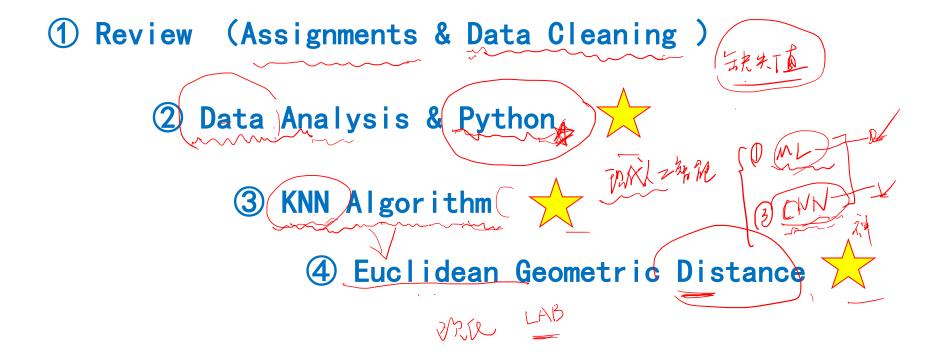
数据挖掘和大数据分析







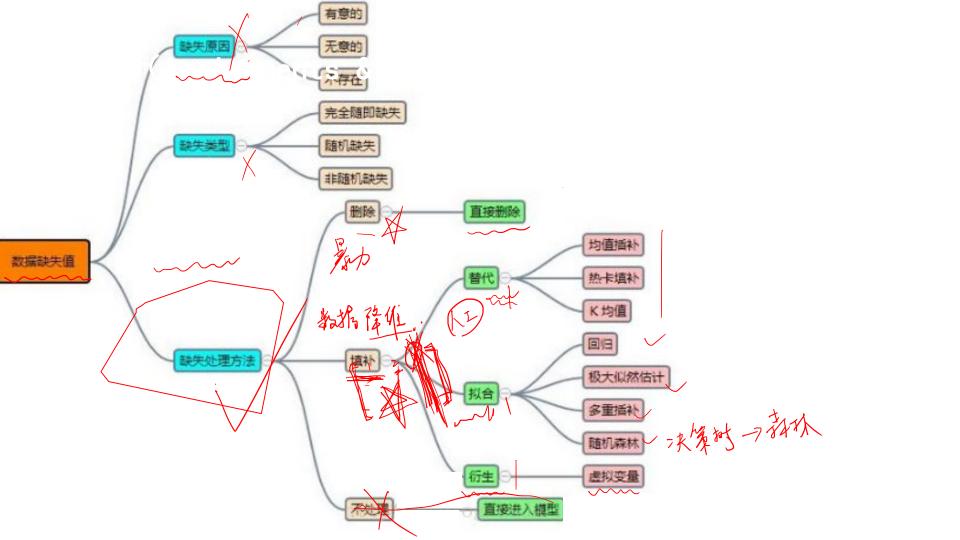


DATA ANALYTICS:

DATA MINING AND BIG DATA







Review (Assignments & Data Cleaning)



[2] 对下图的数据采用删除和填补两种方法进行清洗。

Sepal.Length	Sepal.Width	Petal.Length	Petal Width	Species
5.1	3.5	1.4	0.2	setosa
4.9	3	NA	0.2	setosa
4.7	3.2	NA	0.2	setosa
4.6	3.1	1.5	0.2	setosa
5.1			0.2	setosa
4.6	3.2	1.4	0.2	setosa
5.3	NA	1.5	0.2	setosa
5	3.3	1.4	0.2	setosa
7	7.2	4.7	1.4	versicolor
6.4	3.2	NA	1.5	versicolor
6.3	3.3	6	2.5	virginica
5.8	2.7	5.1	1.9	virginica
NA	3	NA	2.1	virginica
6.3			1.8	virginica
5.9	3	NA	1.8	virginica

data=pd.read_csv("test4.csv")

print(data.dropna(axis=1)) #这一列如果有缺失,删除这一列

print(data.dropna(axis=0)) #这一行如果有缺失,删除这一行

print(data.fillna(0)) #缺失的地方补 0

print(data.fillna(method='pad')) #缺的地方用同列前一行的值填充
print(data.fillna(data.mean())) #缺的地方用同列的均值填充
print(data.fillna(data.median())) #缺的地方用同列的中位数填充

print(data.fillna(data.mode())) #缺的地方用同列的众数填充



Do you finish it in 5 minutes?

