

```

# Latihan 1
# import pandas
import pandas as pd
# import numpy
import numpy as np
# import Library SelectKBest
from sklearn.feature_selection import SelectKBest
# import Library chi kuadrat/squared
from sklearn.feature_selection import chi2

# load dataset
data = pd.read_csv('Iris.csv')
data

```

	Id	SepalLengthCm	...	PetalWidthCm	Species
0	1	5.1	...	0.2	Iris-setosa
1	2	4.9	...	0.2	Iris-setosa
2	3	4.7	...	0.2	Iris-setosa
3	4	4.6	...	0.2	Iris-setosa
4	5	5.0	...	0.2	Iris-setosa
...
145	146	6.7	...	2.3	Iris-virginica
146	147	6.3	...	1.9	Iris-virginica
147	148	6.5	...	2.0	Iris-virginica
148	149	6.2	...	2.3	Iris-virginica
149	150	5.9	...	1.8	Iris-virginica

[150 rows x 6 columns]

```

# Latihan 2
# Menghilangkan kolom Id
df1 = data.drop('Id', 1)
# lalu tampilkan
df1

```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	
Species					
0	5.1	3.5	1.4	0.2	
Iris-setosa					
1	4.9	3.0	1.4	0.2	
Iris-setosa					
2	4.7	3.2	1.3	0.2	
Iris-setosa					
3	4.6	3.1	1.5	0.2	
Iris-setosa					
4	5.0	3.6	1.4	0.2	
Iris-setosa					
..	
...					
145	6.7	3.0	5.2	2.3	Iris-
virginica					

146	6.3	2.5	5.0	1.9	Iris-
virginica					
147	6.5	3.0	5.2	2.0	Iris-
virginica					
148	6.2	3.4	5.4	2.3	Iris-
virginica					
149	5.9	3.0	5.1	1.8	Iris-
virginica					

[150 rows x 5 columns]

```
# Latihan 3
#independent columns --> SepalLengthCm, SepalWidthCm, PetalLengthCm,
PetalWidthCm
X = df1.iloc[:,0:4]
# target columns --> species
y = df1.iloc[:, -1]
```

```
# Latihan 4
#Apply SelectKBest class to extract
```

```
bestfeature = SelectKBest(score_func=chi2, k=4)
fit = bestfeature.fit(X,y)
dfscores = pd.DataFrame(fit.scores_)
dfcolumns = pd.DataFrame(X.columns)
```

```
# Latihan 5
#gabungkan 2 dataframe tersebut untuk visualisasi yang lebih bagus
```

```
featureScores = pd.concat([dfcolumns, dfscores],axis=1)
featureScores.columns = ['Field', 'Score']
print(featureScores.nlargest(10,'Score'))
```

	Field	Score
2	PetalLengthCm	116.169847
3	PetalWidthCm	67.244828
0	SepalLengthCm	10.817821
1	SepalWidthCm	3.594499

```
# Latihan 6
data = pd.read_csv('Iris.csv')
```

```
# Menghilangkan kolom Id
df2 = data.drop('Id', 1)
# lalu tampilkan
df2
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm
Species				
0	5.1	3.5	1.4	0.2
Iris-setosa				

1	4.9	3.0	1.4	0.2	
Iris-setosa					
2	4.7	3.2	1.3	0.2	
Iris-setosa					
3	4.6	3.1	1.5	0.2	
Iris-setosa					
4	5.0	3.6	1.4	0.2	
Iris-setosa					
...	
...					
145	6.7	3.0	5.2	2.3	Iris-
virginica					
146	6.3	2.5	5.0	1.9	Iris-
virginica					
147	6.5	3.0	5.2	2.0	Iris-
virginica					
148	6.2	3.4	5.4	2.3	Iris-
virginica					
149	5.9	3.0	5.1	1.8	Iris-
virginica					

[150 rows x 5 columns]

