



DEEP AUDIO CLASSIFIER

MAJOR PROJECT — I (AD-708)



Team Members

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Guide Details

- Prof. Dheeraj Namdev
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Idea Solution/Prototype

The primary goal of this project is to build a deep learning model that can classify different urban sound events effectively. By leveraging the UrbanSound8K dataset, which provides a rich collection of labeled audio clips, the project integrates Natural Language Processing (NLP) for enhanced data processing and interpretation. The model architecture is designed to preprocess audio features, extract relevant patterns, and apply NLP techniques to improve the understanding and categorization of audio signals, thereby increasing classification accuracy and model robustness.



Abstract

This project develops a deep audio classification system to identify and categorize urban sounds using the UrbanSound8K dataset. It focuses on processing and analyzing sound clips such as sirens, street music, and engine noises. The project incorporates detailed audio signal analysis and data exploration to ensure balanced and comprehensive input for model training. The ultimate goal is to enhance the model's ability to accurately predict and classify different audio events, supporting applications in noise monitoring, smart city management, and accessibility tools.

Technology Stack







Requirements

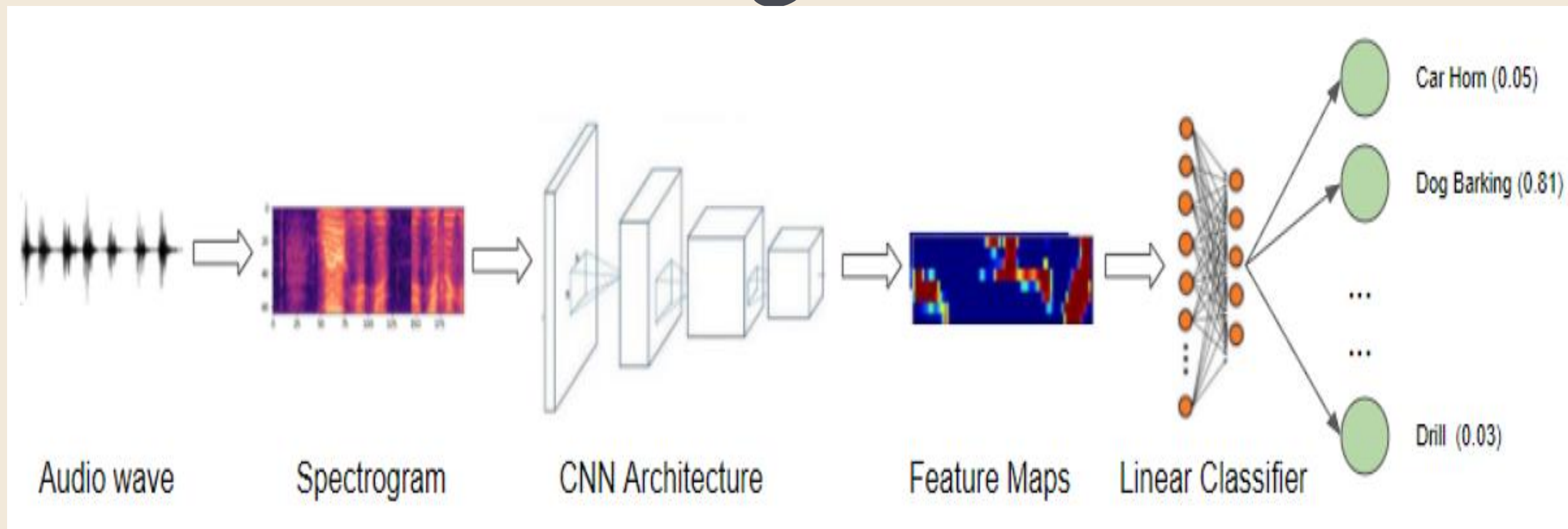
Functional Requirements

- Audio file upload.
- Classification of audio into predefined categories.
- Integration with NLP modules for preprocessing and feature extraction.
- User interface for prediction.

