

# visualisation

September 19, 2024

## 1 Lab 1

```
[1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

### 1.1 Attribute Information

1. lettr capital letter (26 values from A to Z)
2. x-box horizontal position of box (integer)
3. y-box vertical position of box (integer)
4. width width of box (integer)
5. high height of box (integer)
6. onpix total # on pixels (integer)
7. x-bar mean x of on pixels in box (integer)
8. y-bar mean y of on pixels in box (integer)
9. x2bar mean x variance (integer)
10. y2bar mean y variance (integer)
11. xybar mean x y correlation (integer)
12. x2ybr mean of  $x * x * y$  (integer)
13. xy2br mean of  $x * y * y$  (integer)
14. x-ege mean edge count left to right (integer)
15. xegvy correlation of x-ege with y (integer)
16. y-ege mean edge count bottom to top (integer)
17. yegvx correlation of y-ege with x (integer)

### 1.2 Load the data

```
[2]: df = pd.read_csv('letter+recognition/letter-recognition.data')
```

```
[3]: df.head()
```

```
[3]:  lettr  x-box  y-box  width  high  onpix  x-bar  y-bar  x2bar  y2bar  xybar  \
0      T      2      8      3      5      1      8     13      0      6      6
1      I      5     12      3      7      2     10      5      5      4     13
2      D      4     11      6      8      6     10      6      2      6     10
3      N      7     11      6      6      3      5      9      4      6      4
```

4	G	2	1	3	1	1	8	6	6	6	6
---	---	---	---	---	---	---	---	---	---	---	---

	x2ybr	xy2br	x-ege	xegvy	y-ege	yegvx
0	10	8	0	8	0	8
1	3	9	2	8	4	10
2	3	7	3	7	3	9
3	4	10	6	10	2	8
4	5	9	1	7	5	10

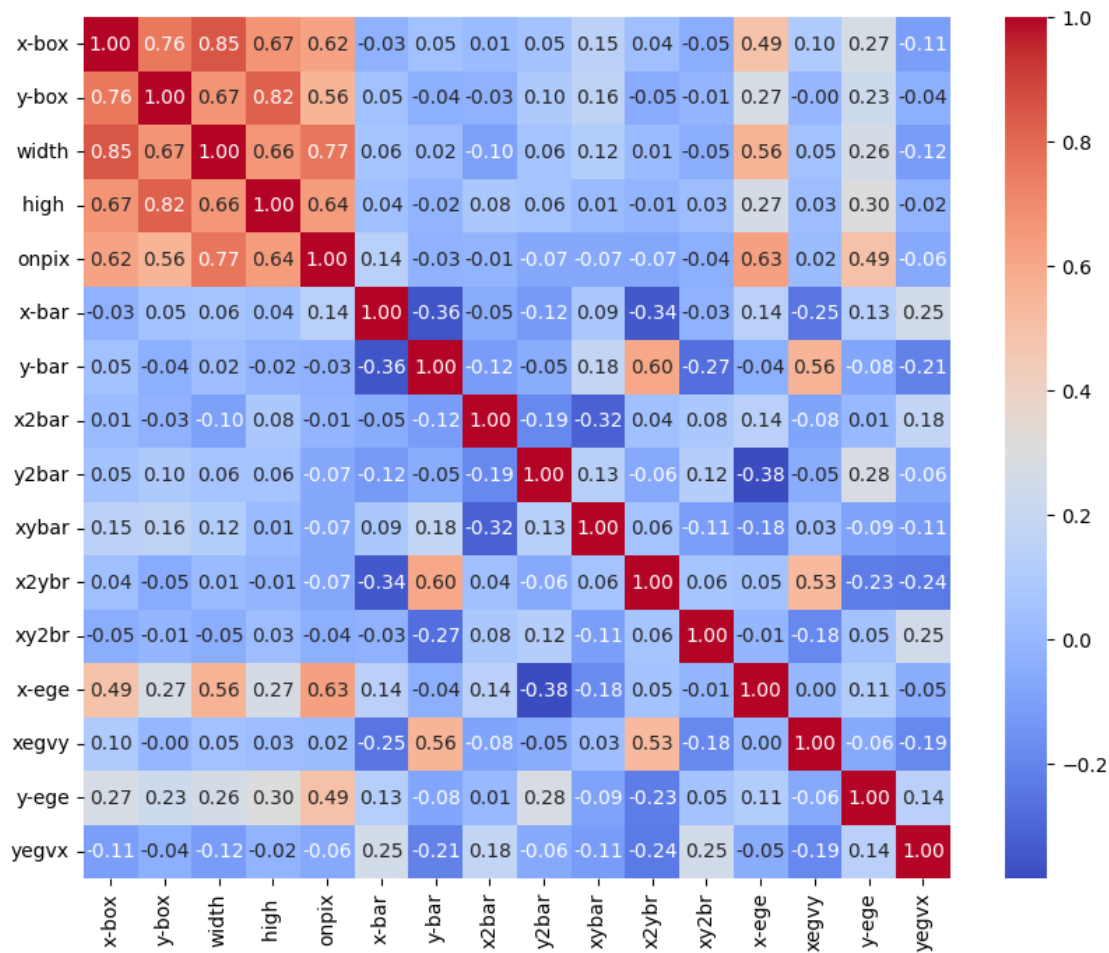
```
[4]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 20000 entries, 0 to 19999
Data columns (total 17 columns):
#   Column      Non-Null Count  Dtype
---  -
0   lettr       20000 non-null   object
1   x-box       20000 non-null   int64
2   y-box       20000 non-null   int64
3   width       20000 non-null   int64
4   high        20000 non-null   int64
5   onpix       20000 non-null   int64
6   x-bar       20000 non-null   int64
7   y-bar       20000 non-null   int64
8   x2bar       20000 non-null   int64
9   y2bar       20000 non-null   int64
10  xybar       20000 non-null   int64
11  x2ybr       20000 non-null   int64
12  xy2br       20000 non-null   int64
13  x-ege       20000 non-null   int64
14  xegvy       20000 non-null   int64
15  y-ege       20000 non-null   int64
16  yegvx       20000 non-null   int64
dtypes: int64(16), object(1)
memory usage: 2.6+ MB
```

