

Machine Learning Project Presentation

Title: Gender and Emotion Classification by Voice



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Voice: used for human communication.

- Audio categorization
- Speech emotion recognition
- Mental Health and Well-being Monitoring

Emotion: used for expressing feelings.

- Enhancing Human-Computer Interaction
- Education and Training
- Automotive Industry



- Voice based Age, Accent and Gender Recognition [\[1\]](#)
 - Utilizing five hidden layers for gender (91% accuracy) and grid search pipeline for age (59% accuracy).
 - Mel-Frequency Cepstral Coefficients (MFCCs) and Spectrogram analysis are the major features extracted for model training.
 - Potential applications could be biometric identification and NLP.
 - Challenges in enhancing age detection accuracy and handling various accents.

Literature Review (2/2)



- Emotion Recognition using pitch , tone and rate.[\[2\]](#)
 - The paper discusses extracting features like pitch, tone, and rate for emotion detection.
 - It reviews several classification schemes, including hidden Markov models (HMMs) and neural networks.
 - Other classification methods mentioned include support vector machines (SVMs) and decision trees.
 - The study also discusses publicly available databases used for training and testing emotion recognition systems.

Dataset Description – Voice Gender Dataset

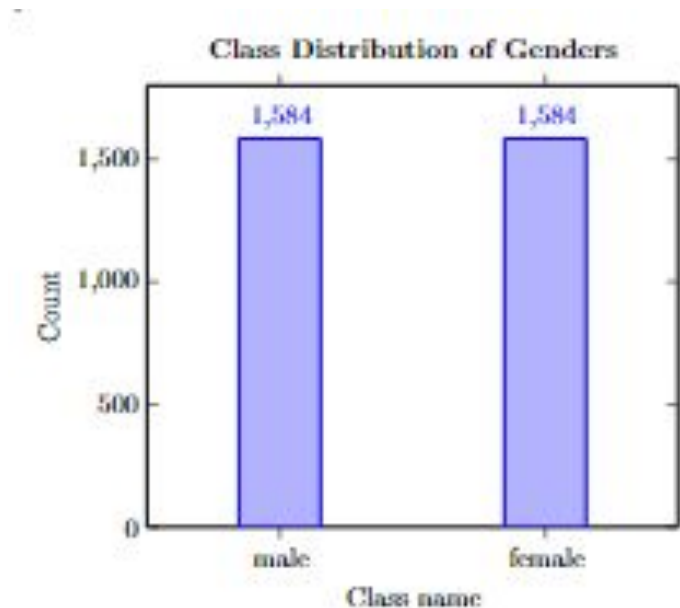


- Dataset consists of 3168 supervised learning samples where each voice recording is labeled with either male or female gender.
- Acoustic properties that help in distinguishing male and female voice are:
 - Mean frequency (Hz)
 - Standard Deviation of frequency
 - Median Frequency
 - Skewness of Frequency Distribution etc.
- [\[Kaggle Link\]](#)

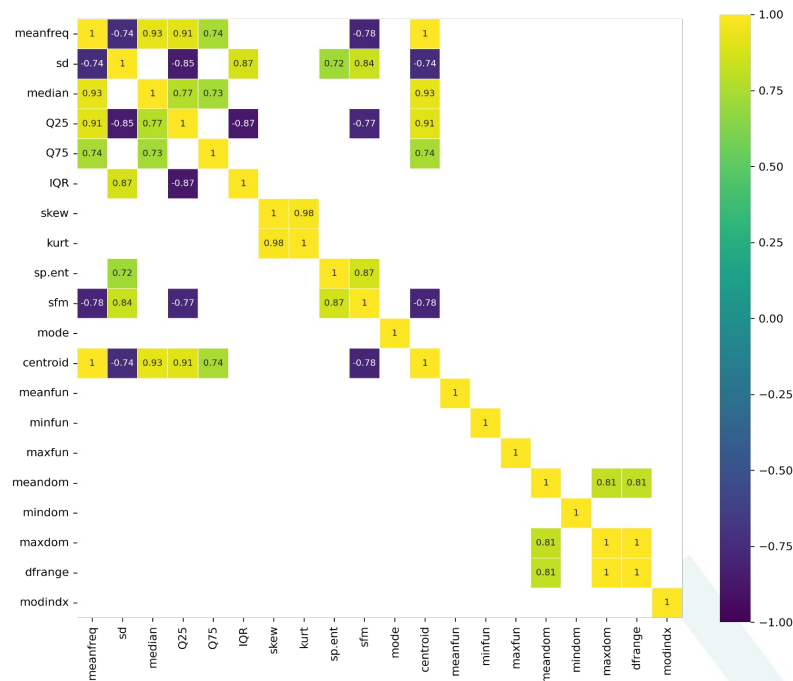
Dataset Description – Emotion Gender Dataset

