

資料前處理

1. 使用 explode 將 Permission 的 list 展開

```
In [134]: df_ = df.explode('Permission')
df_['Permission'].drop_duplicates(inplace=True)
df_.head()
```

```
Out[134]:
```

	Class	Permission	apk
0	Adware	android.permission.ACCESS_FINE_LOCATION	00325582d7caaf0f36ad333869df444a1aa39326e60745...
0	Adware	android.permission.SYSTEM_ALERT_WINDOW	00325582d7caaf0f36ad333869df444a1aa39326e60745...
0	Adware	android.permission.GET_TASKS	00325582d7caaf0f36ad333869df444a1aa39326e60745...
0	Adware	android.permission.RESTART_PACKAGES	00325582d7caaf0f36ad333869df444a1aa39326e60745...
0	Adware	android.permission.VIBRATE	00325582d7caaf0f36ad333869df444a1aa39326e60745...

2. 根據 apk 和 class 將展開後的 Permission 重新分配到相應的欄位。

```
In [135]: df_['count'] = 1
new_df = df_.pivot_table(index=['apk', 'Class'], columns=['Permission'], values='count')
```

```
In [136]: new_df = new_df.sort_index(axis=1, level=1)
new_df = new_df.reset_index()
```

```
In [137]: new_df
```

```
Out[137]:
```

	Permission	apk	Class	ANDROID.PERMISSION.ACCESS_COARSE_LOCATION	ANDROID.PERMISSION.ACCESS_F
0	00325582d7caaf0f36ad333869df444a1aa39326e60745...	Adware		NaN	
1	00567501fe3517b1bf6b363dd6c4d0972a77c591b0d7c2...	Adware		NaN	
2	00621e015191863041e78726b863b7e1374b17fda69036...	Adware		NaN	
3	0149b02468a8e582d76a7ad5094c15a4a07f4f08361ba5...	Adware		NaN	
4	018096f1e732c91fb315a84ab5d851430c5d2c124a7f8d...	Adware		NaN	
...
2269	ffeb97d2f85d1e30f10562c003e695f323a10f905f87c0...	benign		NaN	
2270	ffec366e301184bb458e1375d99cd58b6d918d857a108...	benign		NaN	
2271	ffecd2343d0d4fe17071180b355e7d30744f74915ede4f...	benign		NaN	
2272	fff4c39a07f20e062231781a0d7f2e039f59286d3f5d8d...	SMS		NaN	
2273	fff9e36e72ca18a049929c7f2f584f57b6fa2a03dfbe40...	Banking		NaN	

2274 rows x 880 columns

3. 填補空值

```
In [138]: new_df.fillna(0, inplace=True)
```

```
In [139]: new_df.head()
```

```
Out[139]:
```

	Permission	apk	Class	ANDROID.PERMISSION.ACCESS_COARSE_LOCATION	ANDROID.PERMISSION.ACCESS_F
0	00325582d7caaf0f36ad333869df444a1aa39326e60745...	Adware		0.0	
1	00567501fe3517b1bf6b363dd6c4d0972a77c591b0d7c2...	Adware		0.0	
2	00621e015191863041e78726b863b7e1374b17fda69036...	Adware		0.0	
3	0149b02468a8e582d76a7ad5094c15a4a07f4f08361ba5...	Adware		0.0	
4	018096f1e732c91fb315a84ab5d851430c5d2c124a7f8d...	Adware		0.0	

5 rows x 880 columns

4. 將 Class 改成數值，benign 為 0，其餘為 1

```
In [147]: new_df['Class'] = new_df['Class'].apply(lambda x : 0 if x=='benign' else 1)
new_df.head()
```

```
Out[147]:
```

	Permission	apk	Class	ANDROID.PERMISSION.ACCESS_COARSE_LOCATION	ANDROID.PERMISSION.ACCESS_FINE_LOCATION
0	00325582d7caaf0f36ad333869df444a1aa39326e60745...		1	0.0	0.0
1	00567501fe3517b1bf6b363dd6c4d0972a77c591b0d7c2...		1	0.0	0.0
2	00621e015191863041e78726b863b7e1374b17fda69036...		1	0.0	0.0
3	0149b02468a8e582d76a7ad5094c15a4a07f4f08361ba5...		1	0.0	0.0
4	018096f1e732c91fb315a84ab5d851430c5d2c124a7f8d...		1	0.0	0.0

