	0.74697	704.6
model_pow_4	7.2669e+05	174.01	0.75076	704.6
model_pow_5	5.2965e+05	148.56	0.81834	704.6
model_pow_6	5.0505e+05	145.06	0.82678	704.6
model_pow_7	3.4017e+05	119.05	0.88333	704.6
model_pow_8	2.9652e+05	111.15	0.8983	704.6
model_pow_9	2.7708e+05	107.45	0.90497	704.6
model_pow_10	2.7435e+05	106.92	0.9059	704.6
model_pow_11	2.1882e+05	95,485	0.92495	704.6
model_pow_12	2.1665e+05	95.01	0.92569	704.6
model_pow_13	1.0141e+05	65.002	0.96522	704.6
model_pow_14	1.0117e+05	64.925	0.9653	704.6
model_pow_15	77,362	56.775	0.97347	704.6

#### Ridge Regression

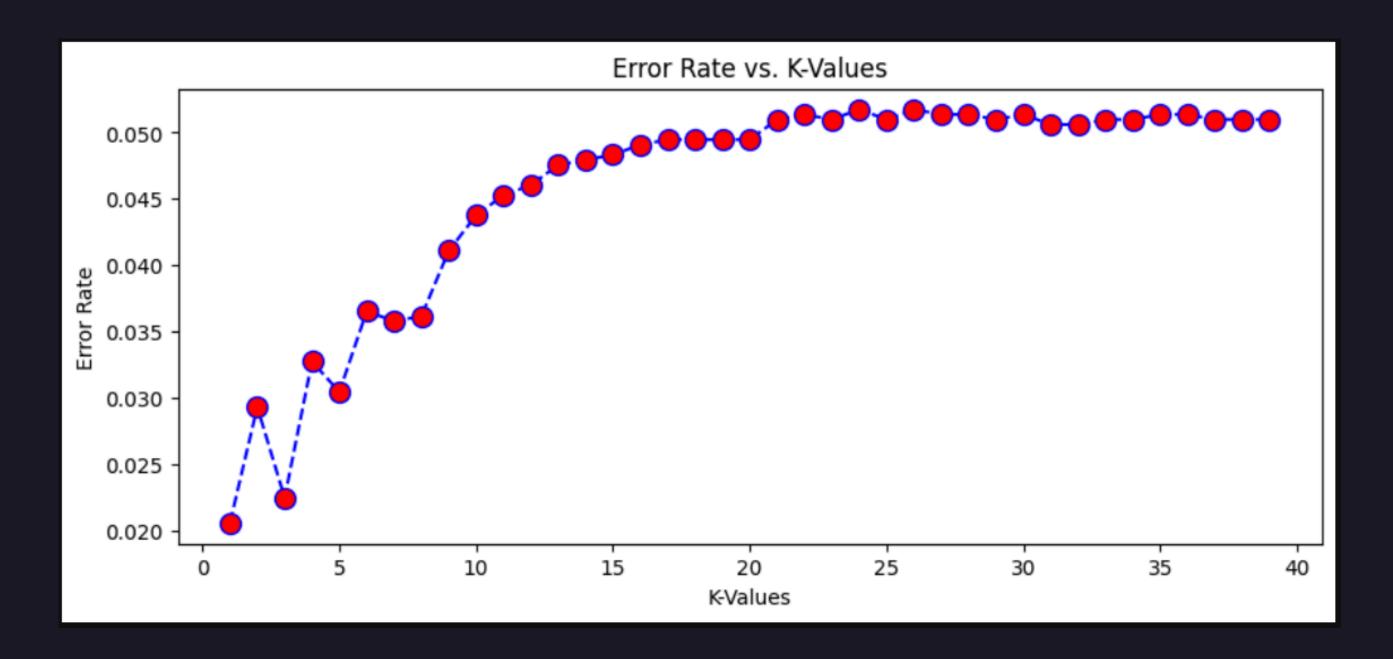
	rss	rmse	r2_score	intercept
alpha_1e-15	1.0258e+05	65.379	0.96482	704.6
alpha_1e-10	2.1195e+05	93.975	0.9273	704.6
alpha_1e-08	2.2886e+05	97.652	0.92151	704.6
alpha_0.0001	3.3532e+05	118.2	0.88499	704.6
alpha_0.001	3.974e+05	128.68	0.8637	704.6
alpha_0.01	5.2566e+05	148	0.81971	704.6
alpha_1	6.3369e+05	162.49	0.78266	704.6
alpha_5	7.346e+05	174.95	0.74805	704.6
alpha_10	8.6187e+05	189.5	0.7044	704.6
alpha_20	1.086e+06	212.72	0.62752	704.6

