Ch1_(1)

February 25, 2025

1

[1]:
```
import pandas as pd
import numpy as np
missdict = {'f1': [1, 2, 3, 4, 5, 6, 7, 8, 9, 10],
            'f2': [10., None, 20., 30., None, 50., 60., 70., 80., 90.],
            'f3': ['A','A', 'A', 'A','B','B','B','B','C','C']}
missdata = pd.DataFrame( missdict )
missdata.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10 entries, 0 to 9
Data columns (total 3 columns):
 #   Column  Non-Null Count  Dtype
---  ------  --------------  -----
 0   f1      10 non-null     int64
 1   f2      8 non-null      float64
 2   f3      10 non-null     object
dtypes: float64(1), int64(1), object(1)
memory usage: 368.0+ bytes
```

[2]:
```
missdata.isna().mean()
```

[2]:
```
f1    0.0
f2    0.2
f3    0.0
dtype: float64
```

[3]:
```
tmpdata1 = missdata.dropna()
tmpdata1
```

[3]:

	f1	f2	f3
0	1	10.0	A
2	3	20.0	A
3	4	30.0	A
5	6	50.0	B
6	7	60.0	B

```
7   8  70.0  B
8   9  80.0  C
9  10  90.0  C
```

[4]:
```
tmpdata2 = missdata.dropna( subset=['f3'] )
tmpdata2
```

[4]:
```
   f1    f2 f3
0   1  10.0  A
1   2   NaN  A
2   3  20.0  A
3   4  30.0  A
4   5   NaN  B
5   6  50.0  B
6   7  60.0  B
7   8  70.0  B
8   9  80.0  C
9  10  90.0  C
```

[5]:
```
numdata = missdata.select_dtypes(include=['int64', 'float64'])
tmpdata3 = numdata.fillna( -999, inplace=False )
tmpdata3.describe()
```

[5]:
```
              f1          f2
count  10.00000   10.000000
mean    5.50000 -158.800000
std     3.02765  443.562297
min     1.00000 -999.000000
25%     3.25000   12.500000
50%     5.50000   40.000000
75%     7.75000   67.500000
max    10.00000   90.000000
```

[6]:
```
numdata.mean()
```

[6]:
```
f1     5.50
f2    51.25
dtype: float64
```

[7]:
```
tmpdata4 = numdata.fillna( numdata.mean(), inplace=False  )
tmpdata4
```

[7]:
```
   f1     f2
0   1  10.00
1   2  51.25
2   3  20.00
3   4  30.00
```

```
4   5  51.25
5   6  50.00
6   7  60.00
7   8  70.00
8   9  80.00
9  10  90.00
```

[8]:
```
missdata.groupby('f3')['f2'].mean()
```

[8]:
```
f3
A    20.0
B    60.0
C    85.0
Name: f2, dtype: float64
```

[9]:
```
missdata.groupby('f3')['f2'].transform('mean')
```

[9]:
```
0    20.0
1    20.0
2    20.0
3    20.0
4    60.0
5    60.0
6    60.0
7    60.0
8    85.0
9    85.0
Name: f2, dtype: float64
```

[10]:
```
tmpdata5 = numdata.copy()
tmpdata5['f2'].fillna( missdata.groupby('f3')['f2'].transform('mean'),␣
 ↪inplace=True)
tmpdata5
```

[10]:
```
    f1    f2
0   1  10.0
1   2  20.0
2   3  20.0
3   4  30.0
4   5  60.0
5   6  50.0
6   7  60.0
7   8  70.0
8   9  80.0
9  10  90.0
```

[11]:
```
missdata_tr = missdata.dropna()
x_tr = missdata_tr[['f1']]
y_tr = missdata_tr['f2']

from sklearn.linear_model import LinearRegression
model = LinearRegression()
model.fit( x_tr, y_tr )

missdata_ts = missdata [ missdata.isnull().any(axis=1) ]
x_ts = missdata_ts[['f1']]

predicted_values = model.predict( x_ts )
tmpdata6 = missdata.copy()
tmpdata6.loc[ tmpdata6['f2'].isnull(), 'f2'] = predicted_values
tmpdata6
```

[11]:

	f1	f2	f3
0	1	10.000000	A
1	2	14.191176	A
2	3	20.000000	A
3	4	30.000000	A
4	5	41.985294	B
5	6	50.000000	B
6	7	60.000000	B
7	8	70.000000	B
8	9	80.000000	C
9	10	90.000000	C

[12]:
```
missdata_num = missdata.copy()
missdata_num['f3']=missdata_num['f3'].map({'A':1, 'B':2, 'C':3})
```

[13]:
```
missdata_num
```

[13]:

	f1	f2	f3
0	1	10.0	1
1	2	NaN	1
2	3	20.0	1
3	4	30.0	1
4	5	NaN	2
5	6	50.0	2
6	7	60.0	2
7	8	70.0	2
8	9	80.0	3
9	10	90.0	3

[14]:
```
from sklearn.impute import KNNImputer
imputer = KNNImputer(n_neighbors=2)
```

```
tmpdata7 = imputer.fit_transform(missdata_num)
```