```
[5]: target = "lettr"
```

### 1.3 Pairwise correlation of columns

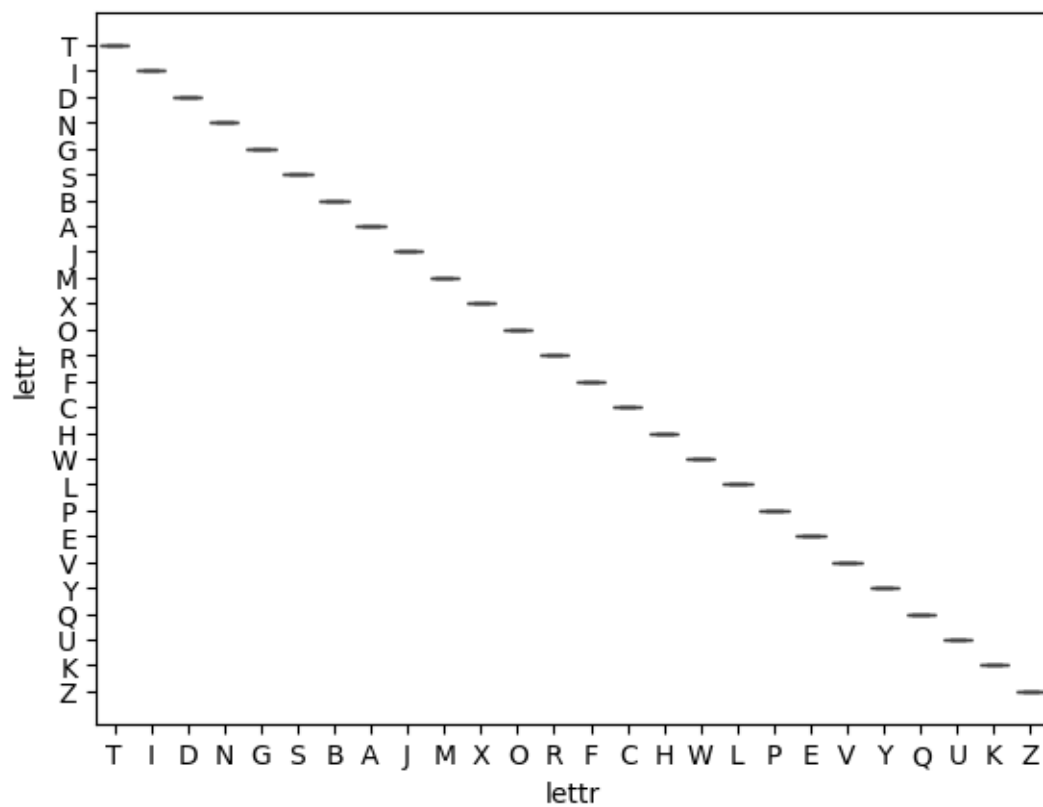
```
[6]: corr = df.corr()
fig, ax = plt.subplots(figsize=(10, 8))
sns.heatmap(corr, cmap='coolwarm', annot=True, fmt=".2f")
plt.show()
```

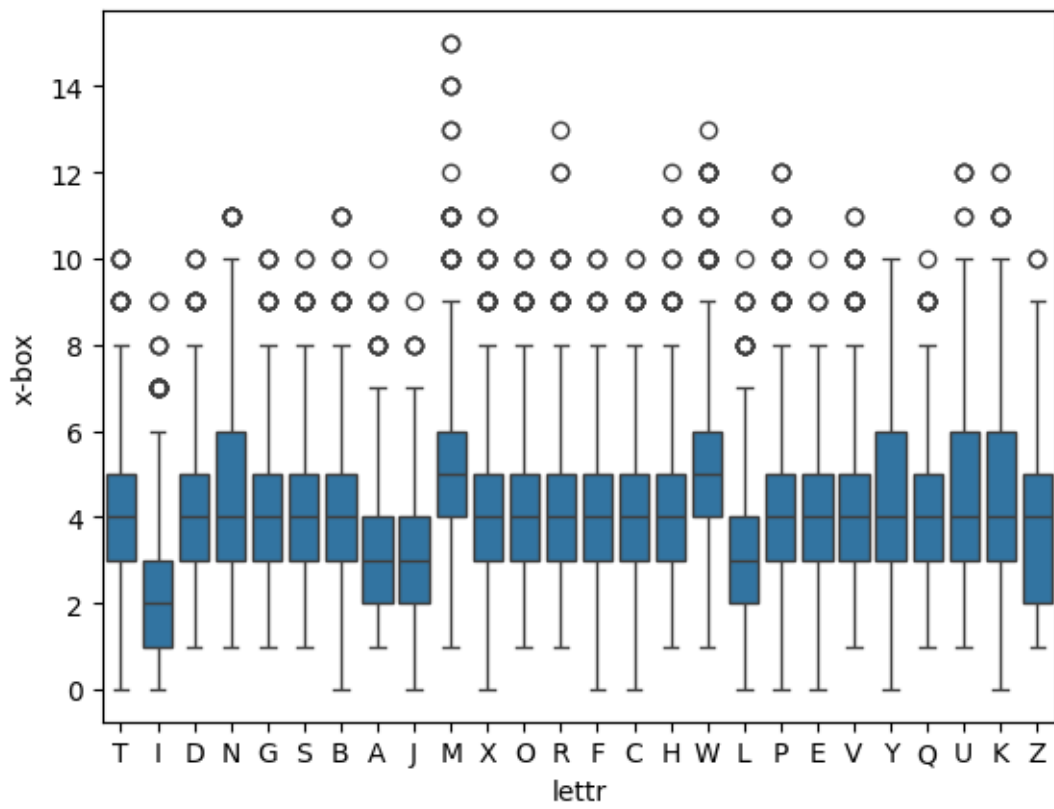


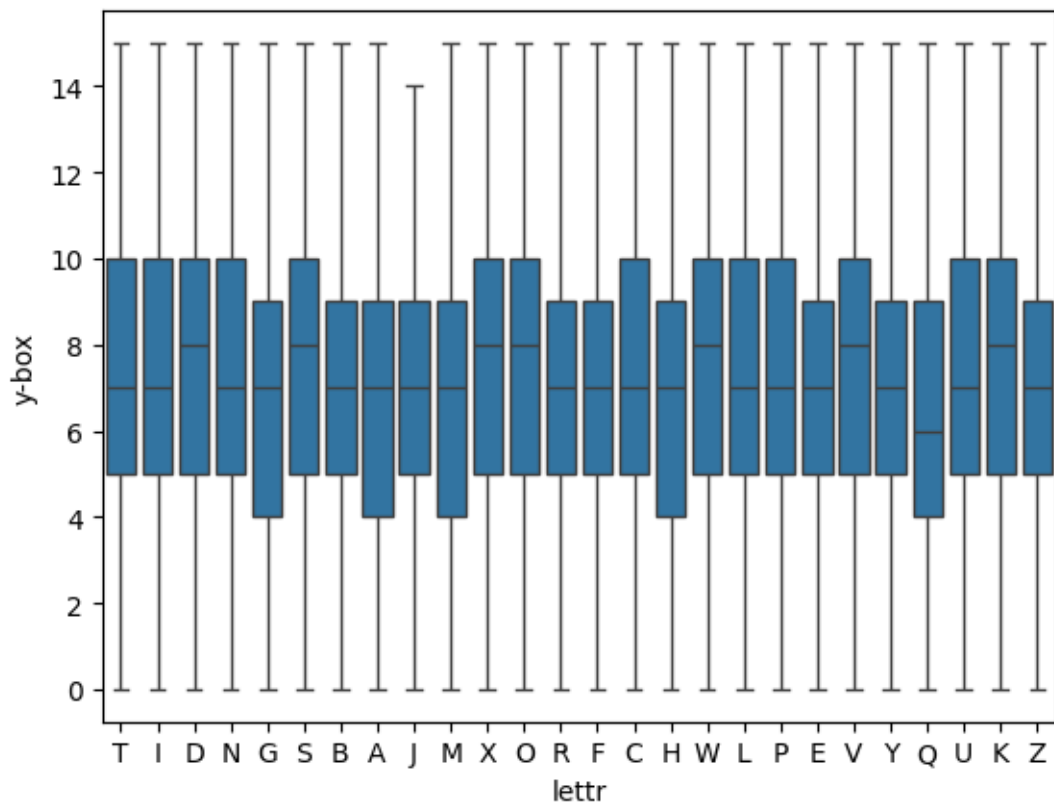
## 1.4 Boxplots

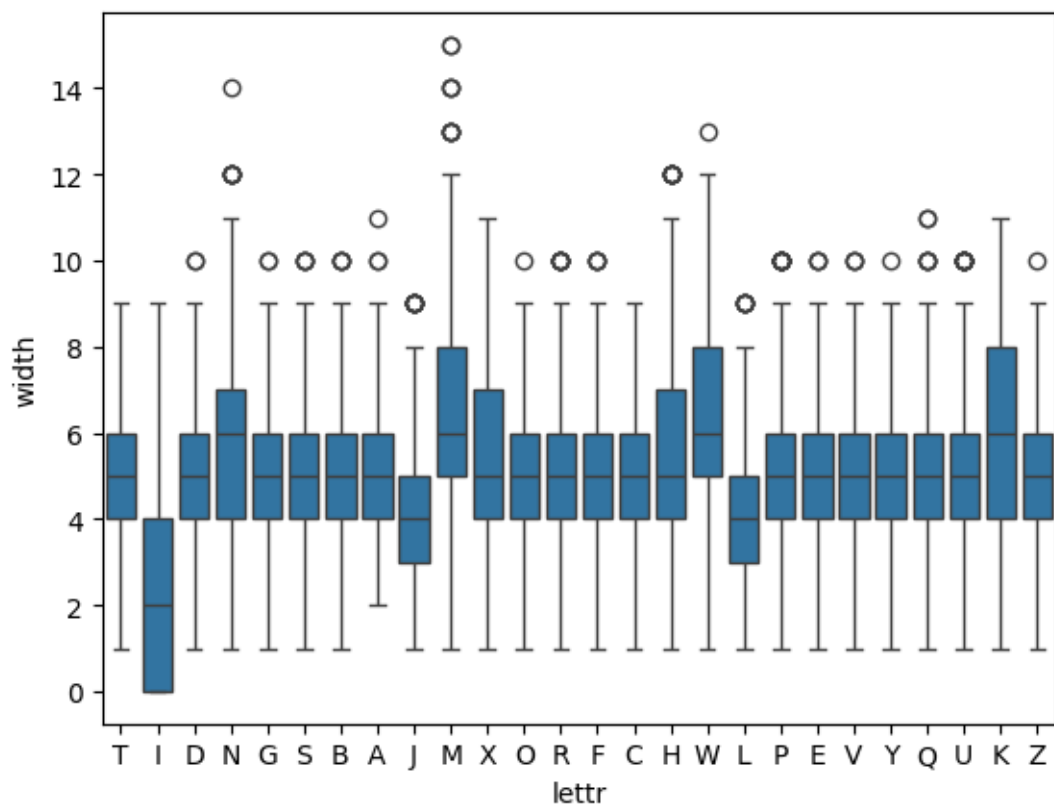
```
[7]: feature_names = df.columns