#### Lasso Regression

	rss	rmse	r2_score	intercept
alpha_1e-15	3.3092e+05	117.42	0.8865	704.6
alpha_1e-10	3.3092e+05	117.42	0.8865	704.6
alpha_1e-08	3.3092e+05	117.42	0.8865	704.6
alpha_1e-05	3.3093e+05	117.42	0.8865	704.6
alpha_0.0001	3.3096e+05	117.43	0.88649	704.6
alpha_0.001	3.3128e+05	117.49	0.88638	704.6
alpha_0.01	3.3475e+05	118.1	0.88519	704.6
alpha_1	6.0005e+05	158.12	0.79419	704.6
alpha_5	6.2946e+05	161.95	0.78411	704.6
alpha_10	6.689e+05	166.95	0.77058	704.6

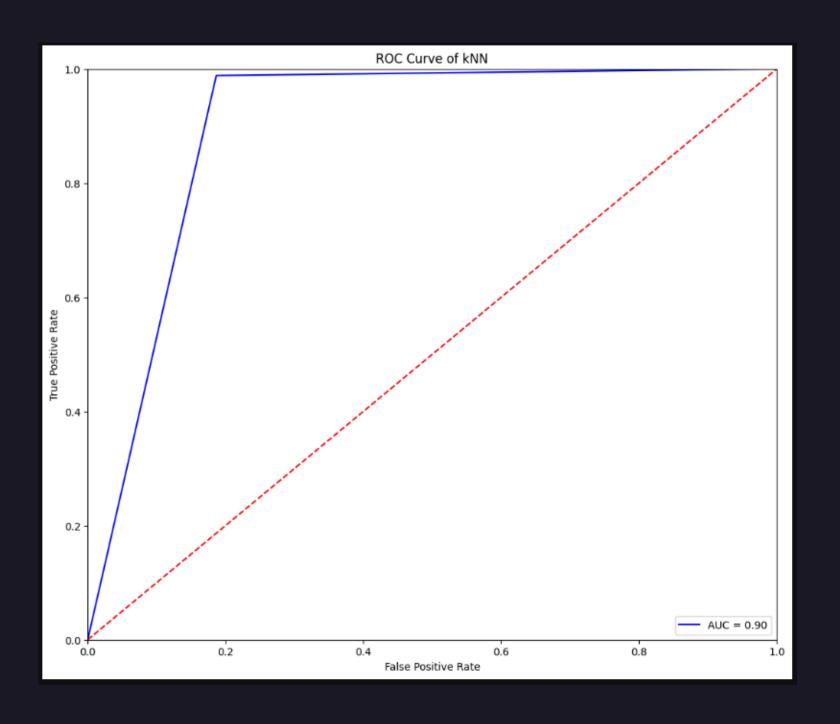
alpha\_1e-05 0
alpha\_0.0001 0
alpha\_0.001 0
alpha\_0.01 0
alpha\_1 10
alpha\_5 11
alpha\_10 11

