# 6. Appendix (Source code)

### StrokePrediction

#### January 19, 2022

```
[1]: import pandas as pd
    import numpy as np
    from sklearn.naive bayes import GaussianNB
    from sklearn.preprocessing import LabelEncoder
[2]: | df = pd.read csv('healthcare-dataset-stroke-data.csv')
            id gender age hypertension heart disease ever married \
[2]:
          9046 Male 67.0 0
                                1
                                     Yes
          51676 Female 61.0
                                           Yes
    2
          31112Male 80.0 0
                               1
                                     Yes
    3
          60182 Female 49.0
                                     0
                                           Yes
         1665 Female 79.0
                                     0
                                           Yes
            ... ...
    5105 18234 Female 80.0
                                     0
                                           Yes
    5106 44873 Female 81.0
                                     0
                                           Yes
    5107 19723 Female 35.0
                               0
                                     0
                                           Yes
    5108 37544 Male 51.0 0
                                     Yes
    5109 44679 Female 44.0
                                     0
                                           Yes
          work_type Residence_type avg_glucose_levelbmi smoking_status \
              Private Urban 228.69 36.6 formerly smoked
    0
              Self-employed Rural 202.21 NaN never smoked
    2 Private Rural 105.92 32.5 never smoked0
                                                      Error! Bookmark not
    defined.
    1
                                                                        3
    2
                                                                        3
    3
                                                                        3
    4
                                                                        3
    3
              Private Urban 171.23 34.4 smokes
              Self-employed
                               Rural 174.12 24.0 never smoked
    5105 Private Urban 83.75 NaN never smoked 5106 Self-employed Urban
    125.20 40.0 never smoked
    5107 Self-employed Rural 82.99 30.6 never smoked 5108 Private Rural
    166.29 25.6 formerly smoked
    5109 Govt job
                                      85.28 26.2 Unknown
                             Urban
         stroke
```

```
5105
               0
     5106
               0
     5107
               0
     5108
     5109
     5110
             rows x 12 columns]
[3]: df.isna().sum()
                          0
[3]: id
                          0
    gender
     age
    hypertension
    heart_disease
    ever married
    work type
    Residence type
    avg glucose level
    bmi
                        201
    smoking status
                          0
    stroke
dtype: int64
[4]: df 1 = df.dropna()
     df 1
            id gender age hypertension heart_disease ever_married \
[4]:
           9046 Male 67.0
                                       0
                                                     1
     0
     2
           31112Male 80.0 0
                                       Yes
     3
           60182 Female 49.0
                                       0
                                             Yes
           1665 Female 79.0
                                             Yes
           56669Male 81.0 0
                                       Yes
     5104 14180 Female 13.0
                                       0
                                                               No
     5106 44873 Female 81.0
                                       0
                                             Yes
     5107 19723 Female 35.0
                                       0
                                             Yes
     5108 37544 Male 51.0 0
                                       Yes
     5109 44679 Female 44.0
                                             Yes
```

```
228.69 36.6 formerly smoked
    0
              Private
                             Urban
    2
                          Rural 105.92 32.5
                                                never smoked
              Private
                          Urban 171.23 34.4
    3
              Private
                                                smokes
                                             174.12 24.0 never smoked
    4
        Self-employed
                             Rural
    5
              Private
                              Urban
                                              186.21
                                                       29.0
                                                               formerly
                                              smoked
    5104
                                             103.08 18.6
             children
                             Rural
                                                                Unknown
    5106 Self-employed
                             Urban
                                             125.20 40.0 never smoked
    5107 Self-employed
                                             82.99 30.6 never smoked
                             Rural
                                                      25.6
    5108
              Private
                                             166.29
                                                               formerly
                             Rural
                                              smoked
                                              85.28 26.2
    5109
             Govt job
                             Urban
                                                                Unknown
         stroke
    0
              1
    2
              1
    3
              1
    4
    5
              1
    5104
              0
    5106
             0
    5107
    5108
             0
    5109
             0
    [4909 rows x 12 columns]
[5]: number = LabelEncoder() df 1['gender'] =
    number.fit transform(df_1['gender']) df_1['ever_married'] =
    number.fit transform(df 1['ever married'])
    df 1['work type'] = number.fit transform(df 1['work type'])
    df 1['Residence type'] =
    number.fit transform(df 1['Residence type'])
    df 1['smoking status'] =
    number.fit transform(df 1['smoking status'])
    C:\Users\asus\AppData\Local\Temp/ipykernel 2096/1938741493.py:2:
    SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a
    DataFrame. Try using .loc[row indexer,col indexer] = value
    instead
```