```
# Latihan 7
#independent columns --> SepalLengthCm, SepalWidthCm, PetalLengthCm,
PetalWidthCm
A = df2.iloc[:,0:4]
# target columns --> species
b = df2.iloc[:, -1]
```

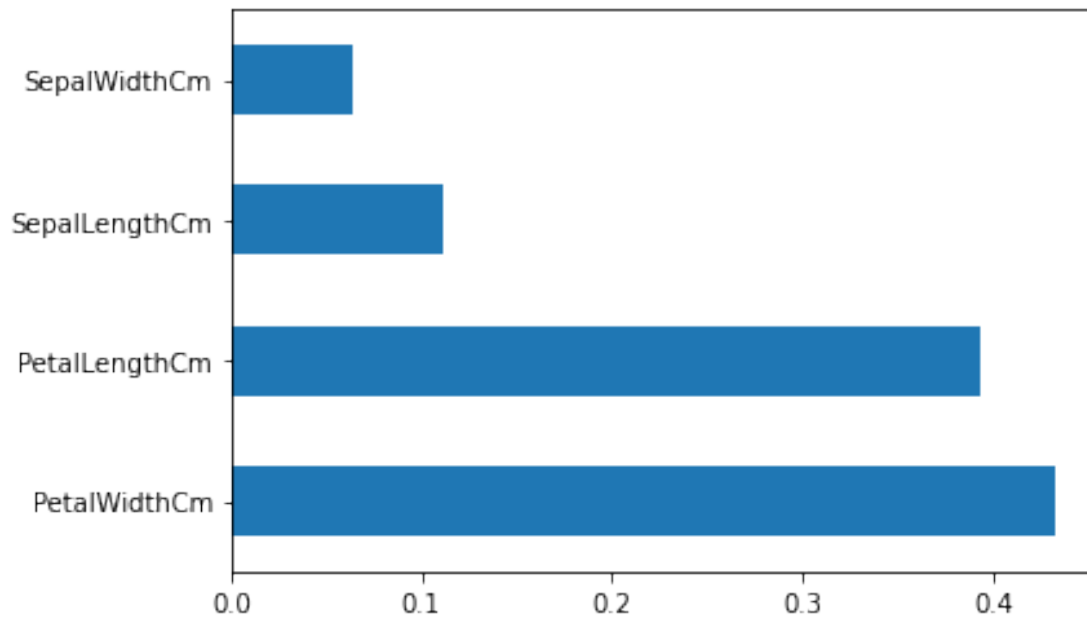
```
# Latihan 8
# Import library ExtraTreesClassifier
from sklearn.ensemble import ExtraTreesClassifier
# Import library matplotlib
import matplotlib.pyplot as plt
```

```
# fit model ExtraTreesClassifier
model = ExtraTreesClassifier()
model.fit(A,b)
```

```
ExtraTreesClassifier()
```

```
# Latihan 9
print(model.feature_importances_)
feat_importance = pd.Series(model.feature_importances_,
index=A.columns)
feat_importance.nlargest(10).plot(kind='barh')
plt.show()
```

