

神經網路又稱為機器學習

決定目標(自變量, 應變量) → 資料清洗 →

模型建立 → 模型診斷

類神經網路優點: 無限制, 效率高, 效率高

缺點: 黑盒子

① 套件匯入

```
import pandas as pd
import numpy as np
from sklearn import datasets, preprocessing
from sklearn.model_selection import train_test_split
from keras.models import Sequential
from keras.layers import Dense, Dropout
from keras.optimizers import SGD, Adam
import matplotlib.pyplot as plt
```

② 叫出iris資料

```
iris = datasets.load_iris()
df = pd.DataFrame(iris['data'], columns=iris['feature_names'])
```

③ 標準化 ⇒ 將單位去除

```
min_max_scaler = preprocessing.MinMaxScaler()
df =
pd.DataFrame(min_max_scaler.fit_transform(df), columns=iris['feature_names'])
```

④ 挑出應變數和自變數

```
labels = np.array(df['sepal length (cm)']) # 應變數
features = df.drop('sepal length (cm)', axis = 1) # 自變數
```

⑤ 隨機抽樣

```
trainX, testX, trainY, testY = train_test_split(features, labels, test_size = 0.3,  
random_state = 42)
```

```
trainY = trainY.reshape(-1, 1) #創行
```

```
testY = testY.reshape(-1, 1) #創行
```

⑥ 建模

```
model = Sequential()  
model.add(Dense(64, input_dim=3, activation='relu'))  
model.add(Dropout(0.5))  
model.add(Dense(1)) #類別型, activation='softmax'  
model.compile(loss='mse', #類別型'categorical_crossentropy'  
optimizer=SGD(lr=0.1), #adam  
metrics=['mse', 'mape']) #類別型'accuracy'  
#%%
```

過度配適

激活函數

可解決梯度
消失的
問題

只保留0.5

隱藏層中神經元的數量

學習率=0.1

自变量

很容易造成局部
最小值

⑦ 模型訓練

```
dnn = model.fit(trainX, trainY, epochs=20, batch_size=30)
```

訓練自变量

訓練隱變量

每次抓30資
料出來擬合

共做20次

一次

⑧ 预测

```
predictions = model.predict(testX)
```

```
predictions = predictions.reshape(-1, 1)
```

```
np.mean(np.abs((testY - predictions))) #mae
```

```
np.mean(np.abs((testY - predictions) / testY)) * 100 #mape
```

看跟前一位预测结果近似