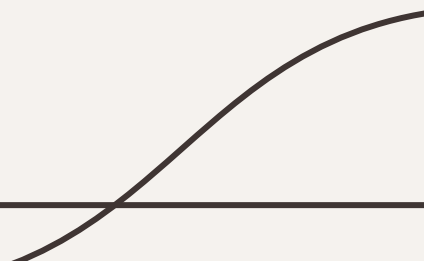




Emotion Recognition from Speech

A classification model utilizing the RAVDESS data set

Samantha Gonzales, Savannah Stewart, Bhagavan Bejjipurapu, Aditya Naredla



The image features two thin, dark horizontal lines spanning the width of the page. The top line has a decorative curve on its left end, and the bottom line has a decorative curve on its right end.

Processing

Data Set

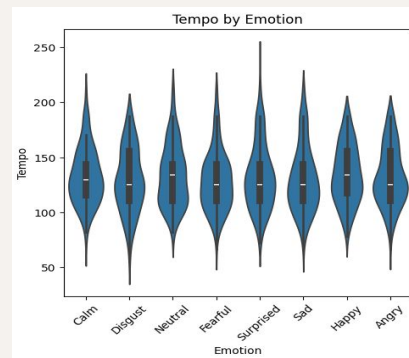
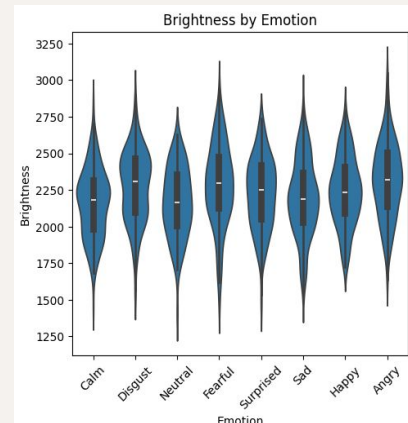
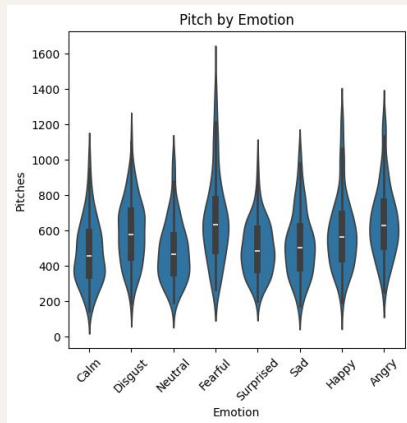
- Data set: Ryerson Audio-Visual Database of Emotional Speech and Song (RAVDESS)
- Available through: [Zenodo](#)
- The data set provided the following:
 - Emotion
 - Emotional Intensity
 - Statement
 - Repetition
 - Actor ID
 - Gender
 - Modality
 - Voice Channel

	Modality	Vocal Channel	Emotion	Intensity	Statement	Repetition	Actor ID	Gender	File Path
0	Audio-only	Speech	Calm	Strong	Kids are talking by the door	1	7	Male	./ravdess_data/Actor_07/03-01-02-02-01-01-07.wav
1	Audio-only	Speech	Disgust	Normal	Dogs are sitting by the door	2	7	Male	./ravdess_data/Actor_07/03-01-07-01-02-02-07.wav
2	Audio-only	Speech	Neutral	Normal	Dogs are sitting by the door	2	7	Male	./ravdess_data/Actor_07/03-01-01-01-02-02-07.wav
3	Audio-only	Speech	Fearful	Normal	Kids are talking by the door	2	7	Male	./ravdess_data/Actor_07/03-01-06-01-01-02-07.wav
4	Audio-only	Speech	Surprised	Strong	Kids are talking by the door	1	7	Male	./ravdess_data/Actor_07/03-01-08-02-01-01-07.wav
5	Audio-only	Speech	Calm	Normal	Kids are talking by the door	1	7	Male	./ravdess_data/Actor_07/03-01-02-01-01-01-07.wav
6	Audio-only	Speech	Calm	Normal	Dogs are sitting by the door	2	7	Male	./ravdess_data/Actor_07/03-01-02-01-02-02-07.wav
7	Audio-only	Speech	Sad	Normal	Dogs are sitting by the door	2	7	Male	./ravdess_data/Actor_07/03-01-04-01-02-02-07.wav
8	Audio-only	Speech	Disgust	Normal	Dogs are sitting by the door	1	7	Male	./ravdess_data/Actor_07/03-01-07-01-02-01-07.wav
9	Audio-only	Speech	Fearful	Normal	Kids are talking by the door	1	7	Male	./ravdess_data/Actor_07/03-01-06-01-01-01-07.wav