5 rows x 880 columns

5. 因為欄位數太多，留下 permission 出現次數超過 5 的欄位

```
In [151]: class_df = new_df.groupby('Class').sum()
delete_permission = []
for col in class_df.columns:
    if class_df[col].sum() < 10:
        delete_permission.append(col)
len(delete_permission)
```

```
Out[151]: 755
```

```
In [152]: new_df.drop(delete_permission, axis=1, inplace=True)
new_df.groupby('Class').sum()
```

```
Out[152]:
```

	Permission	android.permission.ACCESS_COARSE_LOCATION	android.permission.ACCESS_COARSE_UPDATES	android.permission.ACCESS_DOWNLOAD_MANAGER
Class				
0		173.0	3.0	4.0
1		361.0	65.0	15.0

3 rows x 480 columns

6. 將處理好的資料儲存

儲存處理好的資料

```
In [156]: new_df.set_index('Class', inplace=True)
```

```
In [157]: new_df.to_csv('ECT_HW8_107403020.csv')
```

7. 將 apk 轉圖片

```
: permission_array = df.drop(['Class', 'apk'], axis=1).to_numpy()
len(permission_array[0])
```

```
: 123
```

```
: def apk_to_img(apk, file_name):
    apk = apk.reshape(3, 41)
    apk = np.where(apk==1, 255, 0)
    # img = Image.fromarray(apk_2D)
    file_name = 'apk_img/' + file_name
    plt.imshow(file_name, apk, cmap=cm.gray)
    # img.save(file_name)
```

```
: for i in range(len(permission_array)):
    apk = permission_array[i]
    file = 'permission'+str(i)+'.png'
    apk_to_img(apk, file)
```

8. 將處理好的資料儲存

```
In [88]: new_df = df[['Class', 'apk']]
         new_df
```

```
Out[88]:
```

	Class	apk
0	1	permission0.png
1	1	permission1.png
2	1	permission2.png
3	1	permission3.png
4	1	permission4.png
...
2269	0	permission2269.png
2270	0	permission2270.png
2271	0	permission2271.png
2272	1	permission2272.png
2273	1	permission2273.png

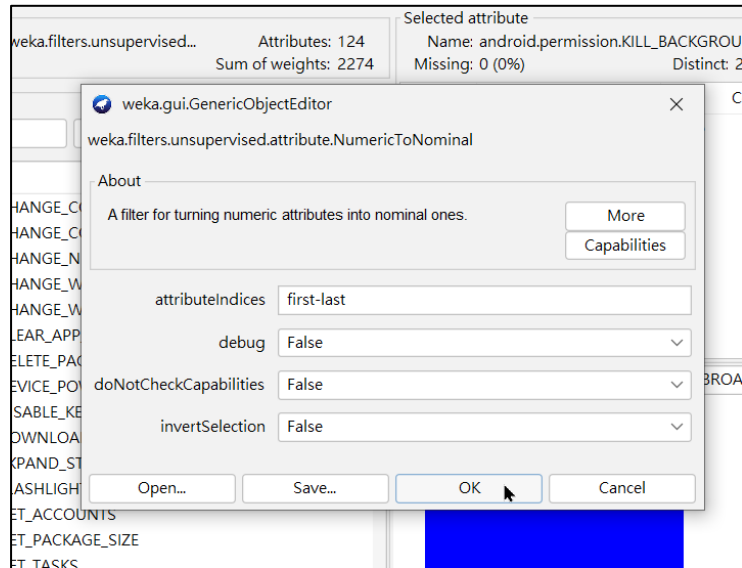
2274 rows x 2 columns

```
In [89]: new_df.set_index('Class', inplace=True)
         new_df.to_csv('ECT_HW8_107403020_CNN.csv')
```

