

Gravitational Wave Detection

CMPE 257 Project
September 9, 2021

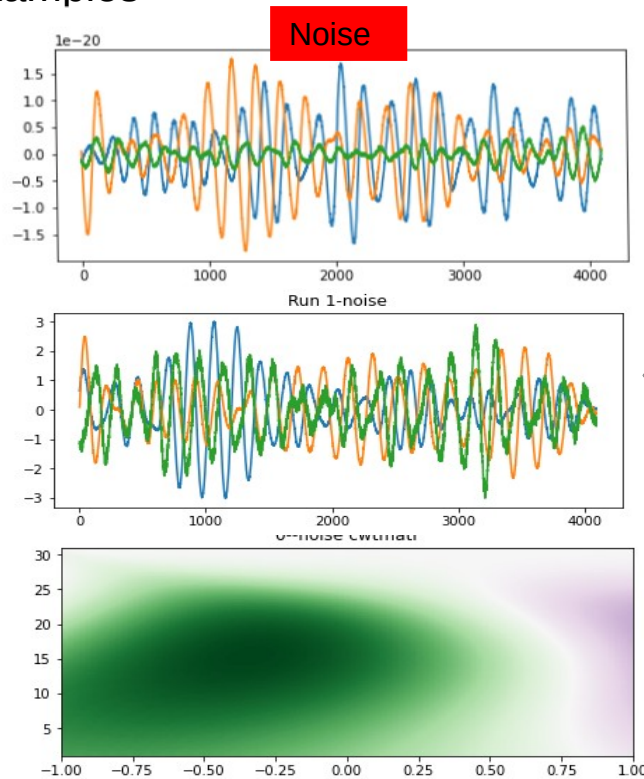
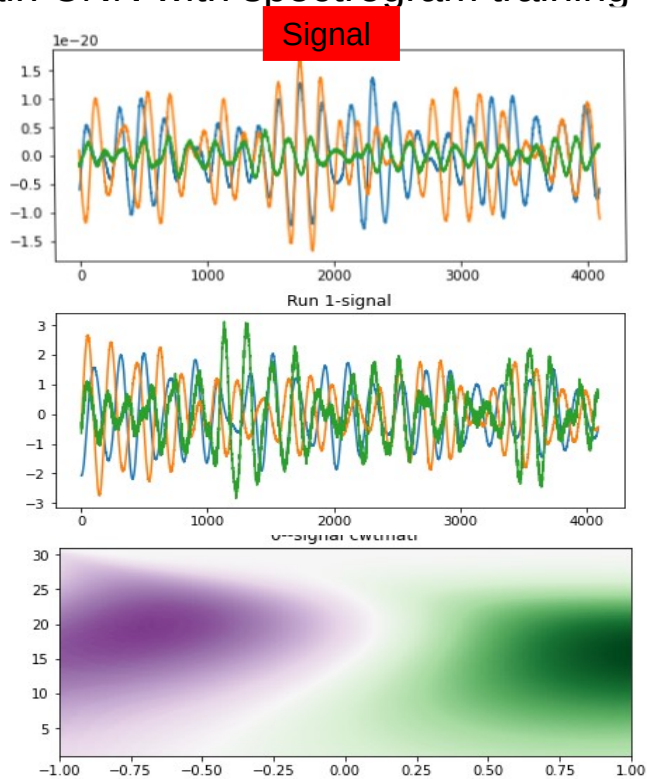
Tom Casaletto
Anthony Fisher
Phil Shirts
Joel Wiser

Outline

- Problem Statement
- CWT/CNN Approach
- CQT/CNN Approach
- Transfer Learning
- Next Steps
- Additional Material

Continuous Wavelet Transform

- Transform detector time series into spectrograms (images)
- Train CNN with spectrogram training examples