Non-Functional requirements

- High model accuracy and low latency.
 - Robust error handling for various input formats.
 - Scalability to adapt to larger datasets or real-world data streams.
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Flowchart Diagram

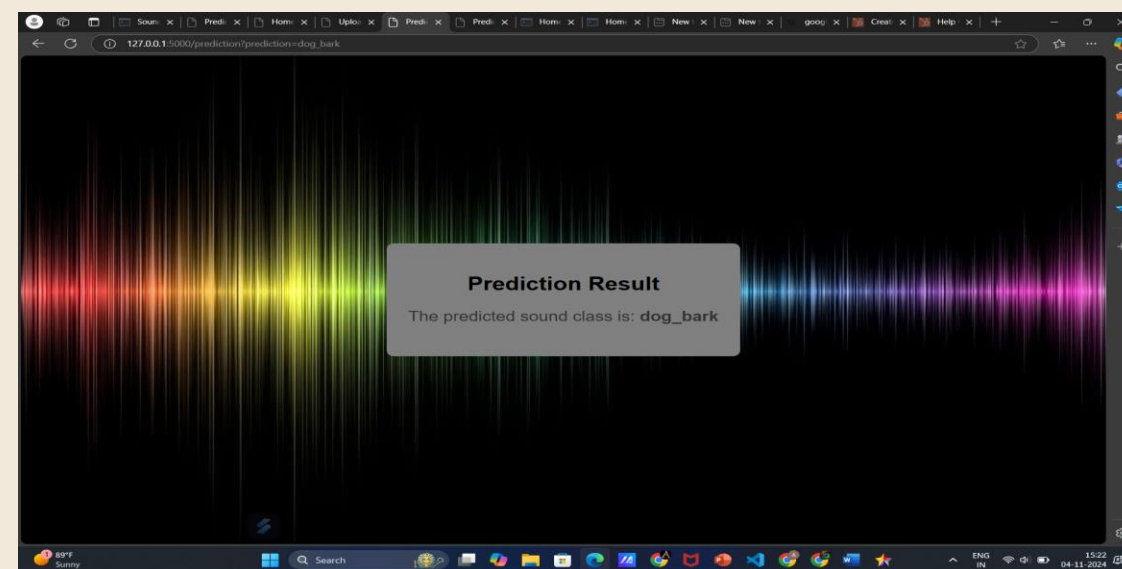
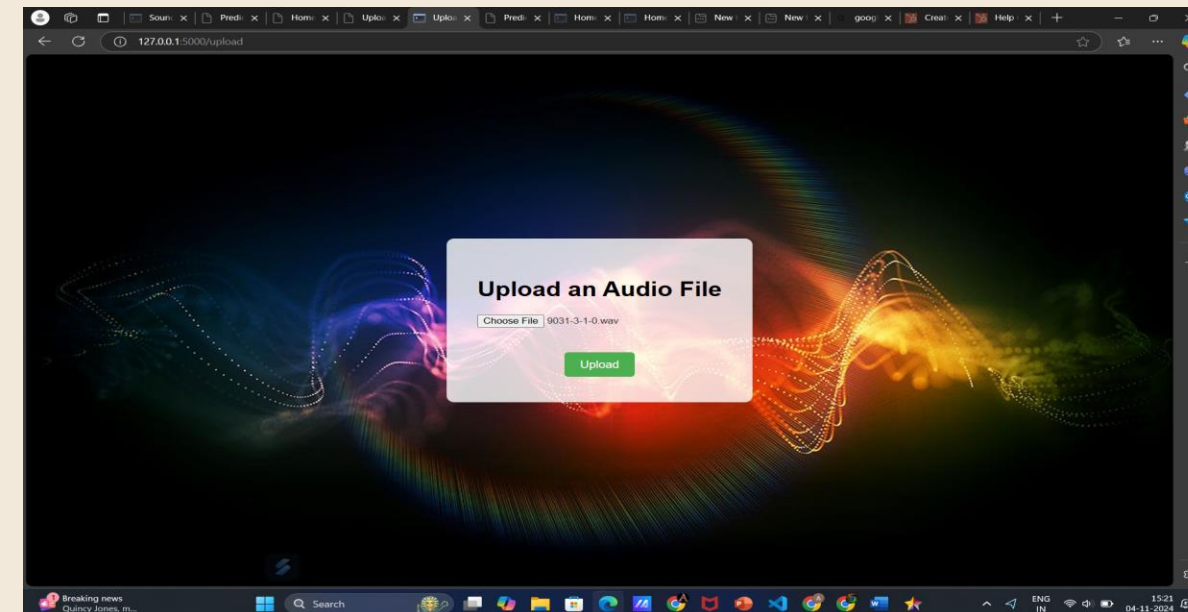
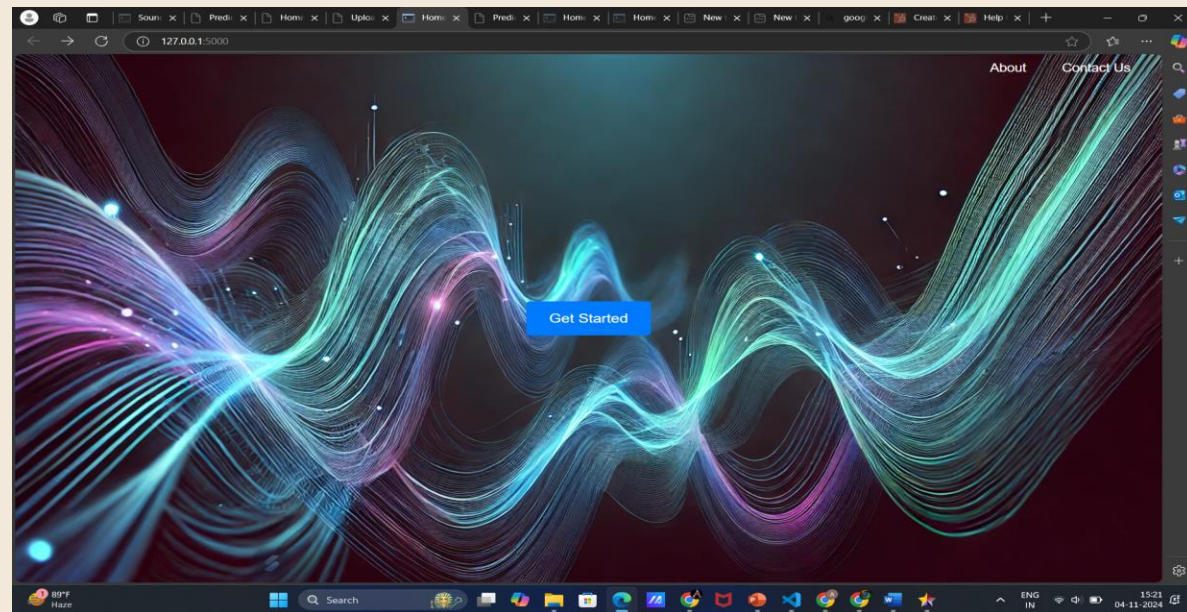




