Linear Regression – Diamonds Prices

Given Data Base:

Raw table (CSV Format) that contains an data on Diamonds. The variables are as following:

- Cut
- Color
- Clarity
- Depth
- Table
- X
- Y
- Z
- Price

The task:

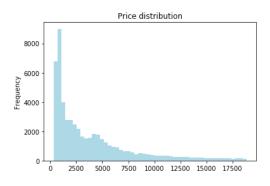
Predict Prices of diamonds based on its characteristics

.

Part I – 2D Linear Regression Model

Explanatory analysis:

Price analysis -



Aggregations:

 max
 18823.000000

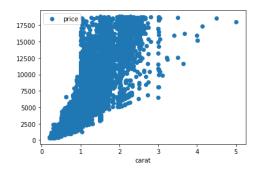
 mean
 3932.799722

 median
 2401.000000

 min
 326.000000

 $\label{eq:df['price'].agg(['max', 'mean', 'median', 'min'])} df['price'].plot(kind='hist', bins=50, color='lightblue', title='Price distribution')$

Correlation between Carat & Price:



We can see that the correlation between Carat and Price is positive.

df.plot(x='carat',y='price', style='o')

2D Linear Regression:

Defining variables & splitting the data -

Independent Variable (Input) - Carat

Dependent Variable (Output) - Price

X = df['carat'].values.reshape(-1,1) # Independed Variable - Input y = df['price'].values.reshape(-1,1) # Depended Variable - Output

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)

Fitting Regressor & Predicting -

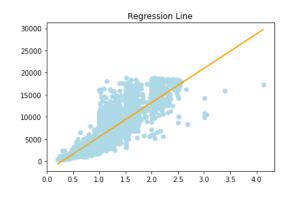
```
regressor = LinearRegression()
regressor.fit(X_train,y_train)
```

y_pred = regressor.predict(X_test)

Results -

R-Squared: 0.8515758113126248 Coefficient: 7745.256582433882

Intercept: -2248.460057551038

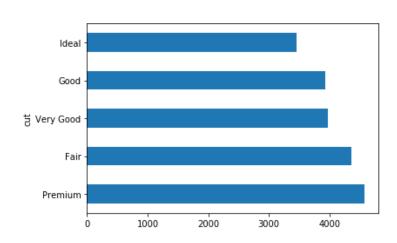


print('R-Squared: ', metrics.r2_score(y_test,y_pred)) print('Coefficient: ',regressor.coef_[0][0]) print('Intercept: ',regressor.intercept_[0]) plt.scatter(X_test,y_test, color='lightblue') plt.plot(X_test,y_pred, color='orange') plt.title('Regression Line') plt.show()

Part II - Multiple Linear Regression Model

Converting 'Cut' into Ordinal Variable –

We can see that the price is affected by the cut level. Therefore, we converted Cut from Categorial into Ordinal variable (level 1 to 5).



df.groupby('cut').mean()['price'].sort_values(ascending=False).plot(kind='barh')

df['cut'].replace({'Ideal': 5, 'Premium': 4, 'Good': 3, 'Very Good': 2, 'Fair':1}, inplace=True)

<u>Creating Additional Variable – Size</u>

With multiplying X, Y and Z measures, we can create a new variable; Size.

```
df['size'] = df['x']*df['y']*df['z']
```

<u>Correlation Table –</u>

	carat	cut	depth	x	у	Z	size	price
carat	1.000000	-0.114426	0.028224	0.975094	0.951722	0.953387	0.976308	0.921591
cut	-0.114426	1.000000	-0.169916	-0.105361	-0.105319	-0.126726	-0.101119	-0.049421
depth	0.028224	-0.169916	1.000000	-0.025289	-0.029341	0.094924	0.009157	-0.010647
x	0.975094	-0.105361	-0.025289	1.000000	0.974701	0.970772	0.956564	0.884435
у	0.951722	-0.105319	-0.029341	0.974701	1.000000	0.952006	0.975143	0.865421
Z	0.953387	-0.126726	0.094924	0.970772	0.952006	1.000000	0.950065	0.861249
size	0.976308	-0.101119	0.009157	0.956564	0.975143	0.950065	1.000000	0.902385
price	0.921591	-0.049421	-0.010647	0.884435	0.865421	0.861249	0.902385	1.000000

df[['carat','cut','depth','x','y','z','size','price']].corr()

Multiple Variables Linear Regression:

Defining variables & splitting the data -

Independent Variable (Input) - Carat, Cut, Size, X, Y, Z

Dependent Variable (Output) - Price

```
X = df[['cut','carat','size','x','y','z']].values
y = df['price'].values
```

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)

Coefficient Table:

cut 170.735240
carat 8418.080310
size 13.083272
x -154.892973
y -601.625678
z -701.238426

v_def = df[['cut','carat','size','x','y','z']] coeff_df = pd.DataFrame(regressor.coef_,v_def.columns) coeff_df

<u>Fitting Regressor & Predicting -</u>

regressor = LinearRegression() regressor.fit(X_train,y_train)

y_pred = regressor.predict(X_test)

<u>Results -</u>

R-Squared: 0. 0.8607280804681006

print('R-Squared: ', metrics.r2_score(y_test,y_pred))