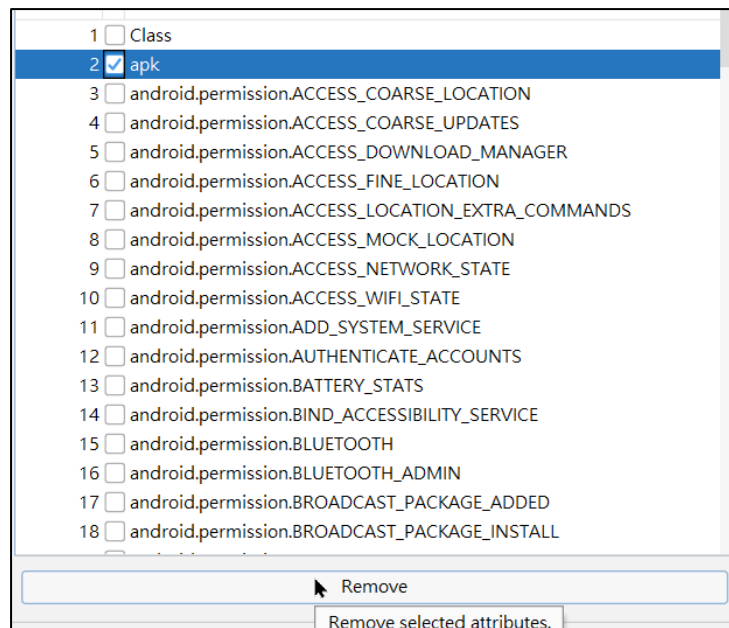
訓練模型

1. 第一類資料，用來訓練 RandomForest、MLP

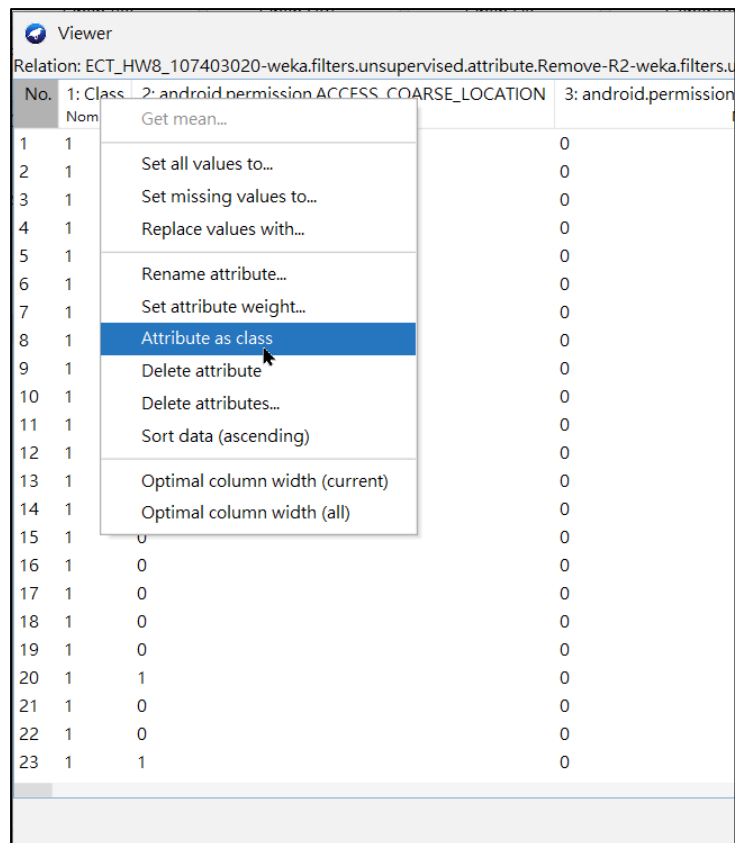
numeric_to_nominal



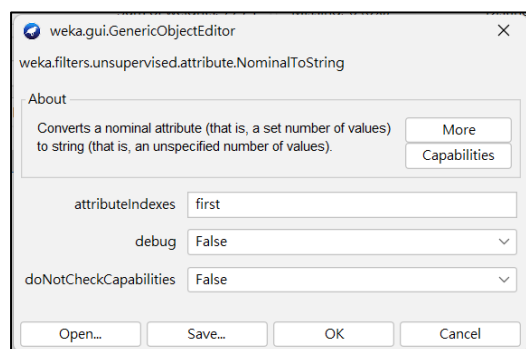
去除 apk 欄位



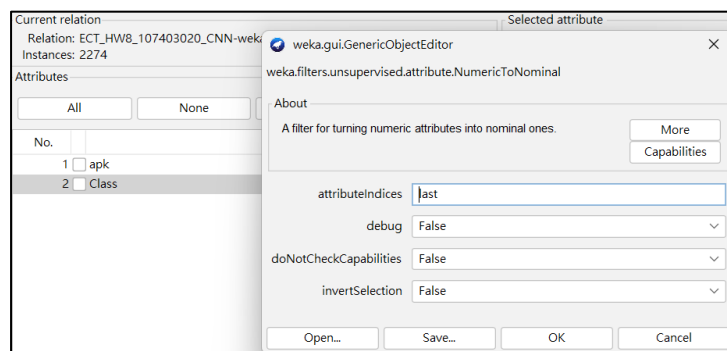
設定 Class



2. 第二類資料 apk, Class (apk 放的是圖片位置)，用來訓練 CNN
將 apk 轉 string

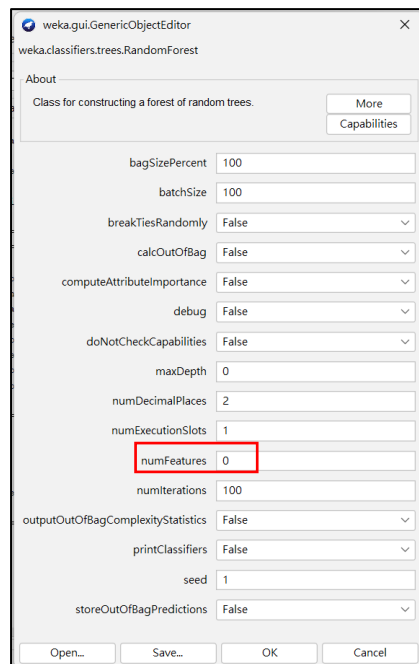


numeric_to_nominal



1. RandomForest

超參數 採用系統預設值。使用 10 Cross-validation 來進行訓練。



```
=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances      2214      97.3615 %
Incorrectly Classified Instances    60      2.6385 %
Kappa statistic                    0.9282
Mean absolute error                 0.0571
Root mean squared error            0.1484
Relative absolute error            15.3316 %
Root relative squared error        34.4049 %
Total Number of Instances          2274

=== Detailed Accuracy By Class ===

               TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC
Weighted Avg.   0.927   0.011   0.965     0.927   0.946     0.928   0.994   0.98

=== Confusion Matrix ===

  a    b  <-- classified as
521   41 |    a = 0
19 1693 |    b = 1
```

2. 將 numFeatures 設為 5，numFeatures 是設定有多少的 features 用於 random selection。

```

=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances      2211      97.2296 %
Incorrectly Classified Instances    63        2.7704 %
Kappa statistic                    0.9246
Mean absolute error                0.0601
Root mean squared error            0.1508
Relative absolute error            16.1434 %
Root relative squared error        34.9571 %
