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From: What is today's lunch?

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Summary

Feature extraction using Librosa was conducted to utilize Machine Learning models to classify payload. The new datasets collected in the data collection trip were used to extract features. Also, Machine Learning with prior dataset was conducted. NN(Neural Network), KNN(K Nearest Neighborhood), GNB(Gaussian Naïve Bayes), and SVM(Support Vector Machine) were utilized.

Deep Learning(DL) including Convolutional Neural Network(CNN) and RNN(Recurrent Neural Network) was employed as well. Furthermore, the combination of five features was applied in CNN and RNN models using data from a prior study.

What 'What is today's lunch?' completed this week:

- **Feature extraction with new dataset**

- Feature extraction (MFCCs, Mel, Chroma, Tonnetz, Contrast)

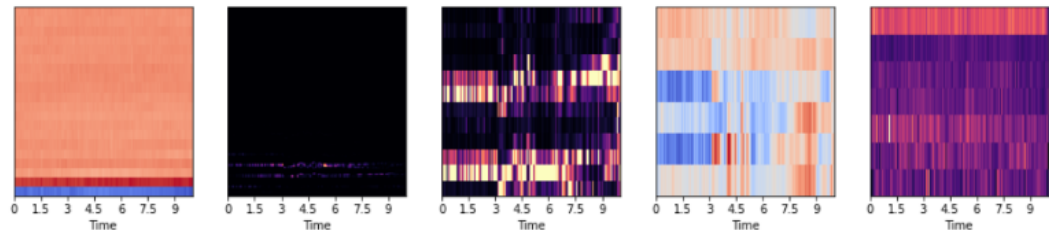


Figure 1. Feature extraction (2 payload)

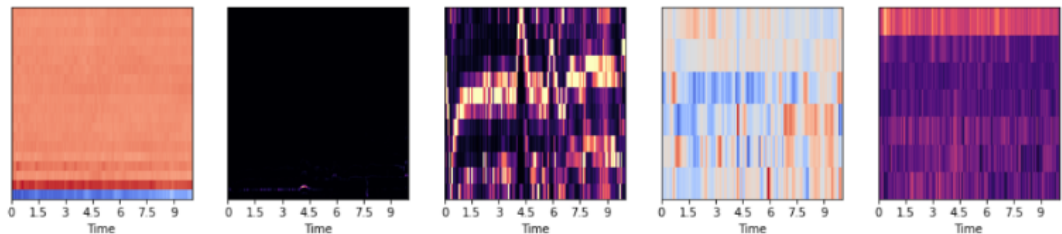


Figure 2. Feature extraction (1 payload)

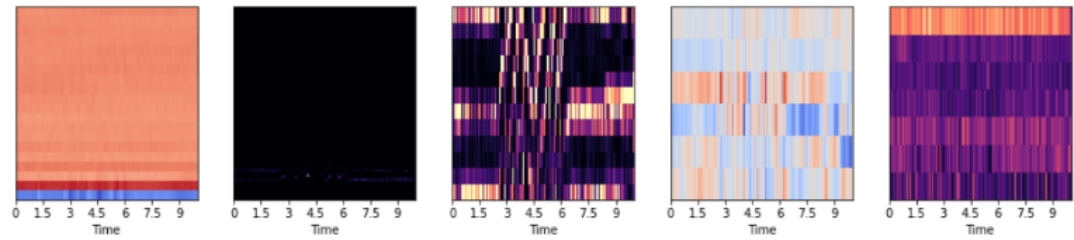


Figure 3. Feature extraction (0 payload)

Feature extraction with new datasets (2payload, 1payload, without payload) was conducted. MFCCs, Mel, Chroma, Tonnetz, and Contrast will be utilized to train and test Machine Learning models.

- Machine Learning with prior dataset
 - NN, KNN, GNB, SVM with MFCC
 - ◆ NN

```

1 # split the dataset
2 from sklearn.model_selection import train_test_split
3 x_train, x_test, y_train, y_test = train_test_split(X, yy, test_size=0.2, random_state = 127)

1 # neural network architecture
2 num_labels = yy.shape[1]
3 filter_size = 2
4
5 model = Sequential()
6 model.add(Dense(256))
7 model.add(Activation('relu'))
8 model.add(Dropout(0.5))
9 model.add(Dense(256))
10 model.add(Activation('relu'))
11 model.add(Dropout(0.5))
12 model.add(Dense(num_labels))
13 model.add(Activation('softmax'))
14 # Compile the model
15 model.compile(loss='categorical_crossentropy', metrics=['accuracy'], optimizer='adam')
16
17 model.build(input_shape=(None, 40))
18 # Display model architecture summary
19 model.summary()
20

1 from keras.callbacks import ModelCheckpoint
2 from datetime import datetime
3 num_epochs = 100
4 num_batch_size = 32
5
6 history=model.fit(x_train, y_train, batch_size=num_batch_size, epochs=num_epochs, validation_data=(x_test, y_test), verbose=1)

1 # Evaluating the model on the training and testing set
2 score = model.evaluate(x_train, y_train, verbose=0)
3 print("Training Accuracy: {}".format(score[1]))
4 score = model.evaluate(x_test, y_test, verbose=0)
5 print("Testing Accuracy: {}".format(score[1]))

Training Accuracy: 100.00%
Testing Accuracy: 100.00%

```

◆ KNN (99%)

```
1 from sklearn.model_selection import train_test_split
2 training_data, validation_data, training_labels, validation_labels = train_test_split(X, yy, test_size = 0.2, random_state = 100)

1 from sklearn.neighbors import KNeighborsClassifier
2 classifier = KNeighborsClassifier(n_neighbors = 3)

1 classifier.fit(training_data, training_labels)

KNeighborsClassifier(n_neighbors=3)

1 print(classifier.score(validation_data, validation_labels))

0.9972602739726028
```