	Α	В	С	D
TV		radio	newspaper	sales
	230.1	37.8	69.2	22.1
	44.5			10.4
	151.5	41.3	58.5	18.5
	180.8			12.9
	8.7	48.9	75	7.2





No

#这一列如果有缺失,删除这一列

#这一行如果有缺失,删除这一行

#缺失的地方补 0

pd='pad')) #缺的地方用同列前一行的值填充

nean())) #缺的地方用同列的均值填充

nedian())) #缺的地方用同列的中位数填充

#缺的地方用同列的众数填充

Review (Assignments & Data Cleaning)



3 对【2】题数据采用回归方法填补。

```
import numpy as np
import pandas
from skleam experimental import enable iterative imputer
from skleary impute import Iterative Imputer
data = pandas.read csv("test4.csv")
print(data)
imp = IterativeImputer(max iter=10, random state=0)#回归填补法
imp.fit(data)
v = np.round(imp.transform(data))
print(v)
```

Review (Assignments & Data Cleaning)



- 3 对【2】题数据采用回归方法填补。
- [3] 对【2】题数据采用回归方法填补。

采用一元线性回归,由于只有 Sepal.Length、Sepal.Width、Petal.Length 三列存在缺失值,将这三列单独拿出来与最后一列各自组成一个 csv 文件,由于最后一列是英文名将其按种类分别映射成数字 0,1,2,也就是 setosa 对应 0、versicolor 对应 0、virginica 对应 1。再分三次回归填补。相当于构建三条回归直线。整体思路都是先删除有缺失值的行,然后再构建线性回归模型,再预测缺失值。



(Assignments & Data Cleaning Review



3 对【2】题数据采用回归方法填补。

#Sepal.Length 列的填补

data1=pd.read_csv("test4-1.csv")

data1=data1.dropna(axis=0)

regr1=linear_model(LinearRegression()

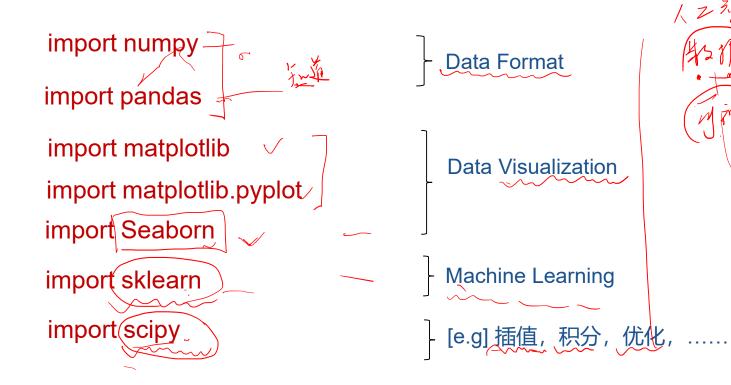
regr1.fit(data1['species'].values.reshape(-1,1),data1['Sepal.Length'])

print(regr1.predict([[2]]))