Total Number of Instances          2274

=== Detailed Accuracy By Class ===

                TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
                -----  -----  -
                0.925    0.012    0.961     0.925    0.943      0.925    0.994     0.980     0
                0.988    0.075    0.976     0.988    0.982      0.925    0.994     0.998     1
Weighted Avg.   0.972    0.059    0.972     0.972    0.972      0.925    0.994     0.993

=== Confusion Matrix ===

  a    b  <-- classified as
520   42 |    a = 0
 21 1691 |    b = 1

```

3. 將 numFeatures 設為 11

```

=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances      2212      97.2735 %
Incorrectly Classified Instances    62        2.7265 %
Kappa statistic                    0.9258
Mean absolute error                0.0548
Root mean squared error            0.1489
Relative absolute error            14.7116 %
Root relative squared error        34.5183 %
Total Number of Instances          2274

=== Detailed Accuracy By Class ===

                TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
                -----  -----  -
                0.927    0.012    0.961     0.927    0.944      0.926    0.994     0.980     0
                0.988    0.073    0.976     0.988    0.982      0.926    0.994     0.998     1
Weighted Avg.   0.973    0.058    0.973     0.973    0.973      0.926    0.994     0.993

=== Confusion Matrix ===

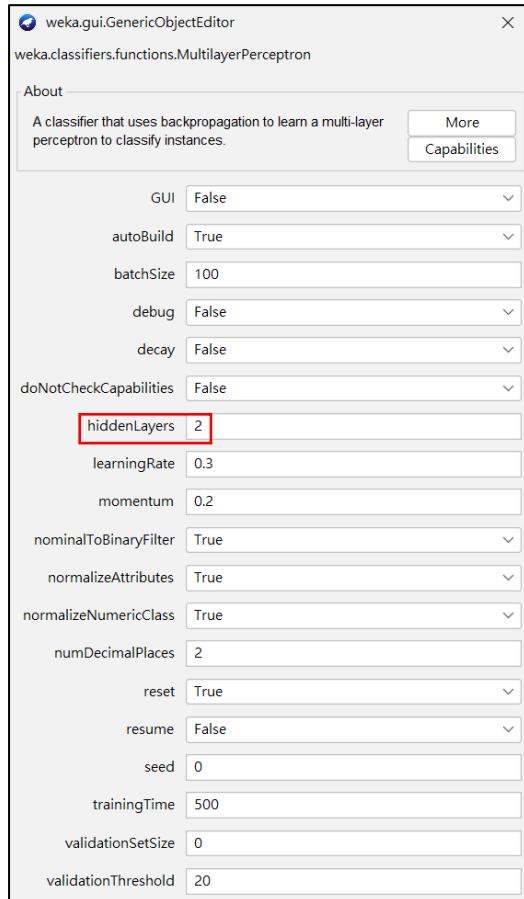
  a    b  <-- classified as
521   41 |    a = 0
 21 1691 |    b = 1