[15]:
```
pd.DataFrame( tmpdata7 )
```

[15]:

	0	1	2
0	1.0	10.0	1.0
1	2.0	15.0	1.0
2	3.0	20.0	1.0
3	4.0	30.0	1.0
4	5.0	40.0	2.0
5	6.0	50.0	2.0
6	7.0	60.0	2.0
7	8.0	70.0	2.0
8	9.0	80.0	3.0
9	10.0	90.0	3.0

2

[16]:
```
outdict = {'A': [10, 0.02, 0.3, 40, 50, 60, 712, 80, 90, 1003],
        'B': [0.05, 0.00015, 25, 35, 45, 205, 65, 75, 85, 3905]}
outdata = pd.DataFrame( outdict )

Q1 = outdata.quantile(0.25)
Q3 = outdata.quantile(0.75)
IQR = Q3 - Q1
lower_bound = Q1 - 1.5 * IQR
upper_bound = Q3 + 1.5 * IQR

((outdata < lower_bound) | (outdata > upper_bound))
```

[16]:

	A	B
0	False	False
1	False	False
2	False	False
3	False	False
4	False	False
5	False	True
6	True	False
7	False	False
8	False	False
9	True	True

[17]:
```
outliers = ((outdata < lower_bound) | (outdata > upper_bound)).any(axis=1)
outliersdata = outdata[ outliers ]
outliersdata
```

[17]:
```
         A       B
5    60.0   205.0
6   712.0    65.0
9  1003.0  3905.0
```

[18]:
```
standardizeddata = (outdata - outdata.mean()) / outdata.std()
standardizeddata
```

[18]:
```
          A         B
0 -0.552206 -0.364647
1 -0.580536 -0.364688
2 -0.579741 -0.344154
3 -0.467047 -0.335940
4 -0.438661 -0.327727
5 -0.410274 -0.196309
6  1.440519 -0.311300
7 -0.353501 -0.303086
8 -0.325115 -0.294872
9  2.266563  2.842723
```

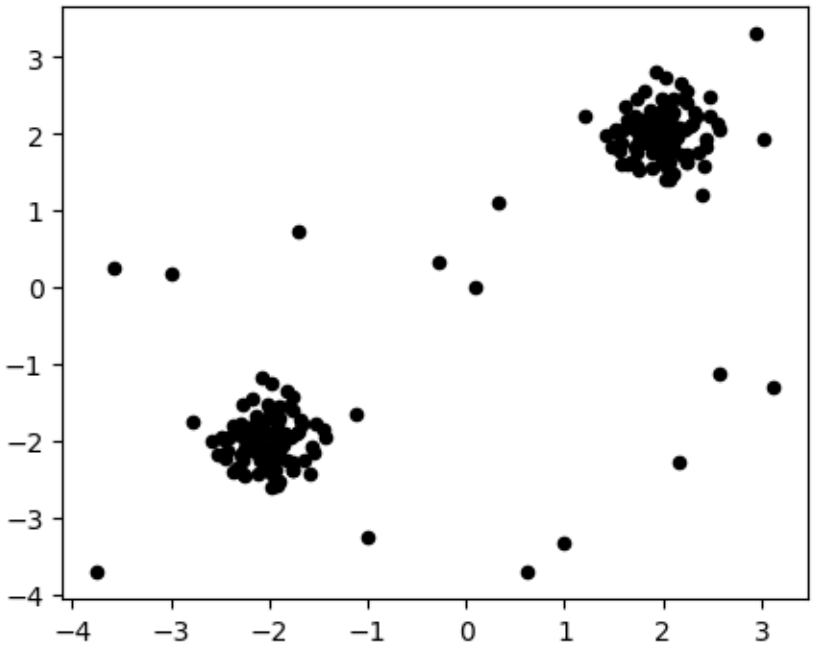
[19]:
```
outliers2 = ((standardizeddata < -3) | (standardizeddata > 3)).any(axis=1)
outliersdata2 = outdata[ outliers2 ]
outliersdata2
```

[19]:
```
Empty DataFrame
Columns: [A, B]
Index: []
```

[20]:
```
import matplotlib.pyplot as plt
np.random.seed(42)
X_inliers = 0.3 * np.random.randn(100, 2)
X_outliers = np.random.uniform(low=-4, high=4, size=(20, 2))
X = np.r_[X_inliers + 2, X_inliers - 2, X_outliers]

plt.figure(figsize=(5, 4))
plt.scatter(X[:, 0], X[:, 1], color='k', s=20)
```