◆ GNB (100%)

```
1 from sklearn.model_selection import train_test_split
2 x_train, x_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 42)

1 from sklearn.naive_bayes import GaussianNB
2 nb = GaussianNB()
3 nb.fit(x_train, y_train)

GaussianNB()

1 print("accuracy: ",nb.score(x_test, y_test))

accuracy: 1.0
```

◆ SVM (linear kernel: 99%, rbf kernel : 98%)

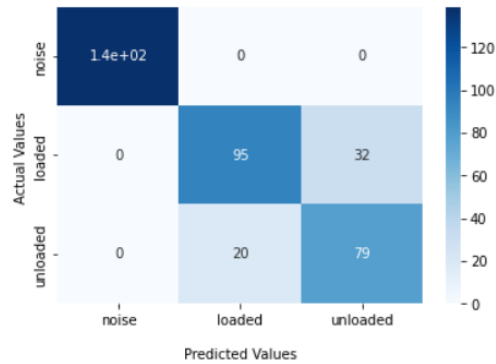
```
1 # Data split
2
3 X_train, X_test, y_train, y_test = train_test_split(X, yy, test_size = 0.3, random_state = 100)
4 import sklearn.svm as svm
5 import sklearn.metrics as mt
6 from sklearn.model_selection import cross_val_score, cross_validate
7
8 # SVM, kernel = 'linear'
9 svm_clf = svm.SVC(kernel = 'linear', random_state=100)
10
11 # cross validation
12 scores = cross_val_score(svm_clf, X, y, cv = 5)
13 scores
14
15 pd.DataFrame(cross_validate(svm_clf, X, y, cv =5))
16
17 print('cross validation average(linear kernel): ', scores.mean())
18
19 # SVM, kernel = 'rbf'
20 svm_clf = svm.SVC(kernel = 'rbf')
21
22 # cross validation
23 scores = cross_val_score(svm_clf, X, y, cv = 5)
24 scores
25
26 pd.DataFrame(cross_validate(svm_clf, X, y, cv =5))
27
28 print('cross validation average(rbf kernel): ', scores.mean())

cross validation average(linear kernel): 0.9928767123287671
cross validation average(rbf kernel): 0.9873957549300016
```

● Deep Learning with prior dataset

- ◆ CNN using 1dimensional convolutional layer and MFCCs (86%)
https://colab.research.google.com/drive/1uWFH6BrzBMzT3XElr84_p37rV4lXCq2e?usp=sharing

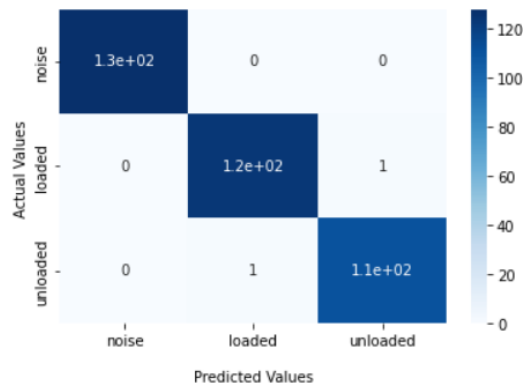
UAV Payload Detection with 1dimensional convolutional layer



Accuracy: 0.86, Total items: 365

- ◆ CNN using 2dimensional convolutional layer and MFCCs (99%)
https://colab.research.google.com/drive/1uWFH6BrzBMzT3XElr84_p37rV4lXCq2e?usp=sharing

UAV Payload Detection with 2dimensional convolutional layer



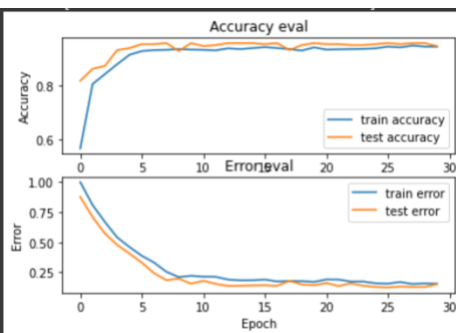
Accuracy: 0.99, Total items: 365

- ◆ RNN with a combination of features. (93.1%)
 Each feature was extracted with the same length and different parts from raw data.

Model: "sequential_8"

Layer (type)	Output Shape	Param #
lstm_16 (LSTM)	(None, 431, 64)	18176
lstm_17 (LSTM)	(None, 64)	33024
dense_16 (Dense)	(None, 64)	4160
dropout_8 (Dropout)	(None, 64)	0
dense_17 (Dense)	(None, 3)	195

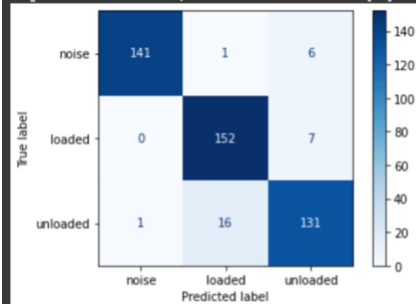
=====
Total params: 55,555
Trainable params: 55,555
Non-trainable params: 0



15/15 [=====] - 2s 103ms/step - loss: 0.1589 - accuracy: 0.9319

Accuracy on test set is : 0.9318681359291077

Expected index: 2, Predicted index: [2]



f1 score: 0.9321677784562447

precision_score: 0.9340294175749214

recall_score: 0.9312708935350446

Things to do by next week

- Going on a data collection trip with Mia.

Problems or challenges:

- Further Data collection would be conducted in the future.

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