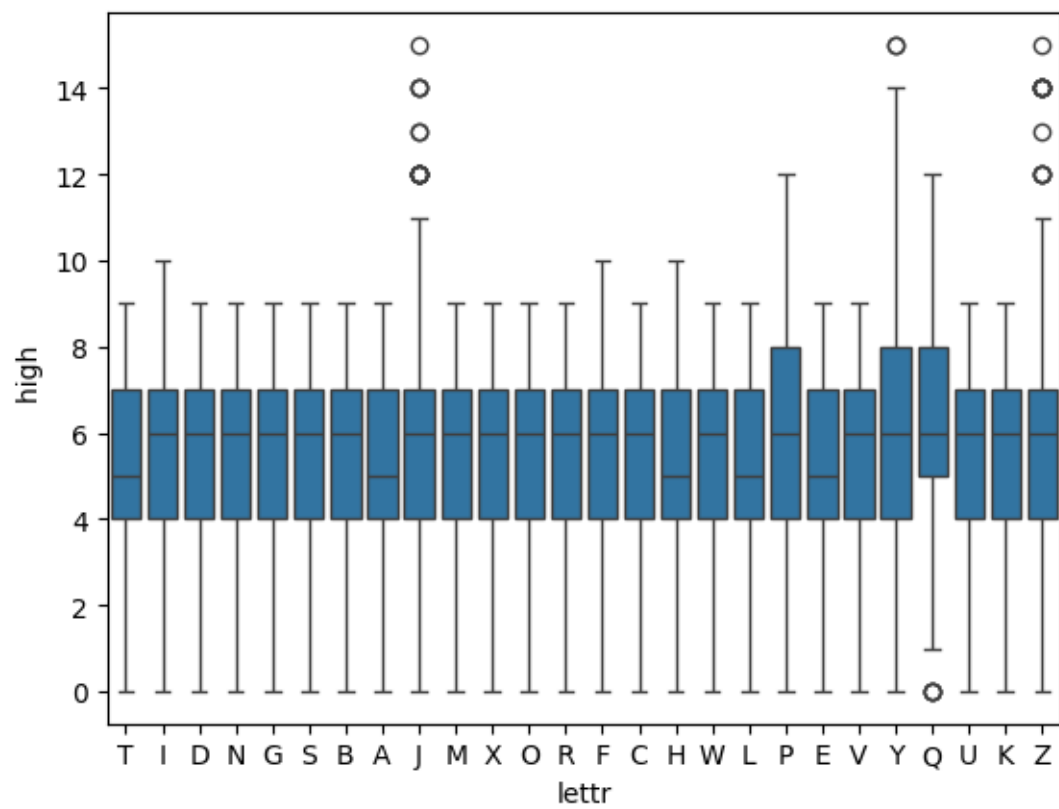
for i in range(len(feature_names)):
    figure = plt.figure()
    ax = sns.boxplot(x=target, y=feature_names[i], data=df)
```



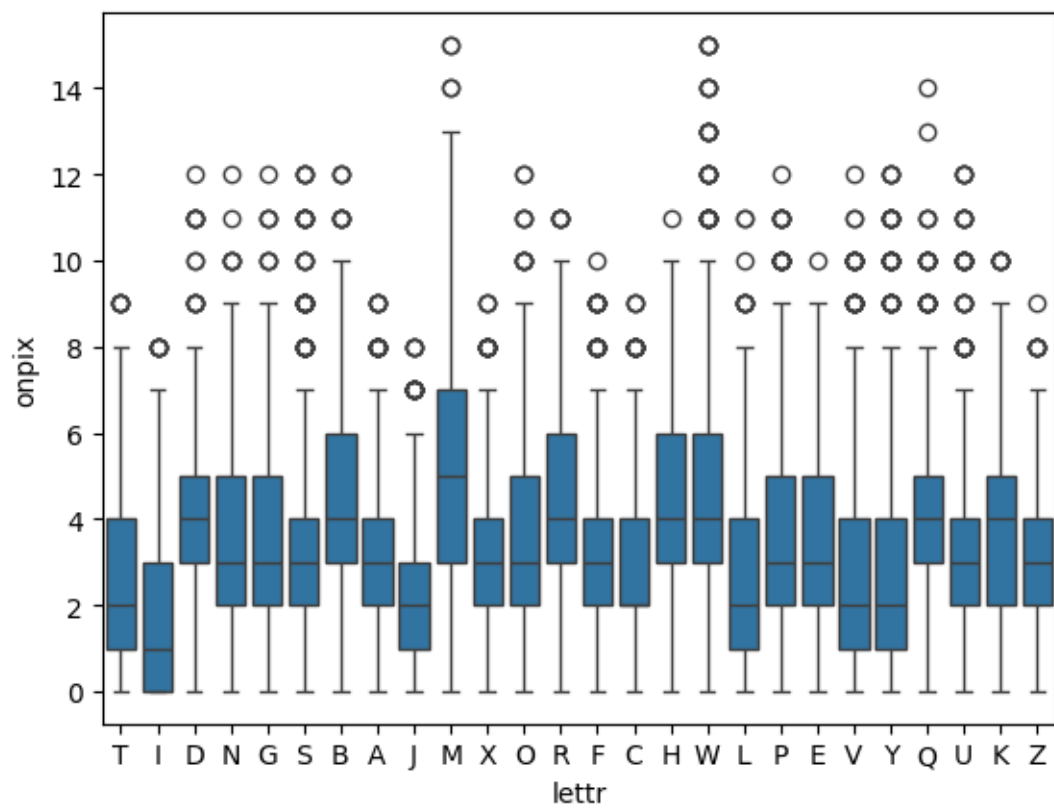


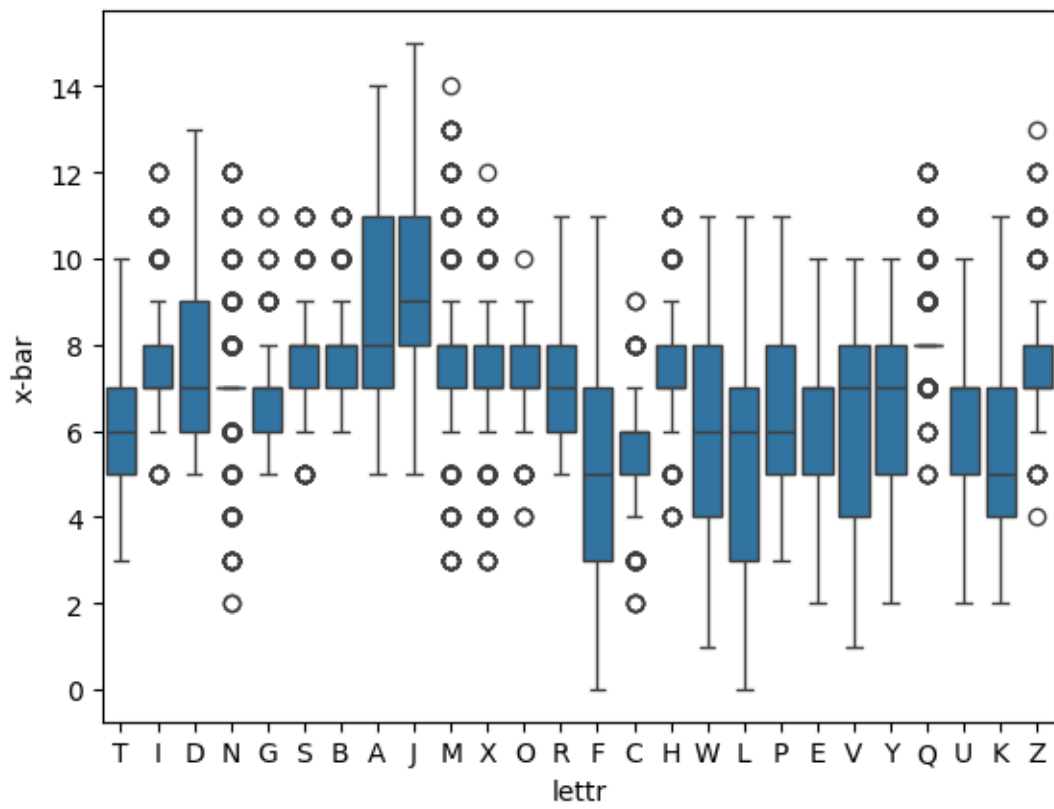