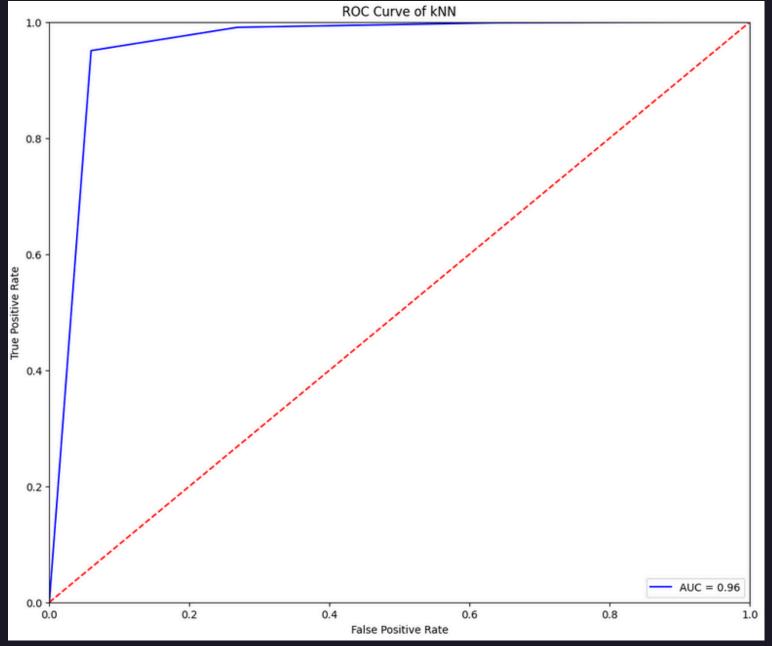
Holidays KNN



	precision	recall	f1-score	support		precision	recall	f1-score	support
	•					•			
Holiday	0.79	0.81	0.80	134	Holiday	0.81	0.73	0.77	134
No Holiday	0.99	0.99	0.99	2494	No Holiday	0.99	0.99	0.99	2494
accuracy			0.98	2628	accuracy			0.98	2628
macro avg	0.89	0.90	0.90	2628	macro avg	0.90	0.86	0.88	2628
weighted avg	0.98	0.98	0.98	2628	weighted avg	0.98	0.98	0.98	2628
TP - True Neg	ative 109				TP - True Negative 98				
FP - False Po	sitive 25				FP - False Positive 36				
FN - False Ne	gative 29				FN - False Negative 23				
TP - True Positive 2465				TP - True Positive 2471					
Accuracy Rate: 0.9794520547945206				Accuracy Rate: 0.9775494672754946					
Misclassification Rate: 0.02054794520547945				Misclassification Rate: 0.022450532724505326					

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Seasons SVM

	precision	recall	f1-score	support
Autumn	0.64	0.58	0.61	680
Spring	0.63	0.65	0.64	624
Summer	0.83	0.86	0.85	659
Winter	0.91	0.94	0.92	665
accuracy			0.76	2628
macro avg	0.75	0.76	0.76	2628
weighted avg	0.76	0.76	0.76	2628

Holidays

**Logistic Regression** 

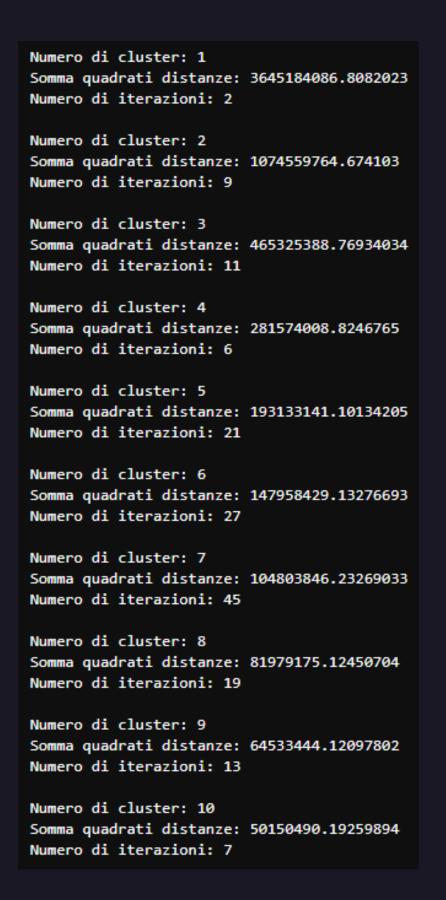
	precision	recall	f1-score	support
Holiday	0.00	0.00	0.00	149
No Holiday	0.95	1.00	0.98	2917
accuracy			0.95	3066
macro avg	0.48	0.50	0.49	3066
weighted avg	0.91	0.95	0.93	3066