Sepal.Width	Petal.Length	Petal.Width	Species
3.5	1.4	0.2	setosa
3	NA	0.2	setosa
3.2	NA _	0.2	setosa
3.1	1.5	0.2	setosa
3.8	NA	0.2	setosa
3.2	1.4	0.2	setosa
NA	1.5	0.2	setosa
33	1.4	0.2	setosa
3.2	4.7	1.4	versicolor
3.2	NA	1.5	versicolor
3.3	6	2.5	virginica
2.7	5.1	1.9	virginica
3	NA	2.1	virginica
2.9	5.6	1.8	virginica
/ 3	NA	1.8	virginica
	NA 3.2 3.1 3.8 3.2 NA 3.3 3.2 3.3 2.7 3.3 2.7	3.5 1.4 3 NA 3.2 NA 3.1 1.5 3.8 NA 3.2 1.4 1.5 3.3 1.4 3.2 4.7 3.2 NA 3.3 6 2.7 5.1 3 NA	3.5

Data Analysis & Python

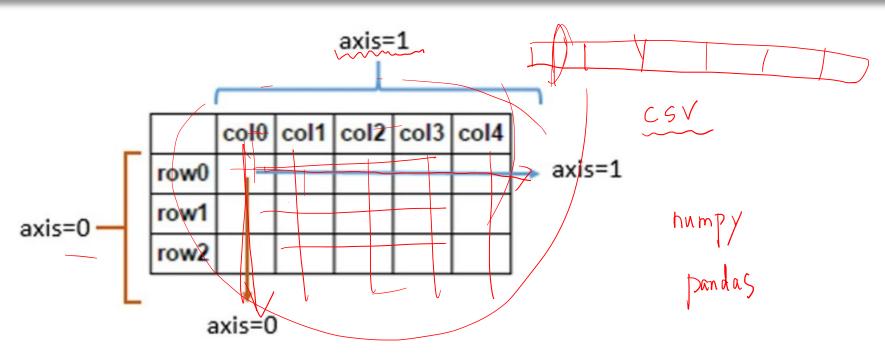




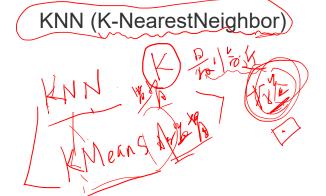
人工程能 Python

Data Analysis & Python

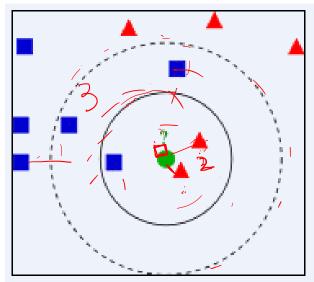


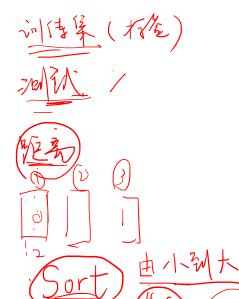






Reading KNN code





欧式距离:
$$d(x,y) = \sqrt{\sum_{k=1}^{n} (x_k - y_{jk})^2}$$
,

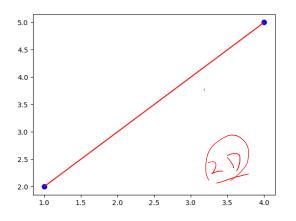
曼哈顿距离:
$$d(x,y) = \sqrt{\sum_{k=1}^{n} |x_k - y_k|}$$