work type Residence type avg glucose levelbmi smoking status \

```
See the caveats in the documentation:
https://pandas.pydata.org/pandasdocs/stable/user guide/indexing.html
#returning-a-view-versus-a-copy df 1['gender'] =
number.fit transform(df 1['gender'])
C:\Users\asus\AppData\Local\Temp/ipykernel 2096/1938741493.py:3:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a
DataFrame. Try using .loc[row indexer,col indexer] = value
instead
See the caveats in the documentation:
https://pandas.pydata.org/pandasdocs/stable/user guide/indexing.html
#returning-a-view-versus-a-copy df 1['ever married'] =
number.fit_transform(df_1['ever_married'])
C:\Users\asus\AppData\Local\Temp/ipykernel 2096/1938741493.py:4:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a
DataFrame. Try using .loc[row indexer,col indexer] = value
instead
See the caveats in the documentation:
https://pandas.pydata.org/pandasdocs/stable/user_guide/indexing.html
#returning-a-view-versus-a-copy df 1['work type'] =
number.fit transform(df 1['work type'])
C:\Users\asus\AppData\Local\Temp/ipykernel 2096/1938741493.py:5:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a
DataFrame. Try using .loc[row indexer,col indexer] = value
instead
See the caveats in the documentation:
https://pandas.pydata.org/pandasdocs/stable/user guide/indexing.html
#returning-a-view-versus-a-copy df 1['Residence type'] =
number.fit transform(df 1['Residence type'])
C:\Users\asus\AppData\Local\Temp/ipykernel 2096/1938741493.py:6:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a
DataFrame. Try using .loc[row indexer,col indexer] = value
instead
See the caveats in the documentation:
https://pandas.pydata.org/pandasdocs/stable/user guide/indexing.html
#returning-a-view-versus-a-copy
                                     df 1['smoking status'] =
                 number.fit transform(df 1['smoking status'])
```

## 1 Data balancing process

```
[6]: df 1['stroke'].unique()
[6]: array([1, 0], dtype=int64)
[7]: df 1['stroke'].value counts()
[7]: 0
         4700
     1
          209
     Name: stroke, dtype: int64
[8]: df 2 = (df 1.groupby('stroke', as index=False)
           .apply(lambda X: X.sample(n=209))
           .reset index(drop=True))
[9]: df 1.info()
    <class 'pandas.core.frame.DataFrame'>
    Int64Index: 4909 entries, 0 to 5109
     Data columns (total 12 columns):
      Column
                         Non-Null
                                       Count
                          Dtype
    ____
    0
        id
                          4909 non-nullint64
   gender 4909 non-null
                          int32
   age
           4909 non-null
                          float64
   hypertension 4909 non-null int64
3
   heart disease
                    4909 non-null int64
4
5
   ever married 4909 non-null
6
   work type
               4909 non-null int32
   Residence type
                    4909 non-null
7
                                      int32
   avg glucose level 4909 non-null float64
   bmi
          4909 non-null
9
                          float64
10 smoking status
                    4909 non-null
                                       int32
11 stroke 4909 non-null
     dtypes: float64(3), int32(5),
     int64(4) memory usage: 402.7 KB
[11]: df 2['stroke'].value counts()
```

```
209
     Name: stroke, dtype: int64
         Split Dataset
    2
[12]:
                                                                          df 2
            id gender age hypertension heart disease ever married \
[12]:
         30290 0 40.0
                            0
                                  0
                                        1
     0
         62715 1 82.0
     2
         45945 1 46.0
                            0
                                        1
     3
         71533 1 50.0
                                  0
                                        1
                            0
         29869 1 49.0
                                       1
     413
         48405 1 80.0
                            0
                                  1
                                        1
     414 5317 0 79.0
                            0
                                        1
     415 32221 1 60.0
                            0
                                  1
                                       1
                                  0
                                        1
     416 2458 0 78.0
                            0
     417 8899 1 49.0
       work_type Residence_type avg_glucose_levelbmi smoking_status \
                            70.13 23.6 2
     0
                 2
                      1
     1
                 2
                      1
                            57.56 27.5 2
                 2
                      1
                            178.76 24.1
                                             2
                 2
                      1
                            158.31 32.8
                                             1
                 2
                      1
                           199.96 28.6
                                             2
     413 2 1 68.53 24.2 3 414 2 1 214.09 28.2 2
     415
                      1
                            91.92 35.9 3
                 2
     416
                 2
                      0
                                             2
                            235.63 32.3
     417
                 2
                      0
                            104.86 31.9
          stroke
     0
              0
     1
              0
     2
              0
     3
              0
              0
     413
              1
     414
              1
```