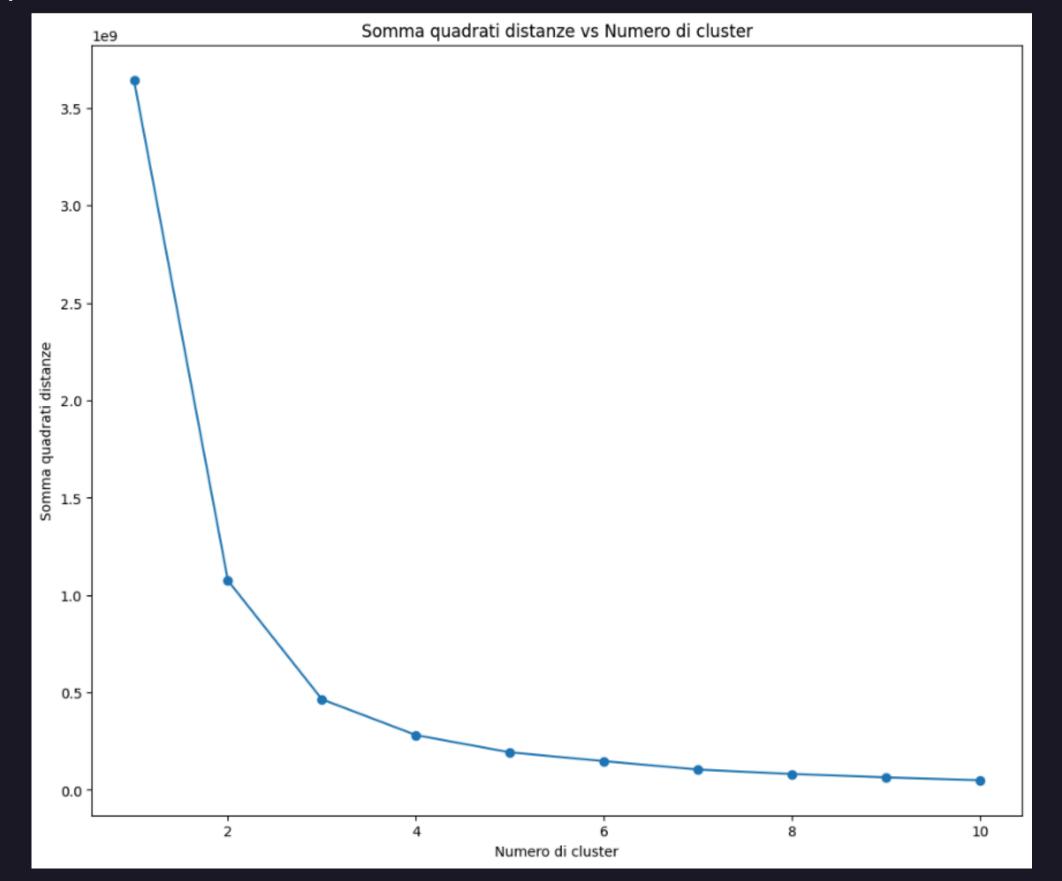
Holidays Random Forest

	precision	recall	f1-score	support
Holiday	0.85	0.15	0.25	149
No Holiday	0.96	1.00	0.98	2917
accuracy			0.96	3066
macro avg	0.90	0.57	0.61	3066
weighted avg	0.95	0.96	0.94	3066

### K-MEANS CLUSTERING

#### Temperature(C°) <-> Rented Bike Count





### K-MEANS CLUSTERING

Temperature(C°) <-> Rented Bike Count

#### Centroidi ottenuti:

[[ 5.88899859 195.33897508] [ 20.82094102 1321.47647449] [ 23.32081737 2170.33971903] [ 17.21933153 744.2899729 ]]

Somma quadrati distanze: 281574008.825

Numero di iterazioni: 6