```

2. MLP

使用 10 Cross-validation 來進行訓練。

超參數：因為預設的模型 hidden layer 層數過多，訓練時間過久。將 hidden layer 改成 2，也就是指有一層 hidden layer，node 個數為 2。



weka.gui.GenericObjectEditor

weka.classifiers.functions.MultilayerPerceptron

About

A classifier that uses backpropagation to learn a multi-layer perceptron to classify instances. [More](#) [Capabilities](#)

GUI: False

autoBuild: True

batchSize: 100

debug: False

decay: False

doNotCheckCapabilities: False

hiddenLayers: 2

learningRate: 0.3

momentum: 0.2

nominalToBinaryFilter: True

normalizeAttributes: True

normalizeNumericClass: True

numDecimalPlaces: 2

reset: True

resume: False

seed: 0

trainingTime: 500

validationSetSize: 0

validationThreshold: 20

```
=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances      2196      96.5699 %
Incorrectly Classified Instances    78       3.4301 %
Kappa statistic                    0.907
Mean absolute error                 0.0414
Root mean squared error             0.1745
Relative absolute error             11.1293 %
Root relative squared error         40.4464 %
Total Number of Instances          2274

=== Detailed Accuracy By Class ===

          TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
          0.918    0.019    0.942     0.918    0.930      0.907    0.974     0.945     0
          0.981    0.082    0.973     0.981    0.977      0.907    0.974     0.988     1
Weighted Avg.   0.966    0.066    0.966     0.966    0.966      0.907    0.974     0.977

=== Confusion Matrix ===

  a    b  <-- classified as
516  46 |    a = 0
 32 1680 |    b = 1
```


超參數：hidden layer 設為 10，有一層 hidden layer，node 個數為 10

```
=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances      2192      96.394 %
Incorrectly Classified Instances    82      3.606 %
Kappa statistic                    0.902
Mean absolute error                 0.0396
Root mean squared error             0.1776
Relative absolute error             10.6328 %
Root relative squared error         41.1793 %
Total Number of Instances          2274

=== Detailed Accuracy By Class ===

                TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
                0.911    0.019    0.941     0.911    0.926      0.902    0.986     0.960     0
                0.981    0.089    0.971     0.981    0.976      0.902    0.986     0.995     1
Weighted Avg.   0.964    0.072    0.964     0.964    0.964      0.902    0.986     0.986

=== Confusion Matrix ===

  a    b  <-- classified as
512  50 |   a = 0
 32 1680 |   b = 1
```