[11]: 0

209 1

```
416
             1
     417
              1
     [418 rows x 12 columns]
[13]: X = df 2.iloc[:,1:10]
     y = df_2['stroke']
[14]: from sklearn.model selection import train test split
→ random state=50, stratify=y)
[16]: print(X train.shape)
     print(X test.shape)
     print(y train.shape)
     print(y test.shape)
    (292, 9)
     (126, 9)
     (292,)
     (126,)
[21]: y train[y train==1].size
[21]: 146
[18]: y_train[y_train==0].size
[18]: 146
[19]: y test[y test==1].size
[19]: 63
[20]: | y_test[y_test==0].size
[20]: 63
    3 Naive Bayes
[22]: nb = GaussianNB()
     nb.fit(X train, y train)
[22]: GaussianNB()
[23]: nb.score(X_train, y_train)
[23]: 0.7671232876712328
[42]: nb.score(X test, y test)
[42]: 0.6984126984126984
```

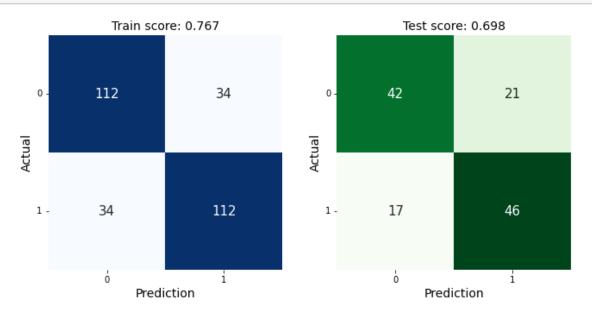
```
[24]: y pred = nb.predict(X test)
[27]: y_pred
  [27]: array([0, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 0, 1, 0,
  1, 0, 0, 1, 0, 0, 0, 0, 1, 1, 0, 1, 1, 0, 1, 0, 1, 0, 0, 1, 1, 0,
                                                            1, 0, 0, 0,
   0, 0, 1, 1, 1, 1, 0, 0, 1, 0, 0, 1, 1, 1, 0, 1, 1, 1, 1, 1, 1, 0, 1,
   1, 0, 0, 1, 1, 1, 1, 0, 0, 1, 1, 1, 1, 1, 0, 0, 0, 0, 1, 0, 0, 1,
   0, 1, 1, 0, 1, 1, 1, 0, 1, 1, 1, 1, 0, 1, 0, 0, 0, 0, 0, 1, 1, 0,
            1, 1, 0, 1, 1, 1, 0, 0, 0, 1, 0, 0, 1, 1, 1, 1], dtype=int64)
[28]: from sklearn.metrics import confusion matrix
     from sklearn.metrics import classification report
[29]: print(confusion matrix(y test, y pred))
     [[42 21]
     [17 46]]
[30]: print(classification report(y test, y pred))
                 precision recall f1-score support
               0
                      0.71
                             0.67
                                        0.69
                                                    63
               1
                      0.69
                             0.73
                                        0.71
                                                    63
        accuracy
                                        0.70
                                                  126
       macro avg
                      0.70
                             0.70
                                        0.70
                                                  126
    weighted avg
                      0.70
                             0.70
                                        0.70
                                                   126
```

# 4 Visualisasi Hasil

[38]: **from jcopml.plot import** plot\_confusion\_matrix

Matplotlib is building the font cache; this may take a moment.

[41]: plot\_confusion\_matrix(X\_train, y\_train, X\_test, y\_test, nb)



[]: