利用ESP32 NodeMCU-32S Edge Al Model 判斷洗腎廔管異常情形

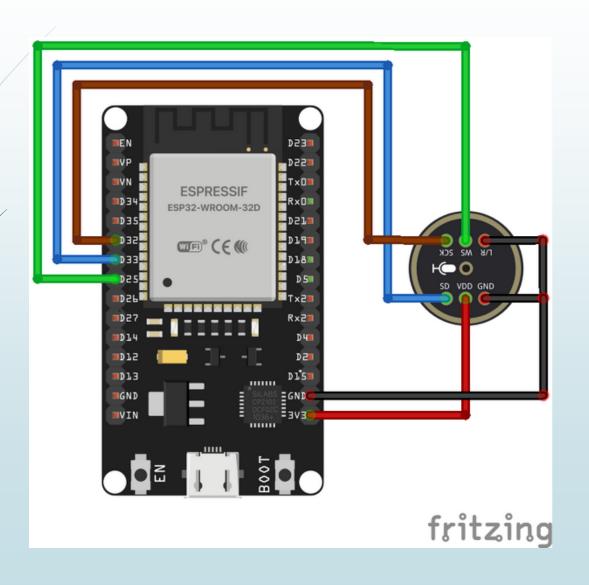
Richard Chiu

設備 NodeMCU-32S

/
NodeMCU-32
25. 4mm (W) *48. 3mm (H) ±0.2 mm
DIP-38 (2.54 间距标准排针)
默认 32Mbits
UART/SPI/SDIO/I2C/PWM/I2S/IR/ADC/DAC
2400~2483. 5MHz
支持 300 ~ 4608000 bps , 默认 115200 bps
蓝牙 4.2 BR/EDR 和 BLE 标准
默认 32Mbit,最大支持 128Mbit
-20℃ ~ 70 ℃
-40 °C ~ 125 °C , < 90%RH
Micro USB 供电电压 4.75V~5.25V, 推荐 5.0V 供电电压 3.0V ~ 3.6V, 供电电流 >500mA, 推荐 3.3V



設備 INMP441



ESP32 套件

■ Arduino IDE

主程式區段

- ► FFT -> 128 維特徵
- 128D feature -> 隨機森林
- ▶ 輸出結果

```
while (wavFile.available()) {
271
             // 讀取音頻樣本
             readSamples(wavFile, FFT SIZE, vReal, vImag, audioFormat);
             FFT->windowing(FFTWindow::Hamming, FFTDirection::Forward);
             FFT->compute(FFTDirection::Forward);
             FFT->complexToMagnitude();
             // 保存FFT結果到文本文件
             for (int i = 0; i < FEATURE SIZE; i++) {
                 char formattedValue[10];
                 dtostrf(vReal[i], 1, 6, formattedValue); // 保留6位小數
283
                 featureFile.print(formattedValue);
                 if (i < FEATURE SIZE - 1) featureFile.print(", ");</pre>
             featureFile.println();
             // 準備進行預測
             saveFeatures(vReal, rfFeatures);
             // 使用隨機森林模型進行預測
             int prediction = predictSample(rfFeatures);
             positiveCount += prediction;
             sampleCount++;
         featureFile.close();
```

訓練資料

- 醫生臨床錄製
- 醫生判斷標籤
- 經 ESP32 FFT 轉換輸出成 data
- ► 於 windows python 環境下訓練
- ML Random forest model