超參數：hidden layer 設為 10, 2，代表有兩層 hidden layer，第一層 node 個數為 10，第二層為 2

```
=== Stratified cross-validation ===
=== Summary ===

Correctly Classified Instances      2184      96.0422 %
Incorrectly Classified Instances    90      3.9578 %
Kappa statistic                    0.8941
Mean absolute error                 0.0473
Root mean squared error             0.1753
Relative absolute error             12.7074 %
Root relative squared error         40.6409 %
Total Number of Instances          2274

=== Detailed Accuracy By Class ===

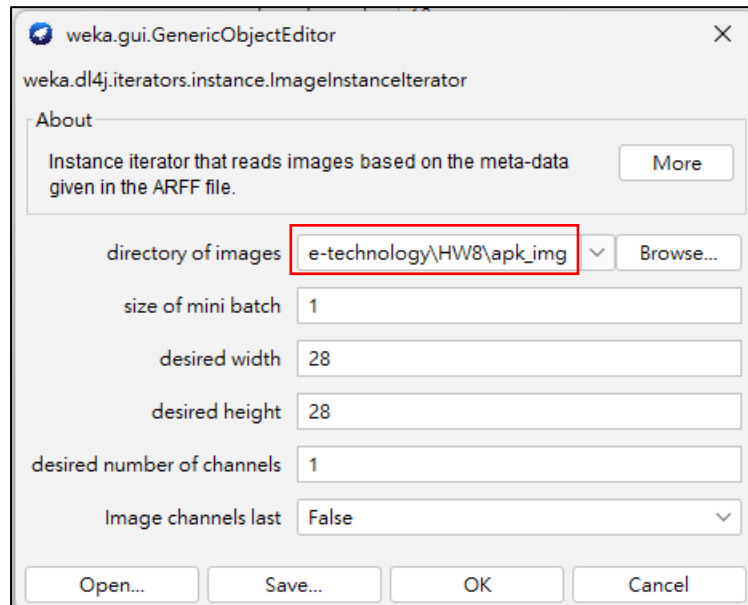
                TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
                0.927    0.029    0.914     0.927    0.920      0.894    0.982     0.956     0
                0.971    0.073    0.976     0.971    0.974      0.894    0.982     0.993     1
Weighted Avg.   0.960    0.062    0.961     0.960    0.961      0.894    0.982     0.984

=== Confusion Matrix ===

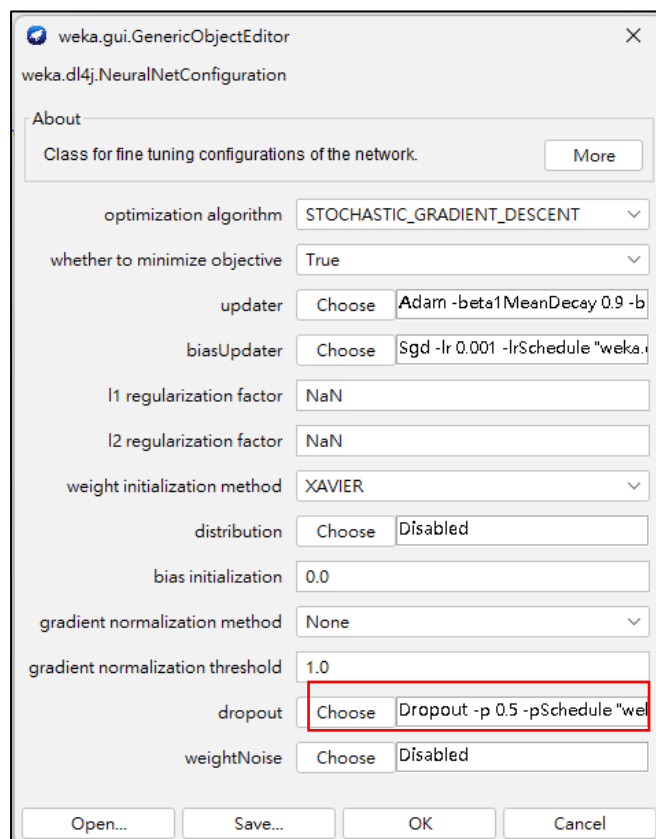
  a    b  <-- classified as
521  41 |   a = 0
 49 1663 |   b = 1
```

3. CNN

最大的問題是電腦效能，太複雜的模型電腦跑不動。**CNN** 考慮到時間因素，因此使用 **percentage split** 來將資料拆成訓練集、測試集，其中訓練集佔 66%。**instance iterator** 選擇 **ImageInstanceIterator**，讓 **weka** 會根據 **apk** 中放的檔名，到選定的資料夾讀取圖片。



設定 **dropout** 訓練時每次會隨機丟棄 0.5% 的神經節點。



- convolution layer 的 nOut 是 filter 的個數，activation 設定成 ReLU。在此的捲積層 kernel 均是 3X3，stride 均設為 1。

weka.gui.GenericObjectEditor
weka.dl4j.layers.ConvolutionLayer

About
A convolution layer from DeepLearning4J. [More](#)

layer name: Convolution layer

number of filters: 32

activation function: Choose **ActivationLReLU -alpha 0.01**

convolution mode: Truncate

CudnnAlgoMode: PREFER_FASTEST

number of rows in kernel: 3

number of columns in kernel: 3

number of rows in stride: 1

number of columns in stride: 1

number of rows in padding: 0

number of columns in padding: 0

Open... Save... OK Cancel

- subsampling 設定為 Max，會將 filter 得到的 feature map 做壓縮。kernel 均設為 2X2，stride 為 2。

weka.gui.GenericObjectEditor
weka.dl4j.layers.SubsamplingLayer

About
A subsampling layer from DeepLearning4J. [More](#)