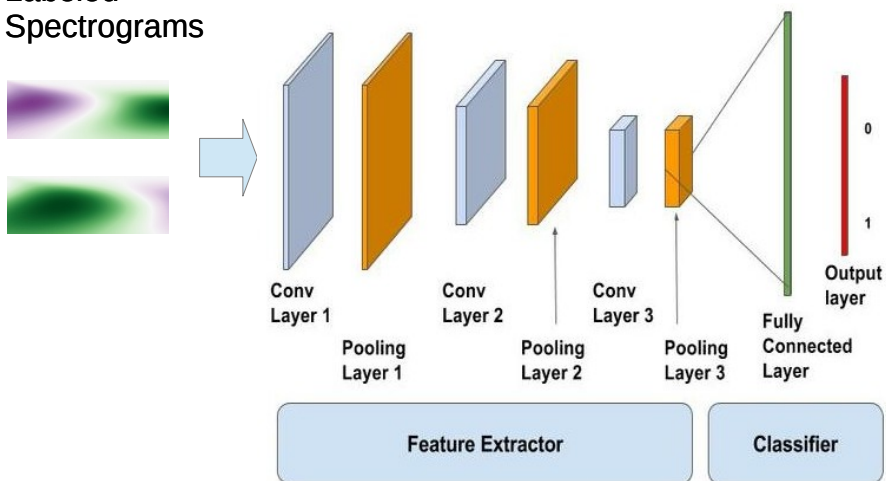


Normalize signals
(z transform)

Perform CWT transform
Apply bandpass filter
(20-500 Hz)

Convolutional Neural Net

Labeled Spectrograms



```
model = models.Sequential()
model.add(layers.Conv2D(32, (10, 10), activation='relu', input_shape=(300,97,1)))
model.add(layers.MaxPooling2D((6, 6)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.MaxPooling2D((2, 2)))
model.add(layers.Conv2D(64, (3, 3), activation='relu'))
model.add(layers.Flatten())
model.add(layers.Dense(64, activation='relu'))
model.add(layers.Dense(2))
model.summary()

model.compile(optimizer='adam',
              loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
              metrics=['accuracy'])
model.fit(X_train,y_train,epochs=10)
```

CWT/CNN Results

Model: "sequential_13"

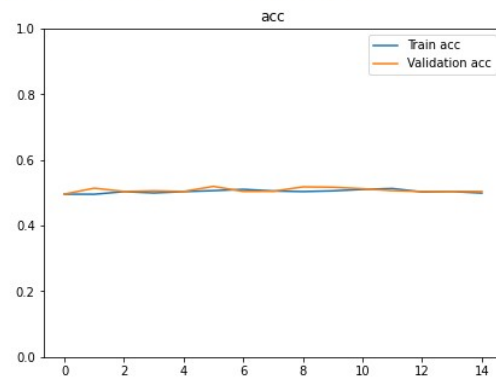
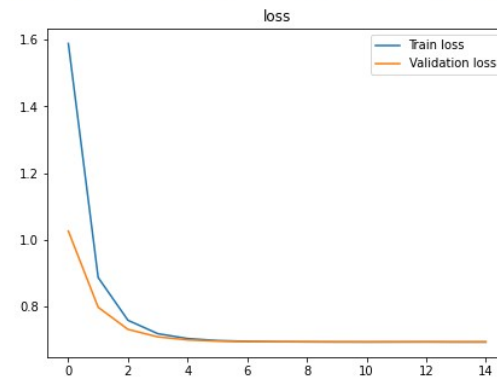
Layer (type)	Output Shape	Param #
conv2d_16 (Conv2D)	(None, 291, 88, 32)	3232
max_pooling2d_11 (MaxPooling (None, 48, 14, 32))		0
conv2d_17 (Conv2D)	(None, 46, 12, 64)	18496
max_pooling2d_12 (MaxPooling (None, 23, 6, 64))		0
conv2d_18 (Conv2D)	(None, 21, 4, 64)	36928
flatten_8 (Flatten)	(None, 5376)	0
dense_16 (Dense)	(None, 64)	344128
dense_17 (Dense)	(None, 2)	130