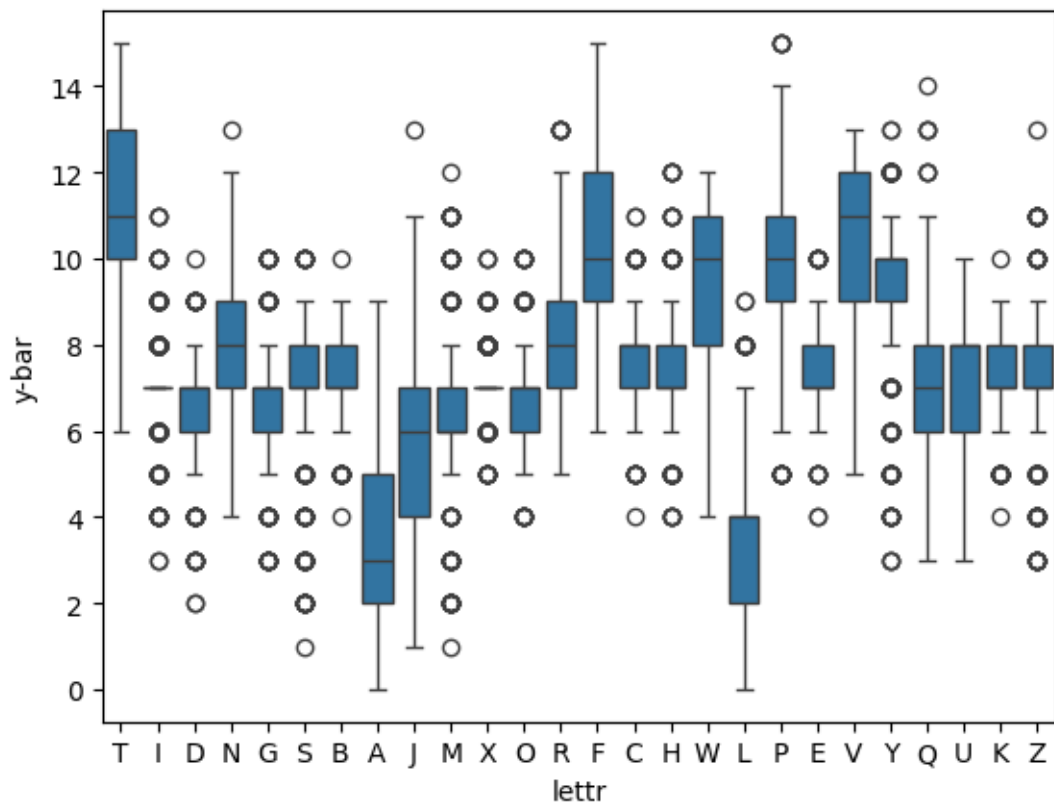


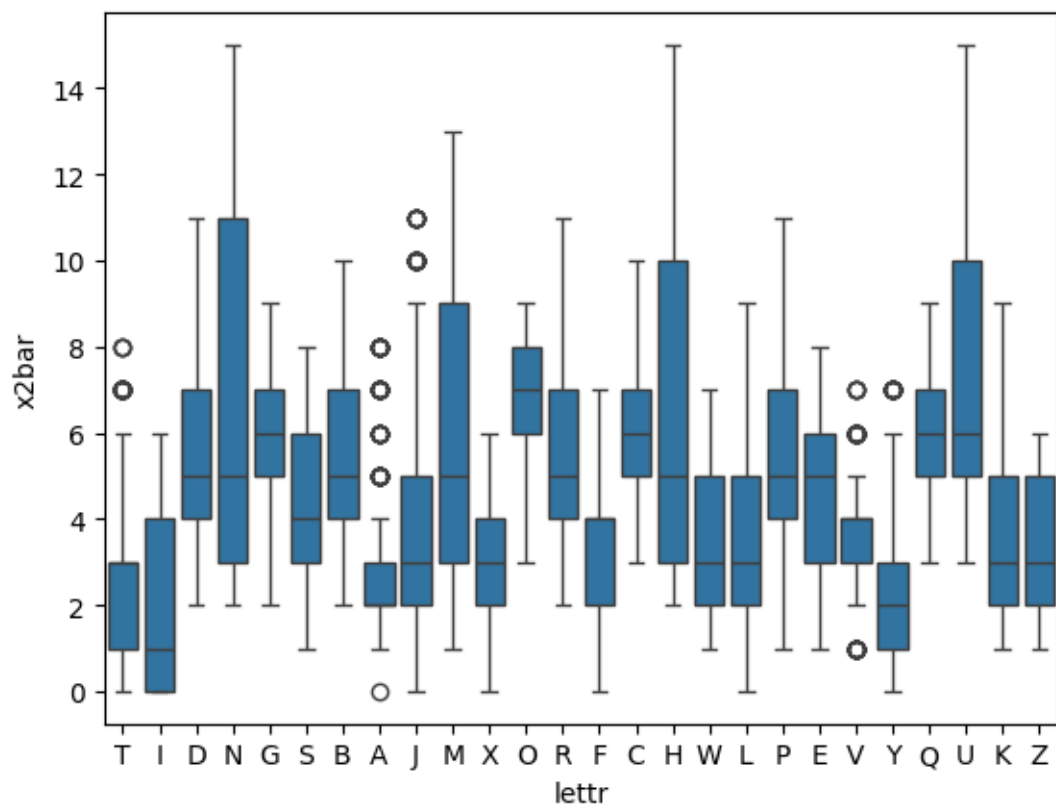


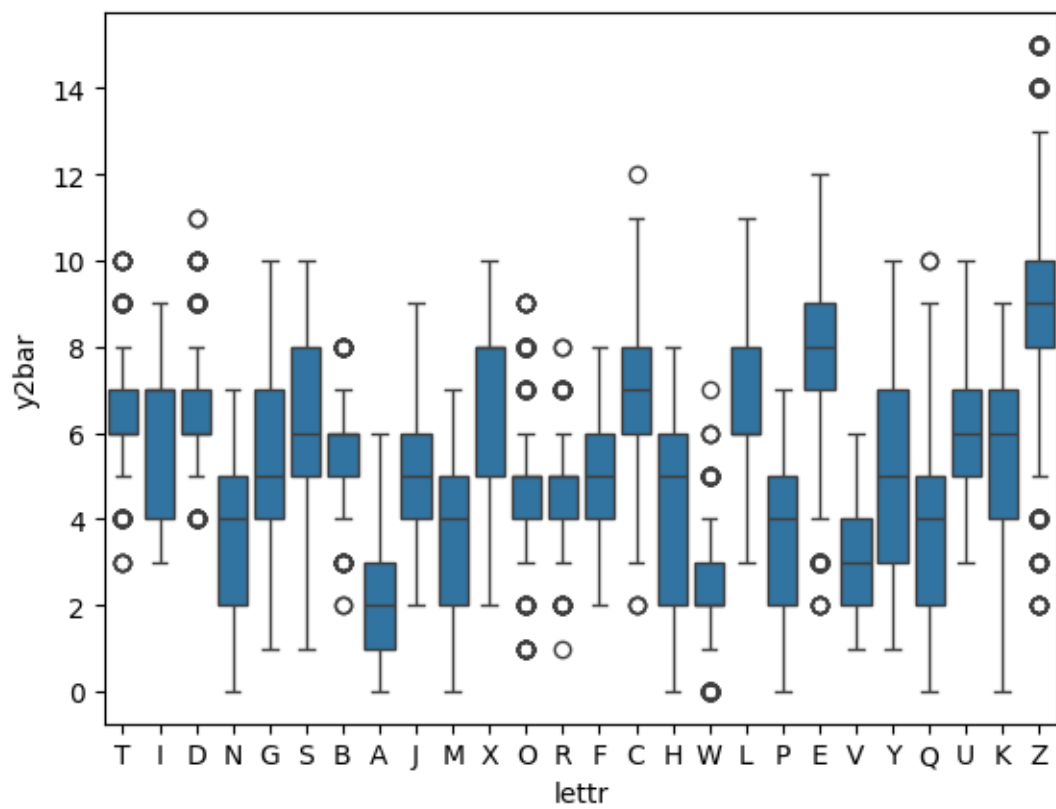


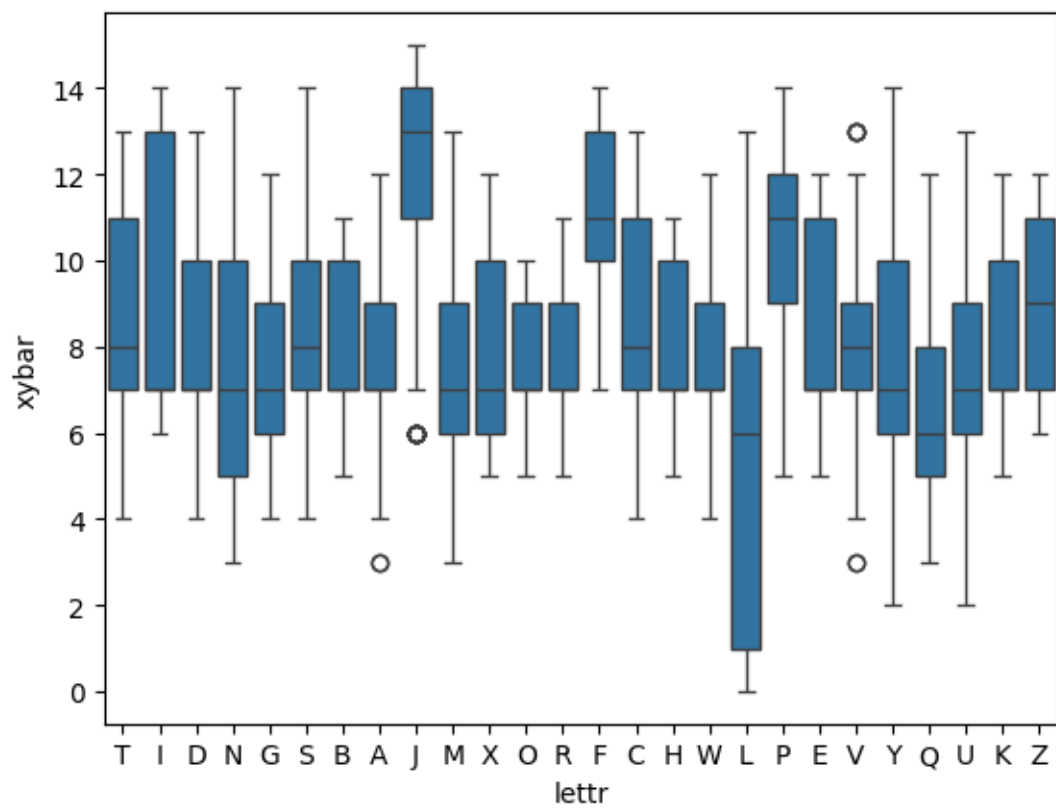


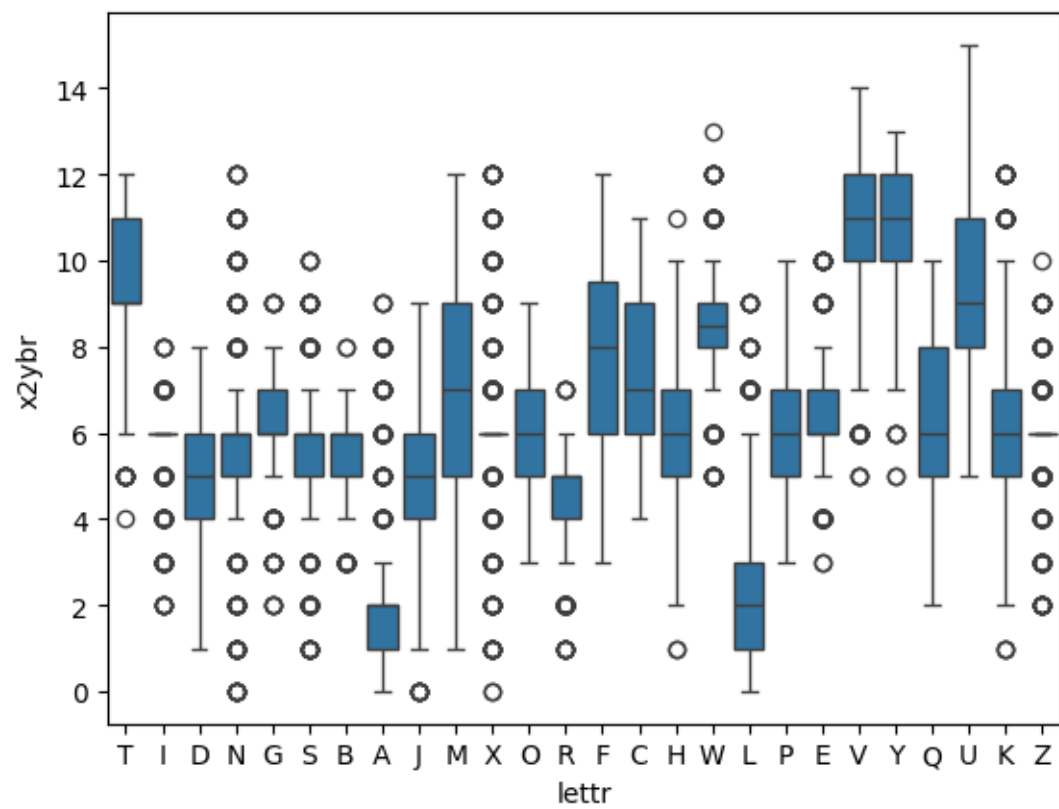


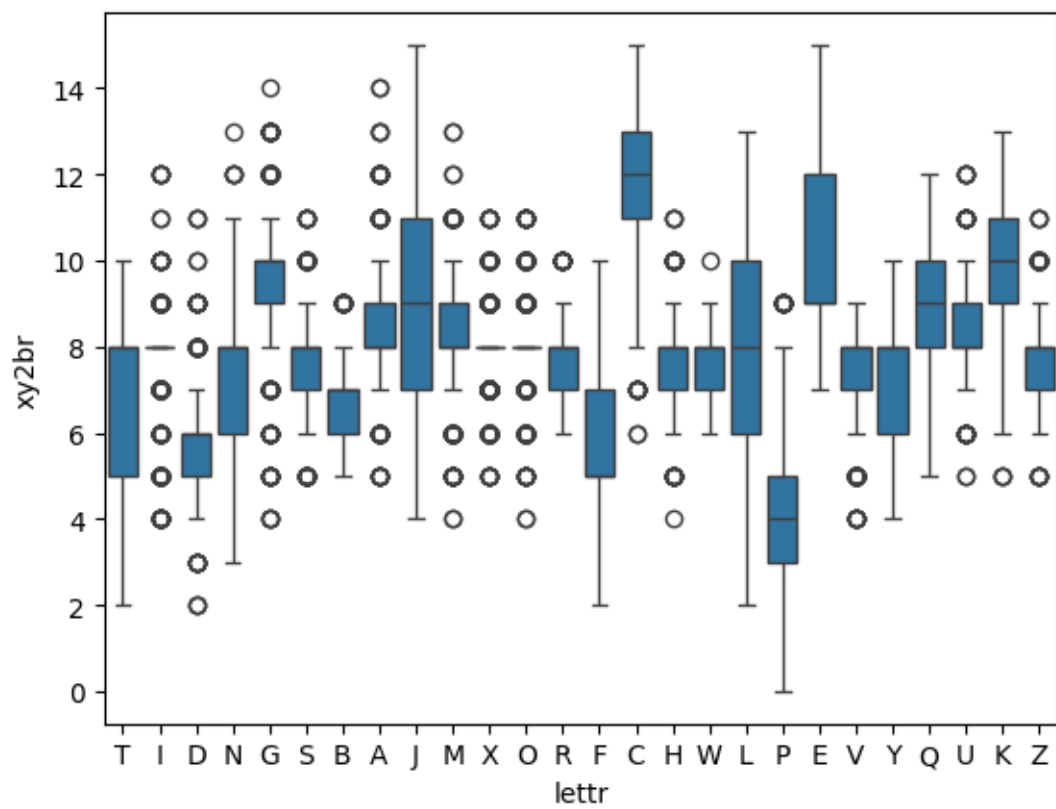




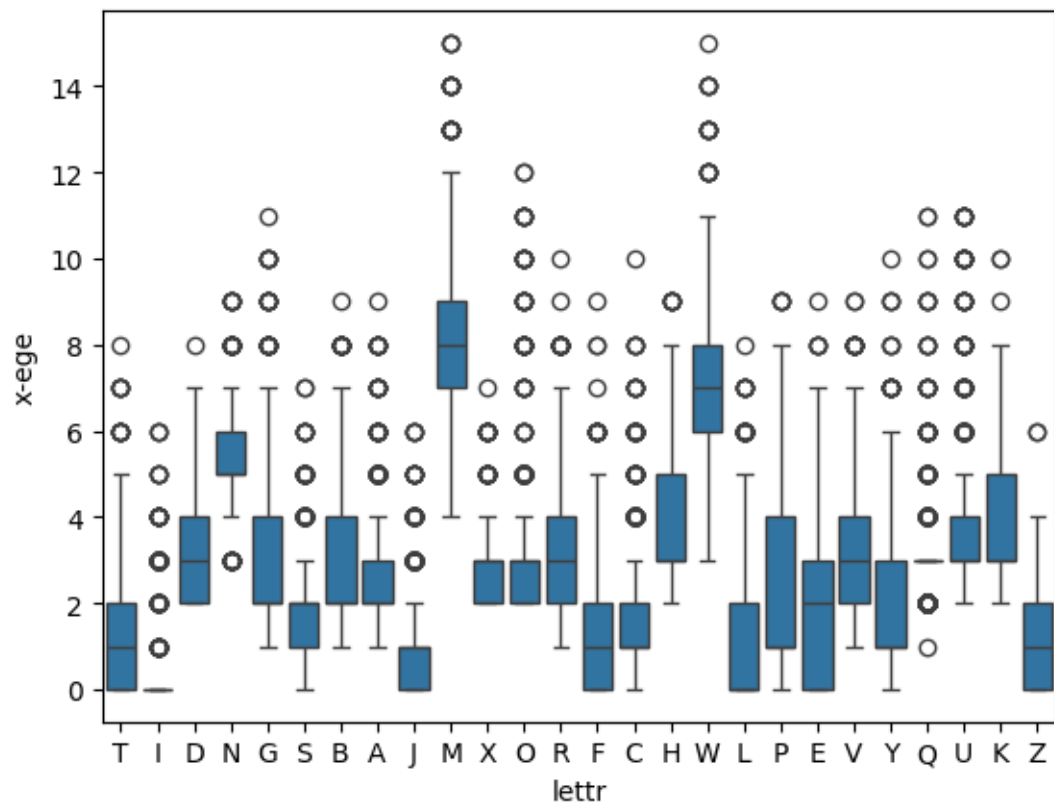


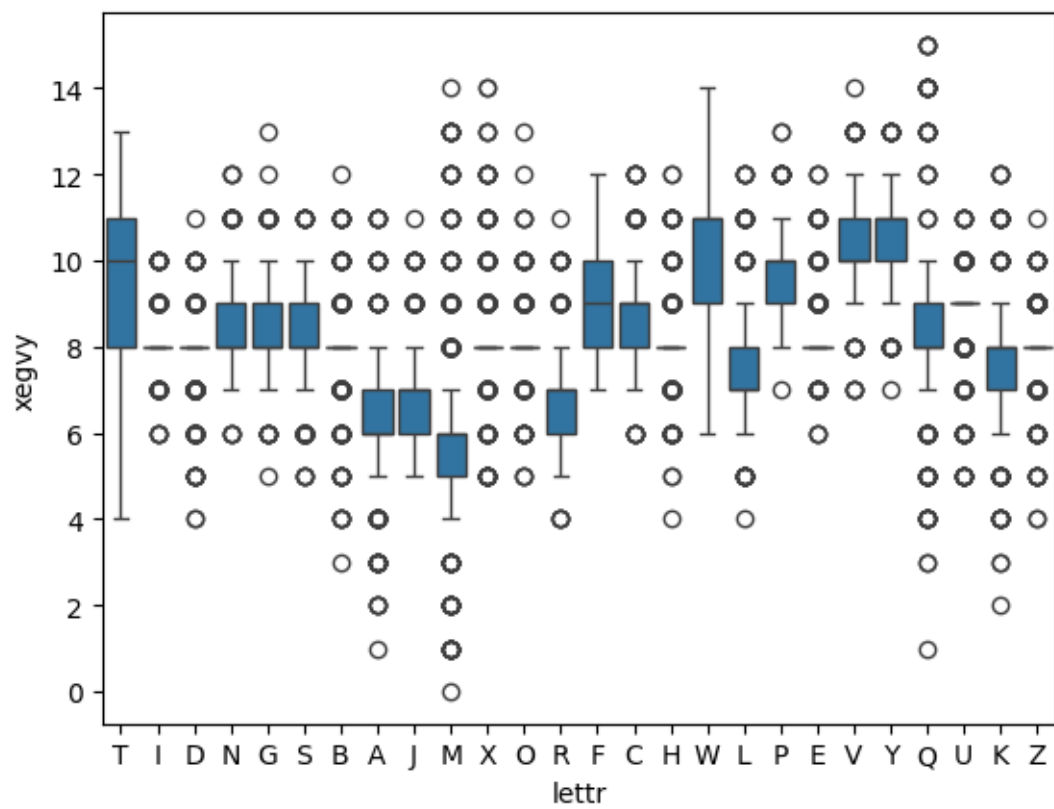


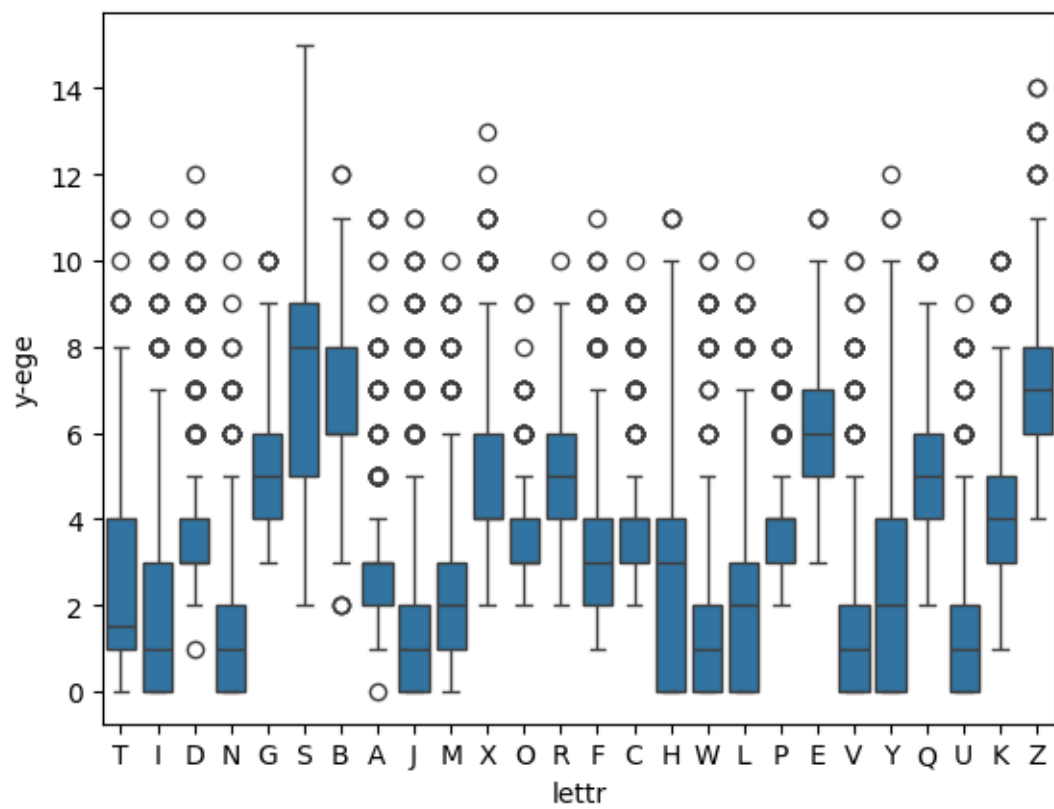


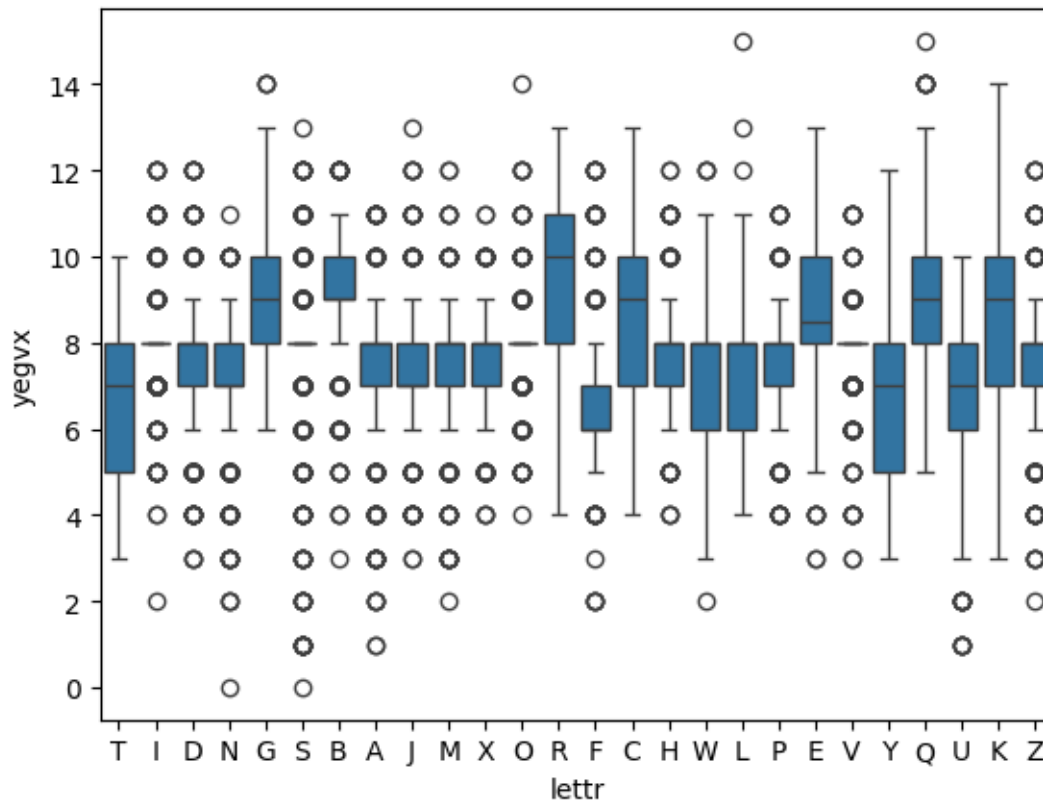










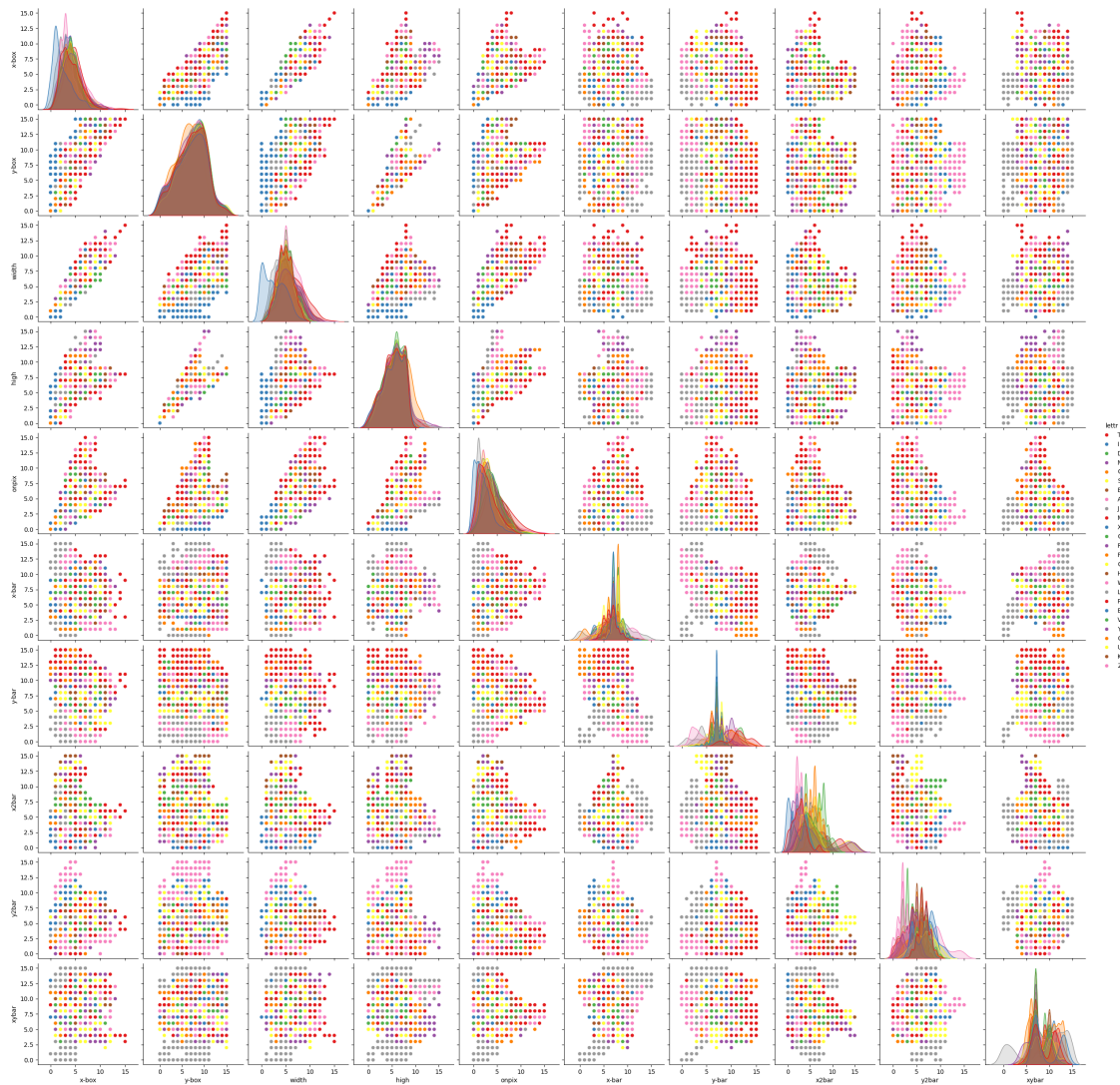


## 1.5 Pairplot

```
[8]: # part = df[["lettr","x-box","y-box","width","high"  
      ↪,"onpix","x-bar","y-bar","x2bar","y2bar","xybar","x2ybr","xy2br","x-ege","xegyvy","y-ege"],  
part = df[["lettr","x-box","y-box","width","high"  
      ↪,"onpix","x-bar","y-bar","x2bar","y2bar","xybar"]]  
sns.pairplot(part, hue=target, palette = 'Set1')
```

```
C:\Users\Lenovo\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfra8p0\LocalCache\local-packages\Python311\site-packages\seaborn\axisgrid.py:123: UserWarning: The figure layout has changed to tight
  self._figure.tight_layout(*args, **kwargs)
```

```
[8]: <seaborn.axisgrid.PairGrid at 0x138725111d0>
```



```
[9]: part2 = df[["lettr", "x2ybr", "xy2br", "x-eg", "xegvy", "y-eg", "yegvx"]]
sns.pairplot(part2, hue=target, palette = 'Set1')
```

C:\Users\Lenovo\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11\_qbz5n2kfra8p0\LocalCache\local-packages\Python311\site-packages\seaborn\axisgrid.py:123: UserWarning: The figure layout has changed to tight

```
self._figure.tight_layout(*args, **kwargs)
```

```
[9]: <seaborn.axisgrid.PairGrid at 0x1386f782b10>
```



## 1.6 Normalisation

## 1.7 Learning

```
[10]: from sklearn.model_selection import train_test_split
      from sklearn.metrics import classification_report, confusion_matrix
```

Divide data

```
[11]: train, test = train_test_split(df, test_size=0.2)

train_classes = train[target]
train = train.drop([target], axis=1)
test_classes = test[target]
test = test.drop([target], axis=1)
```

### 1.7.1 KNN

```
[12]: from sklearn.neighbors import KNeighborsClassifier
knn = KNeighborsClassifier(n_neighbors=4, weights="distance", algorithm="brute")
knn.fit(train, train_classes)
prediction = knn.predict(test)

print(classification_report(test_classes, prediction))
confusion_matrix(test_classes, prediction)
```

	precision	recall	f1-score	support
A	0.99	0.99	0.99	168
B	0.91	0.90	0.91	164
C	0.99	0.97	0.98	157
D	0.98	0.98	0.98	160
E	0.95	0.92	0.94	149
F	0.93	0.94	0.93	158
G	0.96	0.98	0.97	159
H	0.89	0.91	0.90	138
I	0.95	0.96	0.96	129
J	0.97	0.96	0.96	150
K	0.94	0.86	0.90	159
L	0.99	0.99	0.99	149
M	0.99	0.99	0.99	168
N	0.97	0.95	0.96	145
O	0.94	0.99	0.97	148
P	0.96	0.90	0.93	145
Q	0.97	0.94	0.96	151
R	0.85	0.94	0.89	157
S	0.96	0.98	0.97	162
T	0.98	0.98	0.98	161
U	0.99	0.98	0.98	159
V	0.96	0.97	0.97	148
W	0.99	0.99	0.99	150
X	0.95	0.96	0.95	166
Y	0.99	0.97	0.98	156
Z	0.99	0.99	0.99	144
accuracy			0.96	4000
macro avg	0.96	0.96	0.96	4000
weighted avg	0.96	0.96	0.96	4000

```
[12]: array([[167,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
         0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
         0, 148,  0,  0,  2,  0,  0,  2,  0,  0,  2,  0,  0,
         0,  0,  0,  0,  7,  2,  0,  0,  1,  0,  0,  0,  0])
```

```

[ 0, 0, 152, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0,
  0, 1, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0, 0],
[ 0, 0, 0, 157, 0, 0, 0, 2, 0, 0, 0, 0, 0,
  0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0],
[ 0, 1, 1, 0, 137, 1, 2, 0, 0, 0, 1, 1, 0,
  0, 0, 0, 1, 0, 1, 0, 0, 0, 2, 0, 1],
[ 0, 0, 0, 0, 0, 149, 0, 0, 0, 0, 0, 0, 0,
  2, 0, 4, 0, 0, 0, 2, 0, 0, 0, 0, 1],
[ 0, 0, 0, 1, 0, 0, 156, 0, 0, 0, 0, 0, 0,
  0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 2, 0, 0, 0, 0, 2, 126, 0, 0, 1, 0, 1,
  1, 1, 0, 0, 4, 0, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 124, 5, 0, 0, 0,
  0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 6, 144, 0, 0, 0,
  0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 1, 0, 0, 1, 0, 0, 6, 0, 0, 137, 0, 0,
  0, 0, 0, 0, 8, 0, 0, 0, 0, 6, 0, 0],
[ 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 1, 147, 0,
  0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 166,
  0, 0, 0, 0, 0, 0, 0, 1, 1, 0, 0, 0],
[ 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 0, 0,
  138, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
  0, 147, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 1, 0, 0, 0, 11, 0, 1, 0, 0, 0, 1, 1,
  0, 0, 130, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 1, 0, 2, 0, 0, 0, 0, 0, 0,
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[ 0, 3, 0, 1, 0, 0, 0, 1, 0, 0, 2, 0, 0,
  2, 0, 0, 0, 148, 0, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
  0, 0, 0, 0, 4, 158, 0, 0, 0, 0, 0, 0],
[ 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
  0, 0, 0, 0, 0, 157, 0, 1, 0, 0, 1, 0],
[ 1, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
  0, 1, 0, 0, 0, 0, 156, 0, 0, 0, 0, 0],
[ 0, 4, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
  0, 0, 0, 0, 0, 0, 144, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
  0, 0, 0, 0, 0, 0, 0, 1, 149, 0, 0],
[ 0, 2, 0, 0, 1, 0, 0, 0, 0, 0, 2, 0, 0,
  0, 0, 0, 0, 0, 2, 0, 0, 159, 0, 0],
[ 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
  0, 0, 0, 0, 0, 1, 1, 1, 0, 1, 151],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,

```



```
0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 0, 142]],
dtype=int64)
```

### 1.7.2 Choise tree

```
[13]: from sklearn.tree import DecisionTreeClassifier

tree = DecisionTreeClassifier(criterion="entropy", splitter="best",
    ↪max_depth=20, min_samples_split=2)
tree.fit(train,train_classes)
prediction = tree.predict(test)

print(classification_report(test_classes, prediction))
confusion_matrix(test_classes, prediction)
```

	precision	recall	f1-score	support
A	0.94	0.95	0.94	168
B	0.85	0.85	0.85	164
C	0.89	0.90	0.90	157
D	0.86	0.89	0.87	160
E	0.91	0.86	0.89	149
F	0.89	0.82	0.86	158
G	0.84	0.90	0.87	159
H	0.78	0.79	0.78	138
I	0.89	0.94	0.91	129
J	0.89	0.90	0.90	150
K	0.86	0.82	0.84	159
L	0.88	0.91	0.90	149
M	0.94	0.95	0.94	168
N	0.93	0.88	0.90	145
O	0.84	0.89	0.87	148
P	0.85	0.88	0.86	145
Q	0.90	0.86	0.88	151
R	0.83	0.83	0.83	157
S	0.90	0.88	0.89	162
T	0.88	0.93	0.90	161
U	0.92	0.92	0.92	159
V	0.89	0.91	0.90	148
W	0.96	0.91	0.94	150
X	0.91	0.92	0.91	166
Y	0.90	0.87	0.88	156
Z	0.96	0.92	0.94	144
accuracy			0.89	4000
macro avg	0.89	0.89	0.89	4000
weighted avg	0.89	0.89	0.89	4000

```

[13]: array([[160,  0,  2,  0,  0,  0,  0,  2,  0,  1,  0,  0,  1,
              0,  0,  0,  0,  0,  1,  0,  0,  0,  0,  1,  0],
 [  0, 140,  0,  6,  0,  2,  0,  4,  1,  1,  0,  0,  0,  0,
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 [  0,  0, 142,  0,  2,  0,  3,  0,  0,  0,  0,  3,  2,  0,
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 [  0,  2,  0, 143,  0,  1,  1,  5,  0,  0,  0,  0,  0,  0,
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 [  0,  0,  3,  0, 128,  0,  5,  0,  1,  0,  2,  3,  0,  0,
              0,  0,  0,  0,  1,  1,  0,  0,  1,  0,  1,  2,  1],
 [  0,  1,  1,  1,  1, 130,  0,  1,  1,  1,  0,  0,  1,  0,
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 [  0,  1,  4,  0,  2,  1, 143,  1,  0,  0,  0,  1,  0,  0,
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 [  0,  1,  1,  3,  1,  0,  2,  3,  1,  0, 131,  1,  1,  0,
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 [  2,  0,  0,  0,  0,  0,  1,  1,  0,  0,  0,  0, 159,  0,
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 [  0,  0,  1,  0,  0,  2,  0,  1,  1,  0,  0,  0,  0,  0,
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              1,  3,  0,  0,  0,  1,  0, 147,  0,  0,  0,  0,  1],
 [  0,  2,  0,  0,  0,  0,  1,  3,  0,  0,  0,  2,  0,  0,
              0,  1,  3,  0,  0,  0,  1,  0, 135,  0,  0,  0,  0],
 [  1,  0,  0,  0,  0,  1,  0,  0,  0,  0,  0,  0,  2,  0,
              0,  2,  0,  1,  0,  0,  1,  0,  5, 137,  0,  0,  0],
 [  0,  3,  0,  1,  1,  0,  0,  0,  0,  0,  1,  3,  0,  0,

```

```

    0, 0, 0, 0, 2, 3, 0, 0, 0, 0, 152, 0, 0],
[ 2, 2, 1, 0, 0, 2, 0, 1, 1, 1, 0, 0, 0,
 0, 1, 0, 0, 0, 0, 6, 0, 3, 1, 0, 135, 0],
[ 1, 0, 0, 0, 1, 0, 1, 0, 1, 2, 0, 0, 0,
 1, 0, 0, 1, 0, 1, 1, 0, 0, 0, 2, 0, 132]],
dtype=int64)

```

### 1.7.3 SVM

```

[14]: from sklearn.svm import SVC
      from sklearn.model_selection import GridSearchCV

      parameters = {'gamma':('scale', 'auto'), 'C':[10, 15, 20, 25, 30]}

      svc = SVC()
      clf = GridSearchCV(svc, parameters)
      clf.fit(train, train_classes)

```

```

[14]: GridSearchCV(estimator=SVC(),
                  param_grid={'C': [10, 15, 20, 25, 30], 'gamma': ('scale', 'auto')})

```

```

[18]: svc = SVC(C=20, gamma='auto')
      svc.fit(train, train_classes)
      prediction = svc.predict(test)

      print(classification_report(test_classes, prediction))
      confusion_matrix(test_classes, prediction)

```

	precision	recall	f1-score	support
A	0.99	0.99	0.99	168
B	0.95	0.96	0.96	164
C	0.99	0.99	0.99	157
D	0.98	0.99	0.98	160
E	0.98	0.96	0.97	149
F	0.98	0.97	0.98	158
G	0.95	0.98	0.96	159
H	0.92	0.91	0.92	138
I	0.97	0.97	0.97	129
J	0.98	0.97	0.98	150
K	0.97	0.93	0.95	159
L	0.99	0.99	0.99	149
M	0.98	0.99	0.99	168
N	0.98	0.97	0.97	145
O	0.98	0.99	0.98	148
P	0.98	0.95	0.97	145
Q	0.99	0.98	0.99	151
R	0.93	0.95	0.94	157

S	0.99	0.99	0.99	162
T	1.00	0.99	1.00	161
U	0.99	0.99	0.99	159
V	1.00	0.99	1.00	148
W	0.99	0.99	0.99	150
X	0.98	0.99	0.99	166
Y	0.98	0.99	0.98	156
Z	0.99	1.00	1.00	144
accuracy			0.98	4000
macro avg	0.98	0.98	0.98	4000
weighted avg	0.98	0.98	0.98	4000

```
[18]: array([[166,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  2,  0],
       [  0, 158,  0,  0,  1,  0,  0,  1,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  3,  1,  0,  0,  0,  0,  0,  0],
       [  0,  0, 156,  0,  0,  0,  1,  0,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  0,  0, 158,  0,  0,  1,  1,  0,  0,  0,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  1,  0,  0, 143,  1,  2,  0,  0,  0,  0,  0,  1,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  1],
       [  0,  0,  0,  1,  0, 154,  0,  1,  0,  0,  0,  0,  0,  0,
        0,  0,  2,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  0,  2,  0,  0,  0, 156,  0,  0,  0,  0,  0,  0,  0,
        0,  1,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  2,  0,  1,  0,  0,  3, 126,  0,  0,  0,  0,  0,  2,
        0,  1,  0,  0,  2,  0,  0,  1,  0,  0,  0,  0],
       [  0,  0,  0,  0,  0,  0,  0,  0,  0, 125,  3,  0,  0,  0,
        1,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  0,  0,  0,  0,  0,  0,  0,  0,  4, 146,  0,  0,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  0,  0,  0,  0,  0,  0,  2,  0,  0,  0, 148,  0,  0,
        0,  0,  0,  0,  6,  0,  0,  0,  0,  3,  0,  0],
       [  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  1, 148,  0,
        0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0, 167,
        1,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  0,  0,  2,  0,  0,  0,  2,  0,  0,  0,  0,  1,  0,
       140,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0],
       [  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,  0,
        0, 146,  0,  1,  0,  0,  0,  1,  0,  0,  0,  0],
       [  0,  1,  0,  0,  2,  2,  1,  0,  0,  0,  0,  0,  0,  0,
        0,  0, 138,  0,  0,  0,  0,  0,  1,  0,  0,  0],
       [  0,  0,  0,  0,  0,  0,  1,  0,  0,  0,  0,  0,  0,  0,
```

```

0, 1, 1, 148, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 2, 0, 0, 0, 0, 0, 2, 0, 0, 3, 0, 0,
 1, 0, 0, 0, 149, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 161, 0, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 160, 0, 0, 0, 0, 0, 0],
[ 1, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 0, 157, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 0, 0, 147, 0, 0, 1, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1,
 0, 0, 0, 0, 0, 0, 0, 0, 0, 149, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 165, 0, 0],
[ 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 0, 0, 0, 1, 154, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 144]],
dtype=int64)

```

#### 1.7.4 Random Forest

```

[19]: from sklearn.ensemble import RandomForestClassifier

rfc = RandomForestClassifier(n_estimators=130, criterion="entropy",
    ↪max_depth=20, min_samples_split=2)
rfc.fit(train,train_classes)
prediction = rfc.predict(test)

print(classification_report(test_classes, prediction))
confusion_matrix(test_classes, prediction)

```

	precision	recall	f1-score	support
A	0.99	1.00	0.99	168
B	0.89	0.96	0.92	164
C	0.99	0.97	0.98	157
D	0.95	0.98	0.97	160
E	0.95	0.93	0.94	149
F	0.94	0.92	0.93	158
G	0.95	0.97	0.96	159
H	0.91	0.91	0.91	138
I	0.94	0.97	0.95	129
J	0.97	0.93	0.95	150
K	0.95	0.91	0.93	159
L	0.99	0.97	0.98	149
M	0.99	0.98	0.98	168

N	0.96	0.95	0.96	145
O	0.96	0.98	0.97	148
P	0.95	0.92	0.94	145
Q	0.94	0.97	0.95	151
R	0.92	0.95	0.93	157
S	0.99	0.96	0.97	162
T	0.98	0.98	0.98	161
U	0.98	1.00	0.99	159
V	0.98	0.97	0.98	148
W	0.99	0.99	0.99	150
X	0.96	0.98	0.97	166
Y	0.99	0.99	0.99	156
Z	0.99	0.98	0.99	144
accuracy			0.96	4000
macro avg	0.96	0.96	0.96	4000
weighted avg	0.96	0.96	0.96	4000

```
[19]: array([[168, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 157, 0, 0, 1, 0, 0, 2, 0, 0, 1, 0, 0,
1, 0, 0, 0, 0, 0, 0, 1, 0, 1, 0, 0],
[ 0, 0, 152, 0, 2, 0, 1, 0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 1, 0, 0, 0, 1, 0, 0],
[ 0, 0, 0, 157, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0,
2, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 1, 1, 0, 138, 1, 3, 0, 0, 0, 0, 0, 2, 0,
0, 0, 0, 1, 0, 0, 0, 0, 1, 0],
[ 0, 2, 0, 1, 0, 146, 0, 1, 0, 0, 1, 0, 0, 0,
0, 0, 6, 0, 0, 0, 1, 0, 0, 0],
[ 0, 1, 0, 1, 1, 0, 154, 0, 0, 0, 0, 0, 0, 0,
0, 0, 0, 2, 0, 0, 0, 0, 0, 0],
[ 0, 2, 0, 0, 0, 1, 1, 126, 0, 0, 2, 0, 0, 0,
0, 0, 0, 1, 2, 0, 0, 1, 0, 0],
[ 0, 0, 0, 0, 0, 1, 0, 0, 125, 2, 0, 0, 0, 0,
0, 0, 1, 0, 0, 0, 0, 0, 0, 0],
[ 1, 0, 0, 0, 0, 0, 0, 0, 1, 8, 139, 0, 0, 0,
0, 0, 0, 0, 0, 1, 0, 0, 0, 0],
[ 0, 0, 0, 1, 1, 0, 0, 2, 0, 0, 0, 144, 0, 0,
0, 0, 0, 0, 7, 0, 1, 0, 0, 0],
[ 0, 2, 0, 0, 0, 0, 1, 0, 0, 1, 0, 145, 0, 0,
0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 164,
2, 0, 0, 0, 0, 0, 0, 0, 1, 0, 0],
[ 0, 0, 0, 2, 0, 0, 0, 2, 0, 0, 0, 0, 0, 0,
138, 1, 0, 0, 1, 0, 0, 0, 1, 0, 0]]
```

```

[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 145, 0, 2, 0, 0, 0, 1, 0, 0, 0, 0, 0],
[ 0, 1, 0, 0, 1, 6, 1, 1, 0, 0, 0, 0, 0,
 0, 1, 134, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0],
[ 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 3, 0, 146, 1, 0, 0, 0, 0, 0, 0, 0, 0],
[ 0, 3, 0, 1, 0, 0, 0, 1, 0, 0, 2, 0, 0,
 1, 0, 0, 0, 149, 0, 0, 0, 0, 0, 0, 0],
[ 0, 1, 0, 0, 0, 0, 0, 2, 0, 1, 0, 0, 0,
 0, 0, 0, 1, 2, 155, 0, 0, 0, 0, 0, 0],
[ 0, 0, 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 158, 0, 1, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 0, 159, 0, 0, 0, 0],
[ 0, 4, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 0, 0, 144, 0, 0, 0],
[ 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 1,
 0, 1, 0, 0, 0, 0, 0, 0, 0, 148, 0, 0],
[ 0, 1, 0, 1, 0, 0, 0, 0, 0, 0, 0, 1, 0,
 0, 0, 0, 0, 0, 0, 0, 0, 0, 163, 0, 0],
[ 0, 1, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 0, 0, 0, 1, 0, 0, 0, 154, 0],
[ 0, 0, 0, 0, 1, 0, 0, 0, 0, 0, 0, 0, 0,
 0, 0, 0, 2, 0, 0, 0, 0, 0, 0, 0, 141]],
dtype=int64)

```

### 1.7.5 AdaBoost

```

[20]: from sklearn.ensemble import AdaBoostClassifier

abc = AdaBoostClassifier(n_estimators=70, learning_rate=0.25)
abc.fit(train,train_classes)
prediction = abc.predict(test)

print(classification_report(test_classes, prediction))
confusion_matrix(test_classes, prediction)

```

C:\Users\Lenovo\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11\_qbz5n2kfra8p0\LocalCache\local-packages\Python311\site-packages\sklearn\ensemble\\_weight\_boosting.py:527: FutureWarning: The SAMME.R algorithm (the default) is deprecated and will be removed in 1.6. Use the SAMME algorithm to circumvent this warning.

```
warnings.warn(
```

	precision	recall	f1-score	support
A	0.71	0.07	0.13	168
B	0.44	0.61	0.51	164

C	0.80	0.48	0.60	157
D	0.63	0.29	0.40	160
E	0.31	0.35	0.33	149
F	0.40	0.54	0.46	158
G	0.38	0.83	0.52	159
H	0.07	0.40	0.13	138
I	0.42	0.06	0.11	129
J	0.27	0.51	0.36	150
K	0.30	0.09	0.14	159
L	0.00	0.00	0.00	149
M	0.52	0.37	0.43	168
N	0.40	0.21	0.28	145
O	0.43	0.81	0.56	148
P	0.76	0.51	0.61	145
Q	0.43	0.17	0.25	151
R	0.45	0.66	0.53	157
S	0.28	0.38	0.32	162
T	0.36	0.21	0.27	161
U	0.71	0.57	0.63	159
V	0.06	0.05	0.05	148
W	0.40	0.03	0.05	150
X	0.44	0.78	0.57	166
Y	0.77	0.06	0.12	156
Z	0.30	0.07	0.11	144
accuracy			0.35	4000
macro avg	0.42	0.35	0.33	4000
weighted avg	0.43	0.35	0.33	4000

```
[20]: array([[ 12,  1,  0,  0,  0,  0,  4, 114,  0,  8,  1,  0,  0,
               0,  1,  0, 12, 10,  4,  0,  0,  0,  0,  1,  0,  0],
 [  0, 100,  0,  0,  0,  0,  1,  3,  1,  1,  5,  0,  0,
               0,  0,  0,  0, 38,  6,  0,  0,  3,  0,  6,  0,  0],
 [  0,  0, 75,  0, 26,  0, 29,  3,  0,  0,  3,  1,  0,
               0,  5,  0,  0,  0,  1,  2,  9,  0,  0,  3,  0,  0],
 [  0, 18,  0, 47,  0,  3,  1, 38,  0,  3,  1,  0,  0,
               1, 36,  0,  0,  6,  5,  1,  0,  0,  0,  0,  0,  0],
 [  0,  2,  3,  0, 52,  0, 28,  1,  0,  0,  7,  5,  0,
               0,  0,  0,  0,  0,  8,  6,  0,  0,  0, 35,  0,  2],
 [  0,  2,  0,  1,  0, 85,  5, 16,  0, 22,  0,  0,  0,
               0,  0,  9,  0,  2,  1, 12,  0,  1,  0,  2,  0,  0],
 [  0,  4,  1,  0,  3,  0,132,  2,  0,  0,  3,  0,  2,
               0,  1,  0,  0,  5,  5,  0,  0,  1,  0,  0,  0,  0],
 [  0,  6,  0,  0,  0,  1,  4, 55,  0,  0,  0,  0,  1,
               0, 49,  0,  0, 15,  6,  0,  0,  1,  0,  0,  0,  0],
 [  0,  3,  1,  0,  3,  3,  1,  0,  8, 76,  0,  0,  0,
```



```

    0, 3, 0, 1, 0, 30, 0, 0, 0, 0, 0, 0, 0],
[ 0, 7, 0, 1, 0, 3, 0, 31, 4, 77, 0, 0, 0],
  0, 13, 0, 3, 4, 4, 0, 0, 0, 0, 3, 0, 0],
[ 0, 5, 4, 1, 12, 0, 15, 49, 0, 0, 14, 0, 0],
  0, 10, 0, 10, 21, 1, 2, 1, 0, 0, 14, 0, 0],
[ 2, 2, 0, 14, 2, 0, 8, 7, 2, 62, 2, 0, 0],
  0, 0, 0, 4, 1, 39, 2, 0, 0, 0, 2, 0, 0],
[ 0, 5, 1, 0, 0, 0, 0, 66, 0, 0, 3, 0, 62],
  9, 0, 0, 0, 1, 0, 0, 13, 2, 6, 0, 0, 0],
[ 1, 2, 0, 1, 0, 2, 0, 53, 0, 1, 1, 0, 27],
  31, 14, 0, 0, 2, 0, 0, 4, 6, 0, 0, 0, 0],
[ 0, 4, 1, 0, 0, 0, 8, 0, 0, 0, 0, 0, 3],
  0, 120, 0, 1, 7, 0, 0, 1, 3, 0, 0, 0, 0],
[ 0, 11, 0, 0, 0, 33, 6, 12, 2, 0, 0, 0, 0],
  0, 1, 74, 0, 0, 0, 0, 0, 3, 0, 1, 2, 0],
[ 0, 5, 1, 6, 1, 0, 76, 3, 0, 1, 1, 0, 4],
  0, 15, 0, 26, 0, 10, 0, 0, 2, 0, 0, 0, 0],
[ 0, 21, 0, 2, 0, 0, 16, 11, 0, 0, 1, 0, 0],
  0, 1, 0, 0, 103, 0, 0, 0, 2, 0, 0, 0, 0],
[ 1, 13, 0, 0, 9, 1, 1, 8, 2, 7, 2, 0, 0],
  0, 0, 0, 3, 3, 61, 1, 0, 0, 0, 32, 0, 18],
[ 0, 1, 0, 1, 0, 53, 6, 30, 0, 19, 0, 0, 0],
  0, 1, 0, 0, 1, 2, 34, 1, 0, 0, 11, 0, 1],
[ 0, 3, 6, 1, 0, 0, 3, 40, 0, 0, 0, 0, 2],
  0, 7, 0, 0, 1, 1, 3, 91, 1, 0, 0, 0, 0],
[ 0, 8, 0, 0, 0, 2, 1, 114, 0, 0, 0, 0, 0],
  1, 0, 12, 0, 0, 0, 0, 2, 7, 0, 0, 1, 0],
[ 1, 2, 0, 0, 0, 0, 0, 14, 0, 0, 0, 0, 18],
  35, 0, 0, 0, 1, 0, 0, 4, 71, 4, 0, 0, 0],
[ 0, 3, 0, 0, 2, 0, 0, 3, 0, 0, 2, 0, 0],
  0, 1, 0, 0, 5, 6, 11, 1, 0, 0, 130, 0, 2],
[ 0, 1, 1, 0, 1, 27, 2, 65, 0, 0, 1, 0, 0],
  0, 2, 3, 0, 0, 9, 19, 2, 8, 0, 5, 10, 0],
[ 0, 0, 0, 0, 57, 0, 1, 0, 0, 4, 0, 0, 0],
  0, 0, 0, 1, 5, 17, 1, 0, 0, 0, 48, 0, 10]],
dtype=int64)

```

, SVC.