layer name: Subsampling layer

convolution mode: Truncate

eps: 1.0E-8

pnorm: 1

number of rows in kernel: 2

number of columns in kernel: 2

number of rows in stride: 2

number of columns in stride: 2

number of rows in padding: 0

number of columns in padding: 0

pooling type: MAX

Open... Save... OK Cancel

- 第一層的 Dense 會把上一層所有的 feature map 展開，排成一行。activation 設定成 ReLU。
- 最後一層的 output layer 的 Loss function 設定成 LossNegativeLogLikelihood。

模型一：

一層 convlution:

```
=== Classifier model (full training set) ===

Network Configuration:
NeuralNetConfiguration(weightInit=XAVIER, biasInit=0.0, dist=weka.dl4j.distribution.Disabled@66, l1=NaN, l2=NaN, dropout=0.0)
Model Summary:

=====
VertexName (VertexType)      nIn,nOut  TotalParams  ParamsShape  Vertex Inputs
=====
input (InputVertex)          -, -      -            -            -
Convolution layer (ConvolutionLayer)  1,32      320          W:{32,1,3,3}, b:{1,32}  [input]
Subsampling layer (SubsamplingLayer)  -, -      0            -            [Convolution layer]
Dense layer 2 1 (DenseLayer)  5408,512  2,769,408    W:{5408,512}, b:{1,512}  [Subsampling layer]
Dense layer 1 1 (DenseLayer)  512,32    16,416       W:{512,32}, b:{1,32}    [Dense layer 2 1]
Output layer 2 (OutputLayer)  32,2      66           W:{32,2}, b:{1,2}      [Dense layer 1 1]
=====

Total Parameters:  2,786,210
Trainable Parameters:  2,786,210
Frozen Parameters:  0
```

```
=== Summary ===

Correctly Classified Instances      749          96.8952 %
Incorrectly Classified Instances    24           3.1048 %
Kappa statistic                    0.9163
Mean absolute error                 0.0339
Root mean squared error             0.1629
Relative absolute error             9.1062 %
Root relative squared error         37.6907 %
Total Number of Instances          773

=== Detailed Accuracy By Class ===

      TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
      -----  -----  -
      0.927    0.017    0.947     0.927    0.937      0.916    0.991    0.974     0
      0.983    0.073    0.976     0.983    0.979      0.916    0.991    0.997     1
Weighted Avg.   0.969    0.059    0.969     0.969    0.969      0.916    0.991    0.991

=== Confusion Matrix ===

  a  b  <-- classified as
178 14 |  a = 0
 10 571 |  b = 1
```

模型二:

兩層 convolution

Network Configuration:
NeuralNetConfiguration(weightInit=XAVIER, biasInit=0.0, dist=weka.dl4j.distribution.Disabled@66, l1=NaN, l2=NaN, dro
Model Summary:

VertexName (VertexType)	nIn,nOut	TotalParams	ParamsShape	Vertex Inputs
input (InputVertex)	-, -	-	-	-
Convolution layer 1 (ConvolutionLayer)	1, 32	320	W:{32,1,3,3}, b:{1,32}	[input]
Subsampling layer 1 (SubsamplingLayer)	-, -	0	-	[Convolution layer 1]
Convolution layer 2 (ConvolutionLayer)	32, 64	18,496	W:{64,32,3,3}, b:{1,64}	[Subsampling layer 1]
Subsampling layer 2 (SubsamplingLayer)	-, -	0	-	[Convolution layer 2]
Dense layer 2 1 (DenseLayer)	1600, 512	819,712	W:{1600,512}, b:{1,512}	[Subsampling layer 2]
Dense layer 1 1 (DenseLayer)	512, 32	16,416	W:{512,32}, b:{1,32}	[Dense layer 2 1]
Output layer 2 (OutputLayer)	32, 2	66	W:{32,2}, b:{1,2}	[Dense layer 1 1]