Total params: 402,914
Trainable params: 402,914
Non-trainable params: 0

Epoch 1/10
188/188 [=====] - 210s 1s/step - loss: 0.7200 - accuracy: 0.5025 - val_loss: 0.6932 - val_accuracy: 0.5040
Epoch 2/10
188/188 [=====] - 199s 1s/step - loss: 0.6959 - accuracy: 0.5040 - val_loss: 0.6923 - val_accuracy: 0.5100
Epoch 3/10
188/188 [=====] - 199s 1s/step - loss: 0.6940 - accuracy: 0.5002 - val_loss: 0.6931 - val_accuracy: 0.5040
Epoch 4/10
188/188 [=====] - 204s 1s/step - loss: 0.6932 - accuracy: 0.5035 - val_loss: 0.6931 - val_accuracy: 0.5040
Epoch 5/10
188/188 [=====] - 200s 1s/step - loss: 0.6932 - accuracy: 0.5042 - val_loss: 0.6931 - val_accuracy: 0.5030
Epoch 6/10
188/188 [=====] - 200s 1s/step - loss: 0.6933 - accuracy: 0.4997 - val_loss: 0.6932 - val_accuracy: 0.5030
Epoch 7/10
188/188 [=====] - 200s 1s/step - loss: 0.6932 - accuracy: 0.4997 - val_loss: 0.6931 - val_accuracy: 0.5025
Epoch 8/10
188/188 [=====] - 200s 1s/step - loss: 0.6932 - accuracy: 0.5005 - val_loss: 0.6931 - val_accuracy: 0.5035
Epoch 9/10
188/188 [=====] - 200s 1s/step - loss: 0.6936 - accuracy: 0.5010 - val_loss: 0.6933 - val_accuracy: 0.4960
Epoch 10/10
188/188 [=====] - 199s 1s/step - loss: 0.6932 - accuracy: 0.4995 - val_loss: 0.6931 - val_accuracy: 0.5040

```
model2.evaluate(X_test,y_test)
show_final_history(history2)
```

63/63 [=====] - 17s 268ms/step - loss: 0.6931 - accuracy: 0.5050



Backup

CWT/CNN Results

1000 examples, 80/20 train/test split

Model: "sequential_13"

Layer (type)	Output Shape	Param #
conv2d_16 (Conv2D)	(None, 291, 88, 32)	3232
max_pooling2d_11 (MaxPooling (None, 48, 14, 32)		0
conv2d_17 (Conv2D)	(None, 46, 12, 64)	18496
max_pooling2d_12 (MaxPooling (None, 23, 6, 64)		0
conv2d_18 (Conv2D)	(None, 21, 4, 64)	36928
flatten_8 (Flatten)	(None, 5376)	0
dense_16 (Dense)	(None, 64)	344128
dense_17 (Dense)	(None, 2)	130

Total params: 402,914

Trainable params: 402,914

Non-trainable params: 0

Epoch 1/10

3/3 [=====] - 1s 107ms/step - loss: 0.9183 - accuracy: 0.3750

Epoch 2/10

3/3 [=====] - 0s 23ms/step - loss: 0.7564 - accuracy: 0.5500

Epoch 3/10

3/3 [=====] - 0s 25ms/step - loss: 0.6912 - accuracy: 0.5875

Epoch 4/10

3/3 [=====] - 0s 24ms/step - loss: 0.6665 - accuracy: 0.6250

Epoch 5/10

3/3 [=====] - 0s 24ms/step - loss: 0.6587 - accuracy: 0.6375

Epoch 6/10

3/3 [=====] - 0s 28ms/step - loss: 0.6100 - accuracy: 0.6625

Epoch 7/10

3/3 [=====] - 0s 24ms/step - loss: 0.5969 - accuracy: 0.6875

Epoch 8/10

3/3 [=====] - 0s 23ms/step - loss: 0.5981 - accuracy: 0.6875

Epoch 9/10

3/3 [=====] - 0s 25ms/step - loss: 0.5399 - accuracy: 0.7125

Epoch 10/10

3/3 [=====] - 0s 24ms/step - loss: 0.5107 - accuracy: 0.6875

Leaderboard [as of 2021-09-08]

European Gravitational Observatory - EGO · 1,000 teams · 22 days to go (15 days to go until merger deadline)

Overview

Data

Code

Discussion

Leaderboard

Rules

Team

My Submissions

Submit Predictions

Public Leaderboard

Private Leaderboard

This leaderboard is calculated with approximately 16% of the test data.