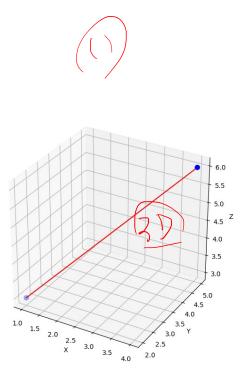


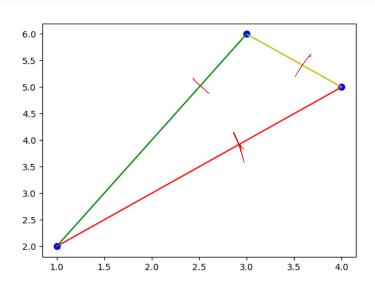


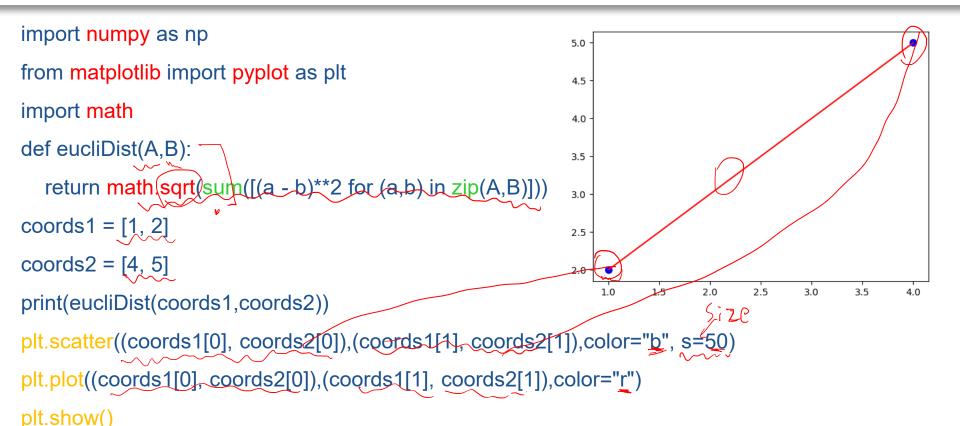
How do you understand KNN and Kmeans?

- A KNN = = K'means
- B KNN! = K'means

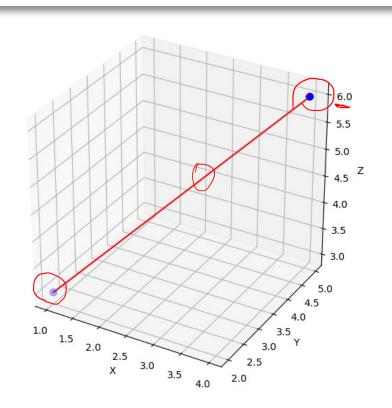


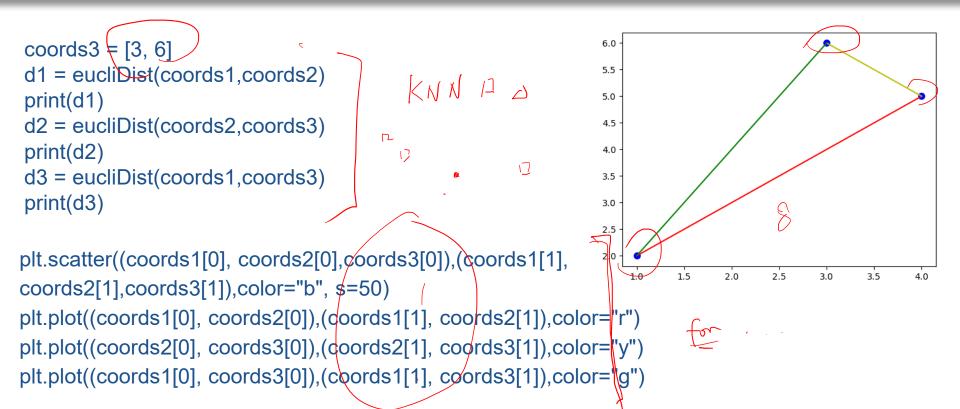






```
from mpl toolkits.mplot3d import Axes3D
from mpl_toolkits.mplot3d import proj3d
coords1 = [1, 2, 3]
coords2 = [4, 5, 6]
fig = plt.figure(figsize = (7,7))
ax = fig.add_subplot(111, projection='3d')
ax.scatter((coords1[0], coords2[0]),(coords1[1],
coords2[1]),(coords1[2], coords2[2]),color="b", s=50)
ax.plot((coords1[0], coords2[0]),(coords1[1],
coords2[1]),(coords1[2], coords2[2]),color="r")
```









贵在坚持!