Total Parameters: 855,010
Trainable Parameters: 855,010
Frozen Parameters: 0

Time taken to test model on test split: 2.29 seconds

=== Summary ===

Correctly Classified Instances	743	96.119 %
Incorrectly Classified Instances	30	3.881 %
Kappa statistic	0.8961	
Mean absolute error	0.0555	
Root mean squared error	0.1846	
Relative absolute error	14.9064 %	
Root relative squared error	42.7167 %	
Total Number of Instances	773	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.922	0.026	0.922	0.922	0.922	0.896	0.984	0.933	0
	0.974	0.078	0.974	0.974	0.974	0.896	0.984	0.995	1
Weighted Avg.	0.961	0.065	0.961	0.961	0.961	0.896	0.984	0.979	

=== Confusion Matrix ===

a b <-- classified as

177	15	a = 0
15	566	b = 1

模型三:

VertexName (VertexType)	nIn,nOut	TotalParams	ParamsShape	Vertex Inputs
input (InputVertex)	-, -	-	-	-
Convolution layer 2 1 (ConvolutionLayer)	1, 32	320	W:{32,1,3,3}, b:{1,32}	[input]
Convolution layer 1 1 (ConvolutionLayer)	32, 64	18,496	W:{64,32,3,3}, b:{1,64}	[Convolution layer 2 1]
Subsampling layer 1 (SubsamplingLayer)	-, -	0	-	[Convolution layer 1 1]
Convolution layer 1 2 (ConvolutionLayer)	64, 128	73,856	W:{128,64,3,3}, b:{1,128}	[Subsampling layer 1]
Convolution layer 2 2 (ConvolutionLayer)	128, 256	295,168	W:{256,128,3,3}, b:{1,256}	[Convolution layer 1 2]
Subsampling layer 2 (SubsamplingLayer)	-, -	0	-	[Convolution layer 2 2]
Dense layer 2 1 (DenseLayer)	4096, 1024	4,195,328	W:{4096,1024}, b:{1,1024}	[Subsampling layer 2]
Dense layer 1 2 (DenseLayer)	1024, 128	131,200	W:{1024,128}, b:{1,128}	[Dense layer 2 1]
Dense layer 1 1 (DenseLayer)	128, 32	4,128	W:{128,32}, b:{1,32}	[Dense layer 1 2]
Output layer 2 (OutputLayer)	32, 2	66	W:{32,2}, b:{1,2}	[Dense layer 1 1]

```

=== Summary ===

Correctly Classified Instances      581      75.1617 %
Incorrectly Classified Instances    192      24.8383 %
Kappa statistic                     0
Mean absolute error                 0.3746
Root mean squared error             0.4321
Relative absolute error             100.5413 %
Root relative squared error         100.001 %
Total Number of Instances          773

=== Detailed Accuracy By Class ===

                TP Rate  FP Rate  Precision  Recall   F-Measure  MCC      ROC Area  PRC Area  Class
                0.000   0.000   ?           0.000   ?           ?       0.500    0.248    0
                1.000   1.000   0.752       1.000   0.858       ?       0.500    0.752    1
Weighted Avg.   0.752   0.752   ?           0.752   ?           ?       0.500    0.627

=== Confusion Matrix ===

  a  b  <-- classified as
  0 192 |  a = 0
  0 581 |  b = 1

```

評估報告

分析結果，需有圖表統整同一個模型不同超參數的表現以及各模型的表現

1. 模型比較

	RandomForest	MLP	CNN
accuracy	97.3615%	96.5699%	96.8952

2. Random Forest

numberFeatures	accuracy
0	97.2615%
5	97.2296%
11	97.2735%

3. MLP

hidden Layer	accuracy
2	96.5699%
10	96.394%
10, 2	96.0422%

4. CNN

	accuracy
模型一	96.8952%
模型二	96.119%
模型三	75.1617%

- 使用的方法的優缺點

1. RandomForest 好處是執行快速且準確率高，但可讀性差。
2. MLP 則是簡單好懂，雖然 hidden layer 沒設很高，但執行的效果還算不錯，但 hidden layer 設太高，電腦執行時間久。
3. CNN 準確率不錯、參數共享，但執行的時間久，且容易 overfitting。

- 是否有改進之處

前處理的部分，可以嘗試刪除多個欄位，嘗試找到最適合的組合。

雖然這次實驗 CNN 模型複雜，對準確度沒有太大的幫助，反而造成準確度下降的問題。推測這可能跟資料量和 overfitting 有關。因此推測可以使用 Dropout 或 regularization 等技術，來避免 overfitting。

- 最後會選擇哪一個模型

最後選擇 RandomForest。考慮到電腦的執行時間，以及準確度，RandomForest 訓練出的結果明顯優於 MLP 和 CNN。