```
[0.1118297  0.06335808 0.39293668 0.43187554]
```



```
import pandas as pd
import numpy as np
import seaborn as sns
```

```
data = pd.read_csv('Iris.csv')
df3= data.iloc[:,1:]
df3
```

	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	
Species					
0	5.1	3.5	1.4	0.2	
Iris-setosa					
1	4.9	3.0	1.4	0.2	
Iris-setosa					
2	4.7	3.2	1.3	0.2	
Iris-setosa					
3	4.6	3.1	1.5	0.2	
Iris-setosa					
4	5.0	3.6	1.4	0.2	
Iris-setosa					
..	
...					
145	6.7	3.0	5.2	2.3	Iris-
virginica					
146	6.3	2.5	5.0	1.9	Iris-
virginica					
147	6.5	3.0	5.2	2.0	Iris-
virginica					
148	6.2	3.4	5.4	2.3	Iris-
virginica					
149	5.9	3.0	5.1	1.8	Iris-

virginica

[150 rows x 5 columns]

Latihan 10

#independent columns --> SepalLengthCm, SepalWidthCm, PetalLengthCm, PetalWidthCm

K = df3.iloc[:,0:4]

target columns --> species

j = df3.iloc[:, -1]

mendapatkan korelasi di setiap fitur dalam dataset

corrmat = df3.corr()

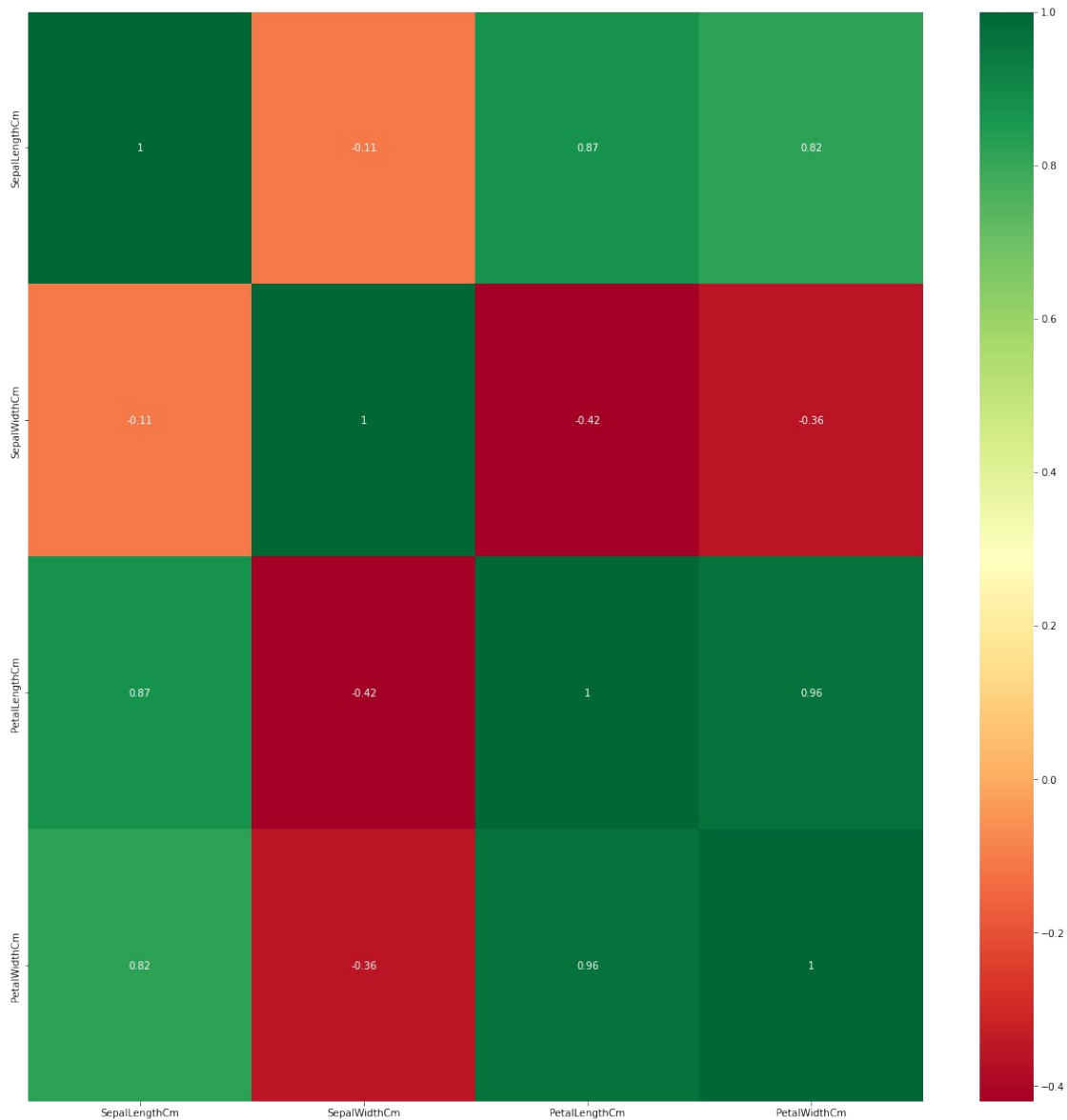
top_corr_features = corrmat.index

plt.figure(figsize=(20,20))

plot heatmap

h =

sns.heatmap(df3[top_corr_features].corr(),annot=True,cmap="RdYlGn")



Kesimpulan :

1. lihat pada baris terakhir yaitu price range, korelasi antara PetalWidthCm dengan fitur lain dimana ada relasi kuat dengan variabel SepalLengthCm dan diikuti oleh var petalLengthCm.
2. sedangkan utk var PetalWidthCm berkorelasi lemah dengan SepalLengthCm