訓練資料

```
C:\Develop\TMI\ESP32\vessel_data\normal\1640575433.txt - Notepad++
檔案(F) 編輯(E) 搜尋(S) 檢視(V) 編碼(N) 語言(L) 設定(T) 工具(O) 巨集(M) 執行(R) 外掛(P) 視窗(W) ?
1640575433.txt 
  1 0.114377, 0.132828, 0.735810, 1.221939, 0.477068, 0.385012, 1.347589, 3.097064, 1.614232, 1.246602, 1.773712, 2.183250, 5.444429, 2.924836, 3.493387, 7.857769, 2.445215, 1.771497,
      2.756210, 8.449602, 13.672212, 4.749329, 4.534157, 4.131658, 8.008426, 7.191376, 2.879943, 5.495909, 3.476709, 1.935445, 1.004915, 0.893154, 1.469310, 1.850733, 2.240760, 1.223796,
      0.991081, 0.156202, 2.157674, 2.125926, 0.589142, 0.908581, 1.443145, 1.325183, 1.500197, 0.576297, 1.033701, 0.500662, 0.917733, 1.818385, 0.756672, 0.879293, 0.498722, 0.072002,
      0.450896, 0.258186, 0.473803, 0.651206, 0.466009, 0.343927, 0.239230, 0.387077, 0.515672, 0.537039, 0.277701, 0.393437, 0.296962, 0.084637, 0.135677, 0.211972, 0.168733, 0.152093,
      0.138125, 0.116879, 0.082999, 0.069237, 0.038599, 0.041380, 0.020566, 0.039254, 0.018766, 0.023761, 0.027665, 0.023737, 0.024624, 0.021774, 0.024219, 0.022839, 0.022154, 0.023065,
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      0.004815, 0.004785
    0.047189, 0.079067, 0.295823, 0.205010, 0.394024, 0.408188, 1.699581, 0.841169, 2.712437, 3.068188, 2.778708, 1.140428, 1.924719, 1.149688, 2.011432, 1.511345, 2.677512, 1.764025,
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      0.006602, 0.006552, 0.006502, 0.006454, 0.006406, 0.006360, 0.006314, 0.006269, 0.006224, 0.006181, 0.006138, 0.006096, 0.006056, 0.006015, 0.005975, 0.005937, 0.005898, 0.005861,
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      0.321219, 0.365518, 0.267837, 0.152396, 0.309232, 0.335301, 0.204778, 0.351338, 0.172949, 0.077017, 0.034174, 0.073463, 0.115427, 0.057611, 0.056826, 0.031008, 0.065475, 0.045297,
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      0.006628, 0.006586
```

```
n_estimators = 8
max_depth = 10

# Train Random Forest model
rf_clf = RandomForestClassifier(n_estimators=n_estimators, max_depth=max_depth, random_state=42) #
rf_clf.fit(X_train, y_train)

# Predict on test set
y_pred = rf_clf.predict(X_test)

# calculate accuracy
accuracy = accuracy_score(y_test, y_pred)

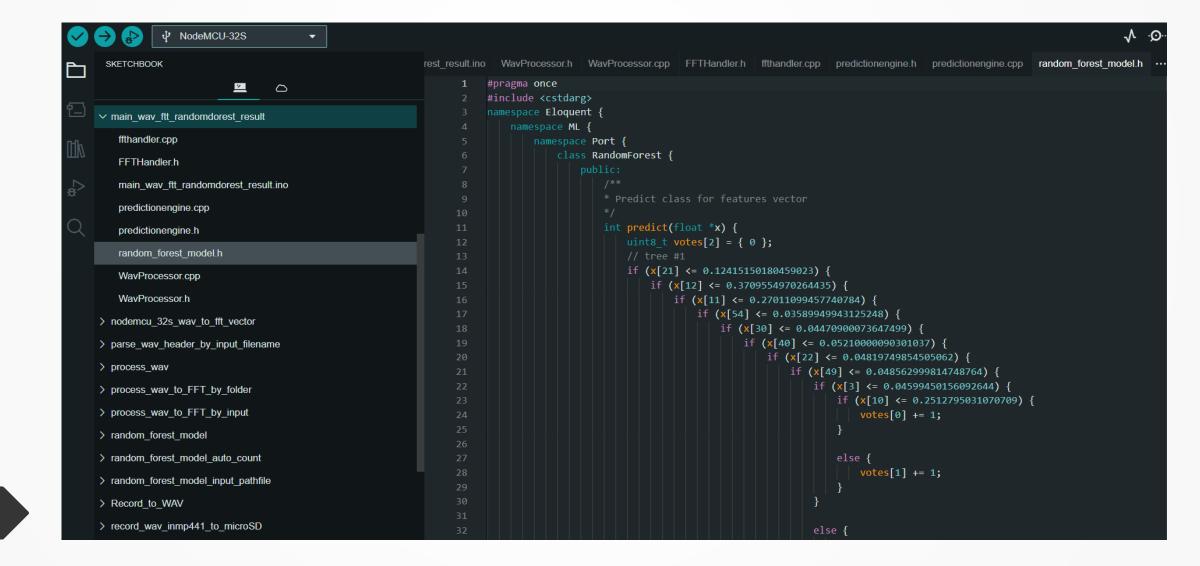
print(f"Model accuracy: {accuracy:.2f} {n_estimators} {max_depth}")

# Save Random Forest model weights
joblib.dump(rf_clf, 'random_forest_model.pkl')

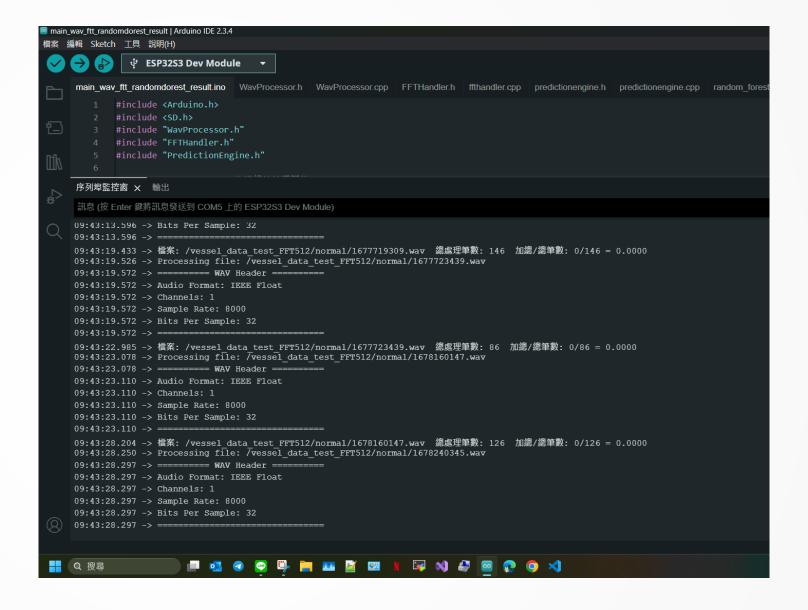
w Model accuracy: 0.91 8 10

"" ['random_forest_model.pkl']
```

模型轉換



模型輸出



開發設備