The final results will be based on the other 84%, so the final standings may be different.

In the money

Gold

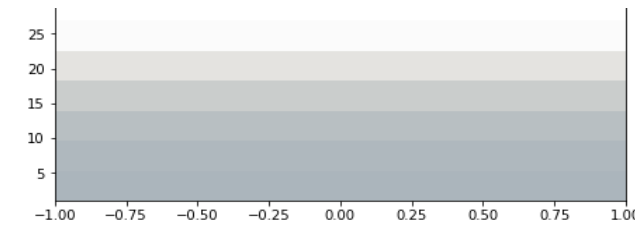
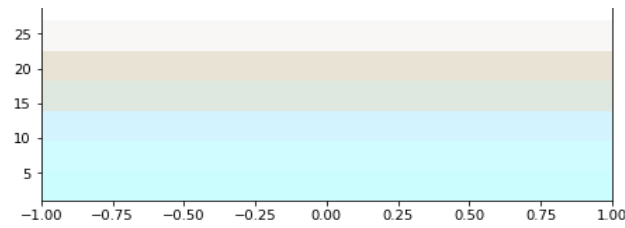
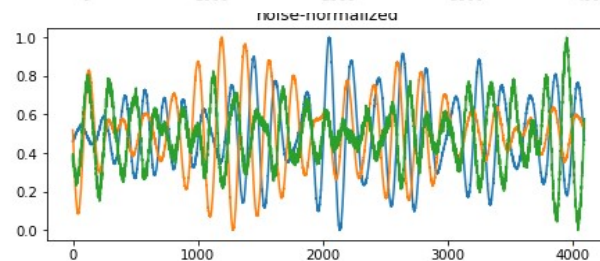
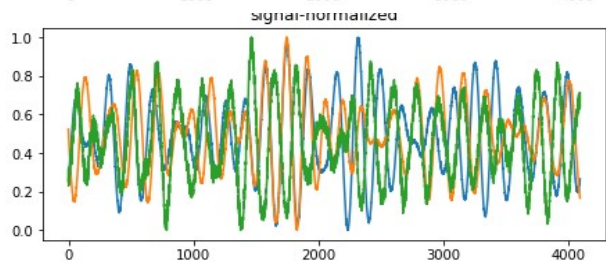
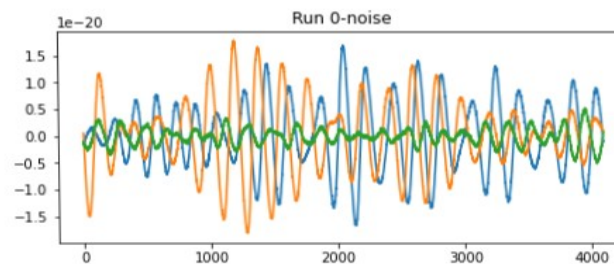
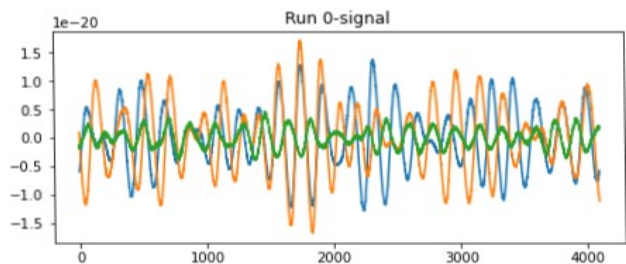
Silver