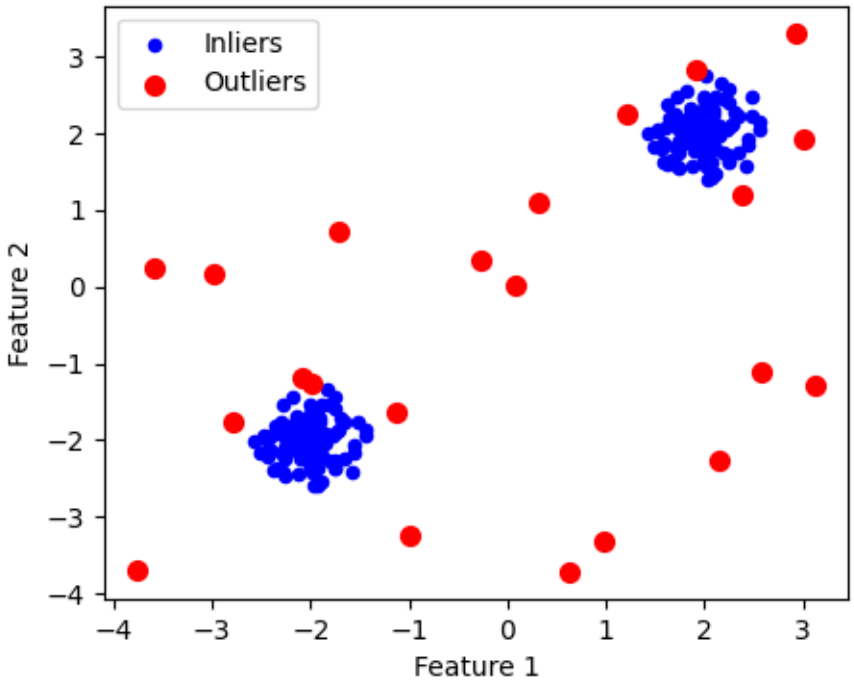
[20]:
```
<matplotlib.collections.PathCollection at 0x7b9fd1eb3790>
```



[21]:
```
from sklearn.neighbors import LocalOutlierFactor
clf = LocalOutlierFactor(n_neighbors=20, contamination=0.1)
y_pred = clf.fit_predict(X) # 1: inlier, -1: outlier
outlier_mask = y_pred == -1

plt.figure(figsize=(5, 4))
plt.scatter(X[:, 0], X[:, 1], color='b', s=20, label='Inliers')
plt.scatter(X[outlier_mask, 0], X[outlier_mask, 1], color='r', s=50,␣
 ↪label='Outliers')
plt.xlabel("Feature 1")
plt.ylabel("Feature 2")
plt.legend()
```

[21]:
```
<matplotlib.legend.Legend at 0x7b9fd1df78e0>
```

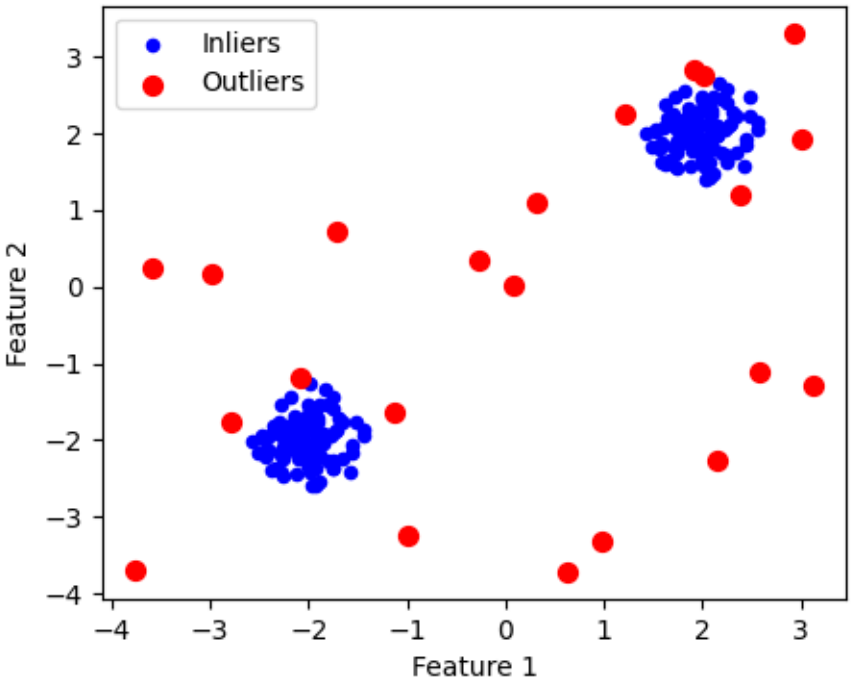


[22]:
```
from sklearn.ensemble import IsolationForest
clf2 = IsolationForest(contamination=0.1)
# contamination :
# n_estimators :      (defalut  100)
# max_features :                (default  1)

clf2.fit( X )
y_pred2 = clf2.predict( X ) # 1: inlier, -1: outlier
outlier_mask2 = y_pred2 == -1

plt.figure(figsize=(5, 4))
plt.scatter(X[:, 0], X[:, 1], color='b', s=20, label='Inliers')
plt.scatter(X[outlier_mask2, 0], X[outlier_mask2, 1], color='r', s=50,␣
 ↪label='Outliers')
plt.xlabel("Feature 1")
plt.ylabel("Feature 2")
plt.legend()
```

[22]:
```
<matplotlib.legend.Legend at 0x7b9fd18bb280>
```


Inliers
Outliers
Feature 2
Feature 1
3
2
1
0
−1
−2
−3
−4
−4
−3
−2
−1
0
1
2
3