Rows 0 - 9 of dataframe from original data set - no feature extraction

Feature Extraction

- **Mel-Frequency Cepstral Coefficients (MFCC):** Captures timbral features
- **Mel-Spectrogram:** Captures frequency content of audio signals over time
- **Tempo:** Estimates the tempo by beats per minute
- **Spectral Centroid (Brightness):** Captures where the energy of sound is concentrated in a frequency
- **Pitch:** Captures pitch and its magnitude in audio signals over time



Feature extraction by Emotion for: Pitch, Brightness, & Tempo



Classification Models

Random Forest

```
forest_classifier = RandomForestClassifier(max_features=6,  
                                         random_state=42,  
                                         n_estimators=700,  
                                         criterion='gini',  
                                         max_depth=None)  
  
forest_classifier.fit(X_train, y_train) # train the model  
y_pred = forest_classifier.predict(X_test) # test the model  
  
y_test_labels = emotion_encoder.inverse_transform(y_test)  
y_pred_labels = emotion_encoder.inverse_transform(y_pred)
```

- Step 1
 - Encode categorical variables into numerical variables for use in model
- Step 2
 - Split into train and test groups
 - Used 80/20 ratio for train and test
- Step 3
 - Define hyperparameters for model
 - Used RandomizedSearchCV to find optimal parameters with our training data
- Step 4
 - Train the model
 - Test the model

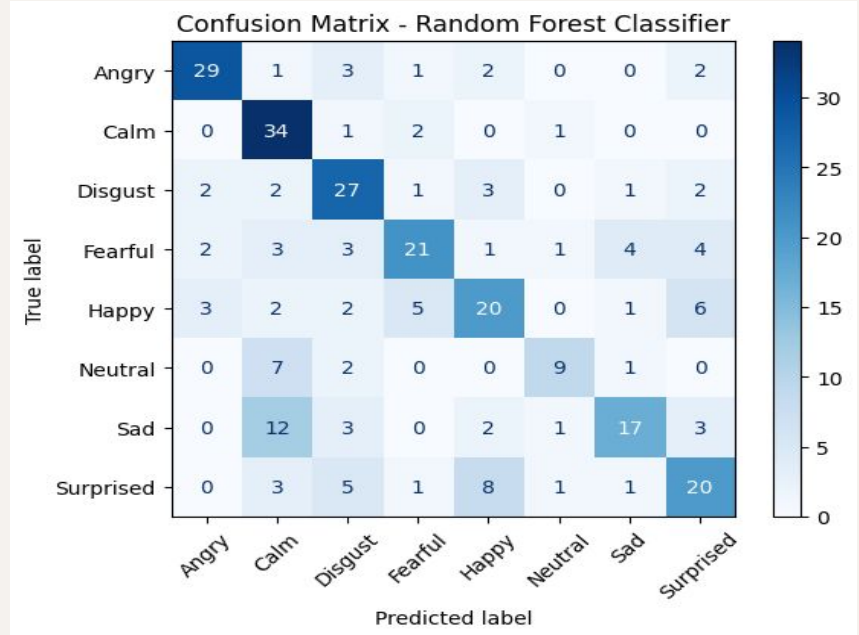
Results

Accuracy: 0.6145833333333334

Classification Report:

	precision	recall	f1-score	support
Angry	0.81	0.76	0.78	38
Calm	0.53	0.89	0.67	38
Disgust	0.59	0.71	0.64	38
Fearful	0.68	0.54	0.60	39
Happy	0.56	0.51	0.53	39
Neutral	0.69	0.47	0.56	19
Sad	0.68	0.45	0.54	38
Surprised	0.54	0.51	0.53	39
accuracy			0.61	288
macro avg	0.63	0.61	0.61	288
weighted avg	0.63	0.61	0.61	288

Accuracy Score



Confusion Matrix

XGB

```
xgb_model = xgb.XGBClassifier(  
    objective='multi:softmax',  
    num_class=len(emotion_encoder.classes_),  
    max_depth=None,  
    learning_rate=0.1,  
    n_estimators=700,  
    subsample=0.9,  
    colsample_bytree=0.9,  
    random_state=42,  
    use_label_encoder=False,  
    eval_metric='mlogloss'  
)  
  
xgb_model.fit(X_train, y_train)  
  
y_pred = xgb_model.predict(X_test)  
  
y_test_labels = emotion_encoder.inverse_transform(y_test)  
y_pred_labels = emotion_encoder.inverse_transform(y_pred)
```