- Consists of 1440 supervised audio samples from 24 actors (12 male + 12 female), capturing a diverse range of emotions.
- Encompasses 8 distinct emotions:
neutral, calm, happy, sad, angry, fearful, disgust, and surprised.
- [\[Kaggle Link\]](#)

Visualizations

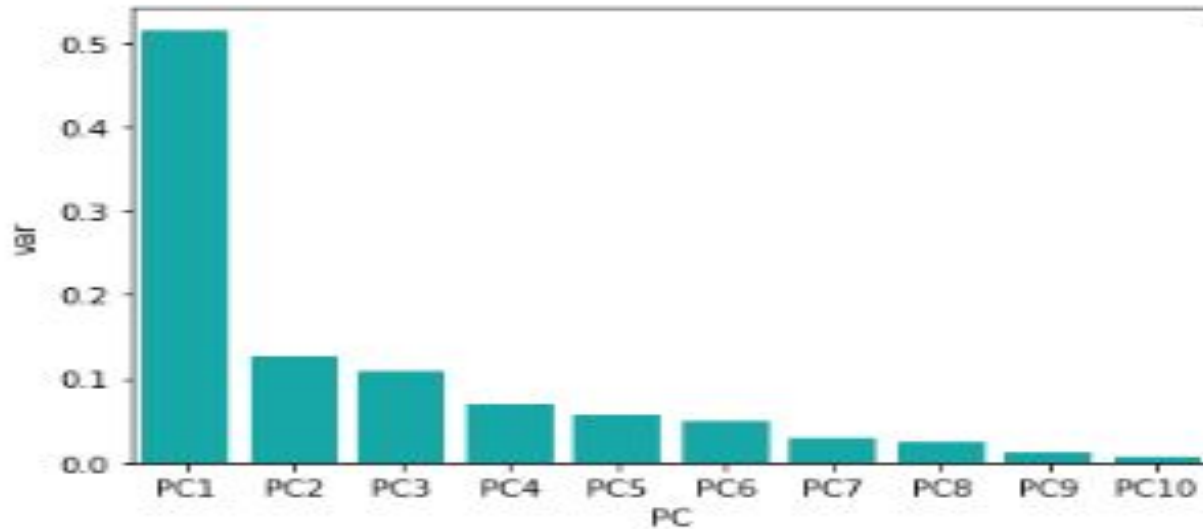


Class Distribution for output label



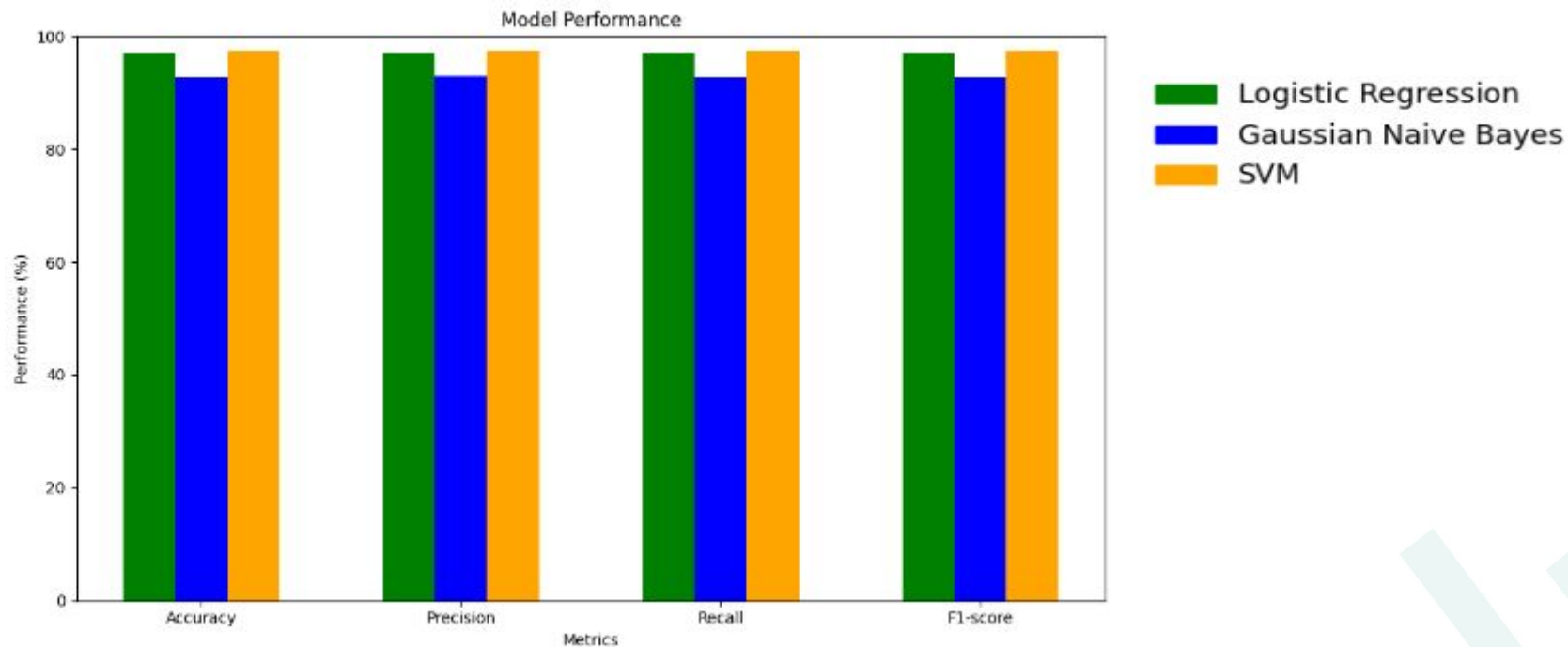
Heatmap for highly correlated columns

Feature Importance Graph using PCA



- Preprocessed and cleaned data by normalizing, null checking, duplicate handling etc.
- Classified the speaker's gender using selected voice attributes.
- The speaker's voice was categorized as either 'male' or 'female.'
- Three different models/methodologies were used for classification:
 - Logistic Regression
 - Naive Bayes:
 - Gaussian
 - Bernoulli
 - Support Vector Machine

Results



Conclusion



- Logistic Regression provided good results for each of the metrics chosen to judge the model.
- Gaussian naive bayes also reported good accuracy and other performance metrics, therefore, it implies that:
 - The Gaussian Naive Bayes model fits perfectly to the data.
 - The voice dataset is produced according to the Gaussian distribution for male and female categories.
- In SVM model, linear kernel was observed to be the best one.

Timeline



Timeline followed:

- Week 1-2: Data Collection , Pre-processing and Data Visualization for Gender Recognition.
- Week 3-4: Logistic Regression, Naïve Bayes, Support Vector Machine.
- Week 5: EDA, Feature Analysis, Heatmaps, PCA, TSNE (Analysis of Models).
- Week 6: Feature Selection and Observing the performance of each model .
- Week 7(after midsem):Data Collection for Emotion Recognition.
- Week 8:Report and PPT.

Contributions



Abdullah Shujat: Pre-processing and Data Visualization, Logistic Regression, Naive Bayes, Report

Anant Kaushal: Data Collection and Visualization, SVM , Feature Extraction , PPT

Anikait Agrawal: Data Collection, Logistic Regression, Naive Bayes , Report

Ansh Varshney: Pre-processing and Data Visualization , Feature Extraction , SVM , PPT



THANK YOU! 🧐💪

