BIKE DEMAND PRESENTATION

Laïd ATTIA 24 November 2022

EXECUTIVE SUMMARY

- Bike demands are influenced by cities, available bicycles for rent, seasons, temperature, hour of the day and holidays
- Linear regression model is recommended to predict bike demand

Introduction

- Project is about how weather would affect bike-sharing demand in urban areas
- Data collection and sources
- Data exploration and analysis
- Data wrangling
- Data modelling
- Interactive Dashboard
- conclusion

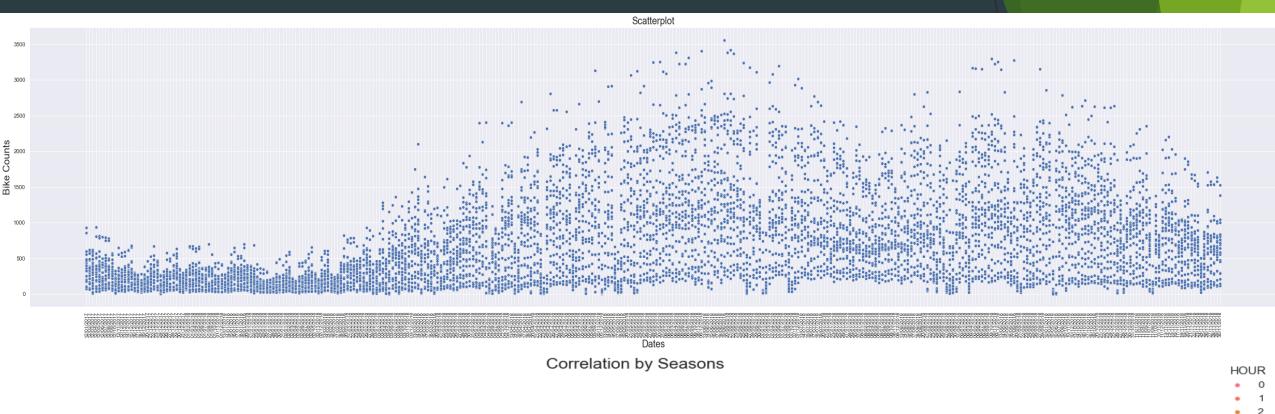
methodology

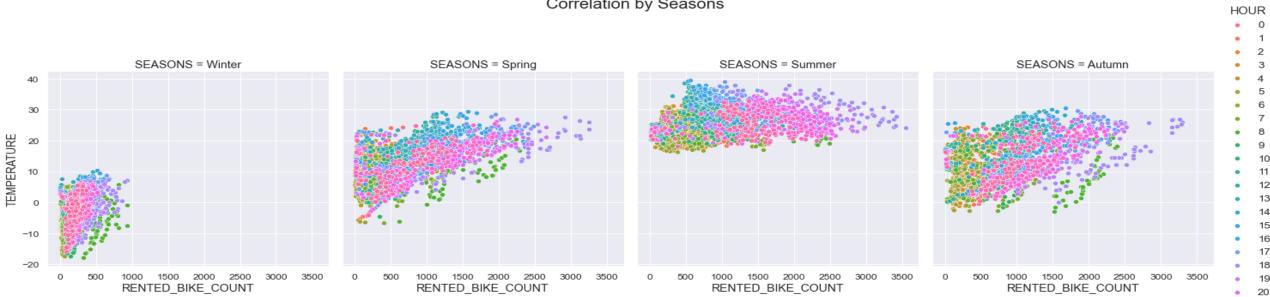
- Web scraping from Wikipedia (global bike sharing systems)
- 5 day weather forecasts for cities using openweather api
- Perform data wrangling on csv files
- Perform data exploration
- Perform data visualisation
- Predict Hourly Rented Bike Count using Basic Linear Regression Models
- Refine the Baseline Regression Models

Exploratory data analysis

- Dataset has 8465 observations for seoul
- Date range is from 1 dec 2017 to 30 Nov 2018 (one year)
- Highest bike count is 3556 on 19 June 2018
- Summer has most bike rentals, winter the least
- Demand varies from city to city

data visualization





Predictive analysis

baseline regression model

improving the model



	Model	MAE	MSE	RMSE	R2	RMSLE	MAPE	TT (Sec)
lr	Linear Regression	283.0143	142660.9000	377.6294	0.6422	0.8924	1.5912	1.1580
ridge	Ridge Regression	282.9418	142645.2344	377.6091	0.6422	0.8941	1.5886	0.0140
lar	Least Angle Regression	283.0429	142767.5492	377.7662	0.6420	0.8922	1.5900	0.0160
lasso	Lasso Regression	283.3254	143644.9844	378.9253	0.6398	0.8928	1.5563	0.0400
llar	Lasso Least Angle Regression	354.6801	227398.6736	476.6986	0.4301	0.9387	1.9962	0.0160
en	Elastic Net	359.0337	235343.4406	484.8899	0.4107	0.9293	1.7635	0.0160

Interactive Dashboard

Bike-Sharing Demand Prediction App

Bike Sharing data

	DATE	RENTED_BIKE_COUNT	HOUR	TEMPERATURE	HUMIDITY	WI
•	2017-01-12T00:00:00+08	254	•	-5.2000	37	
1	2017-01-12T00:00:00+08	204	1	-5.5000	38	
2	2017-01-12T00:00:00+08	173	2	-6	39	
3	2017-01-12T00:00:00+08	107	3	-6.2000	40	
4	2017-01-12T00:00:00+08	78	4	-6	36	
	<					>

Selected Cities Data

	IS03	IS02	COUNTRY	lon	lat	CITY_ASCII	CITY	
	KOR	KR	Korea, South	127	37.5833	Seoul	Seoul	•
	USA	US	United States	-73.9249	40.6943	New York	New York	1
Î	FRA	FR	France	2.3522	48.8566	Paris	Paris	2
Lon	GBR	GB	United Kingdom	-0.1275	51.5072	London	London	3
	CHN	CN	China	120.6164	31.3040	Suzhou	Suzhou	4
>							<	

All

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Selections

- Basic max bike prediction
- A static temperature trend line
- An interactive bike-sharing demand prediction trend line
- A static humidity and bike-sharing demand prediction correlation plot

conclusion

- Bike demands are influenced by cities, available bicycles for rent, seasons, temperature, hour of the day and holidays
- Linear regression model is recommended to predict bike demand
- Explore more sources on bikes demand online for analysis
- Use machine learning algorithms (decision tree, random forest, xgboost, gradient boost)
- Potential use for marketing and business expansions

appendix

Python code sample

```
fig = plt.figure(figsize=(50,10))
sns.scatterplot(x=df.DATE,y=df.RENTED_BIKE_COUNT,data=df, estimator=None)
plt.title("Scatterplot", fontsize=20)
plt.xticks(rotation=90, fontsize=10)
plt.xlabel("Dates", fontsize=20)
plt.ylabel("Bike Counts", fontsize=20)
plt.show()
```

Sql code sample