Bronze

#	Team Name	Notebook	Team Members	Score	Entries	Last
1	KDL			0.8826	17	20h
2	[Aillis.jp] KUMA300			0.8823	124	2h
3	[NVIDIA] KAGRA			0.8812	156	4h
4	Yasuhisa Nakaism			0.8808	88	10h
5	MILIMED			0.8805	33	7h
6	got Sputnik but vaccine not sp...			0.8805	112	15h
7	F.J.Martinez-de-Pison			0.8803	72	7m

Continuous Wavelet Transform

- Transform detector time series into spectrograms (images)
- Train CNN with spectrogram training examples

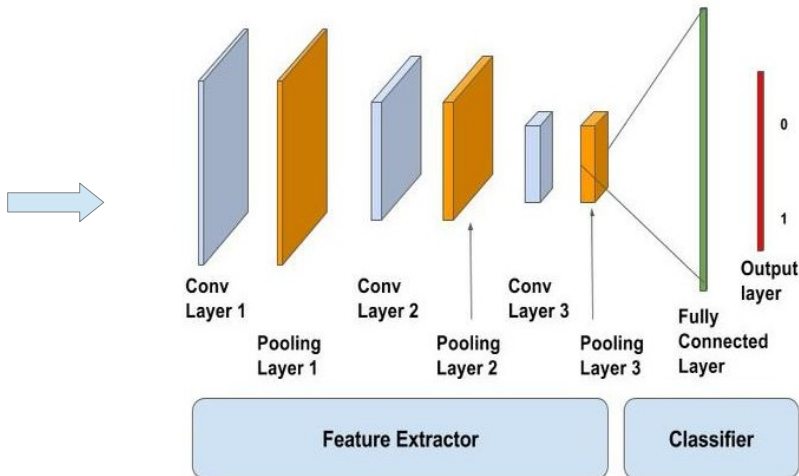
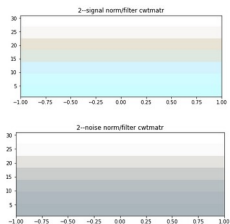


Normalize signals
(z transform)

Perform CWT transform
Apply bandpass filter

Convolutional Neural Net

Labeled Spectrograms



```
model=Sequential()
model.add(Conv2D(32,(3,3),activation='relu',input_shape=(21,4,1)))
model.add(MaxPool2D(2,2))
model.add(Flatten())
model.add(Dense(100,activation='relu'))
model.add(Dense(2,activation='softmax'))
model.compile(loss='sparse_categorical_crossentropy',optimizer='adam',metrics=['accuracy'])
```

CWT/CNN Results

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 19, 2, 32)	320
max_pooling2d (MaxPooling2D)	(None, 9, 1, 32)	0
flatten (Flatten)	(None, 288)	0
dense (Dense)	(None, 100)	28900
dense_1 (Dense)	(None, 2)	202

Total params: 29,422

Trainable params: 29,422

Non-trainable params: 0

Epoch 1/10

250/250 [=====] - 2s 3ms/step - loss: 0.6943 - accuracy: 0.5056

Epoch 2/10

250/250 [=====] - 1s 3ms/step - loss: 0.6937 - accuracy: 0.5006

Epoch 3/10

250/250 [=====] - 1s 3ms/step - loss: 0.6933 - accuracy: 0.5046

Epoch 4/10

250/250 [=====] - 1s 3ms/step - loss: 0.6933 - accuracy: 0.5048

Epoch 5/10

250/250 [=====] - 1s 3ms/step - loss: 0.6931 - accuracy: 0.5001

Epoch 6/10

250/250 [=====] - 1s 3ms/step - loss: 0.6931 - accuracy: 0.4969

Epoch 7/10

250/250 [=====] - 1s 3ms/step - loss: 0.6931 - accuracy: 0.4969

Epoch 8/10

250/250 [=====] - 1s 3ms/step - loss: 0.6932 - accuracy: 0.5027

Epoch 9/10

250/250 [=====] - 1s 3ms/step - loss: 0.6929 - accuracy: 0.4967

Epoch 10/10

250/250 [=====] - 1s 3ms/step - loss: 0.6929 - accuracy: 0.5041

Creating Spectrogram