- Step 1
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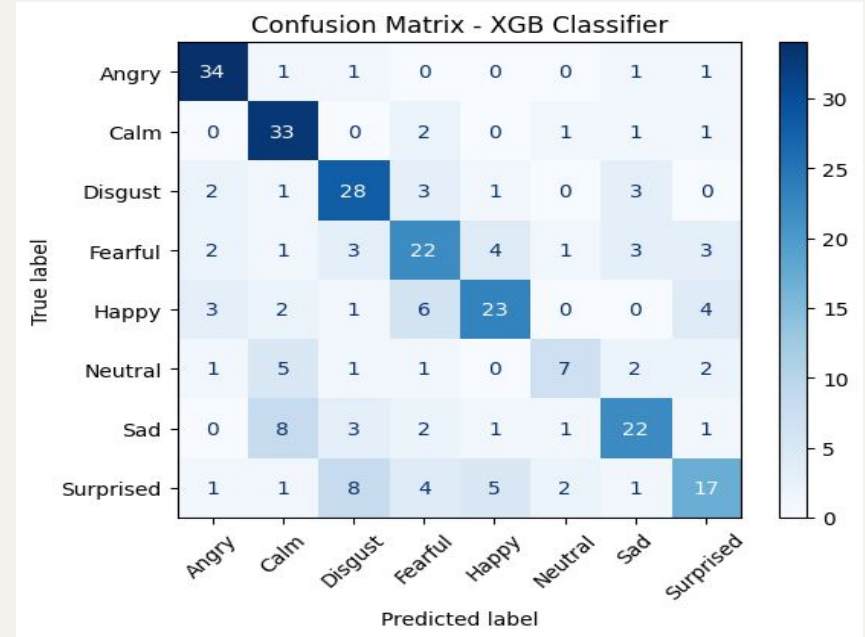
Results

Accuracy: 0.6458333333333334

Classification Report:

	precision	recall	f1-score	support
Angry	0.79	0.89	0.84	38
Calm	0.63	0.87	0.73	38
Disgust	0.62	0.74	0.67	38
Fearful	0.55	0.56	0.56	39
Happy	0.68	0.59	0.63	39
Neutral	0.58	0.37	0.45	19
Sad	0.67	0.58	0.62	38
Surprised	0.59	0.44	0.50	39
accuracy			0.65	288
macro avg	0.64	0.63	0.63	288
weighted avg	0.64	0.65	0.64	288

Accuracy Score



Confusion Matrix

Contributions

Samantha

- Data Processing
- Extract Labels from filename identifiers
- Dataframe creation
- Data Visualizations
- MCFF feature extraction function
- Random Forest Classifier Model
- Confusion matrices
- Feature Importance visualizations

Savannah

- Mel-Spectrogram feature extraction
- Spectral Centroid feature extraction
- Tempo feature extraction
- Pitch feature extraction
- Data Visualizations
- Random Forest Classifier Fine Tuning

Bhagavan

- Encode Categorical Labels into Numerical Labels
- Classification Report for Models
- XGB Classifier Model
- Attempted CNN Modeling for predicting labels

Aditya

- Presented own work

References

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- [2]M. M. Rezapour Mashhadi and K. Osei-Bonsu, "Speech emotion recognition using machine learning techniques: Feature extraction and comparison of convolutional neural network and random forest," PloS one, vol. 18, no. 11, p. e0291500, 2023, doi: <https://doi.org/10.1371/journal.pone.0291500>.
- [3]S. Hamsa, I. Shahin, Y. Iraqi and N. Werghi, "Emotion Recognition From Speech Using Wavelet Packet Transform Cochlear Filter Bank and Random Forest Classifier," in IEEE Access, vol. 8, pp. 96994-97006, 2020, doi: 10.1109/ACCESS.2020.2991811. keywords: {Emotion recognition;Speech recognition;Hidden Markov models;Feature extraction;Transforms;Modulation;Computational modeling;Emotion recognition;noise reduction;cochlear filterbank;feature extraction}
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- [5]"Mel-frequency Cepstral Coefficients (MFCC) for Speech Recognition." GeeksforGeeks. <https://www.geeksforgeeks.org/mel-frequency-cepstral-coefficients-mfcc-for-speech-recognition/> (accessed May 6, 2025)
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Thanks

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