Deployment

- Hosting cost according to the servers selected.
- Domain Purchasing Cost.

Frontend Screenshots



Backend Screenshots

```
app.py 4 UrbanSound.py 9+ x index.html upload.html UrbanSound.ipynb prediction.html
UrbanSound.py > ...
16 import numpy as np
17 import pandas as pd
18 import matplotlib.pyplot as plt
19 import seaborn as sns
20 # %matplotlib inline
21
22 import warnings
23 warnings.filterwarnings("ignore")
24 # -
25
26 # !pip install librosa
27
28 import sys
29 print(sys.executable)
30
31
32 # !python -c "import librosa; print(librosa.__version__)"
33
34
35 import IPython.display as ipd
36 import librosa
37 import librosa.display
38
39 filename = 'UrbanSound8K/fold5/190893-2-0-11.wav'
40
41
42 # Librosa
43 #
44
45
46 librosa_audio_data,librosa_sample_rate = librosa.load(filename)
47
48
49 librosa_audio_data
50
51
52 librosa_sample_rate
53
54
55 librosa.display.waveshow(librosa_audio_data,sr=librosa_sample_rate)
56 ipd.Audio(filename)
```

```
app.py 4 UrbanSound.py 9+ x index.html upload.html UrbanSound.ipynb prediction.html
UrbanSound.py > ...
57
58 # Here Librosa converts the signal to mono, meaning the channel will always be 1
59
60
61
62 # scipy
63 #
64
65
66 from scipy.io import wavfile as wav
67 wave_sample_rate,wave_audio = wav.read(filename)
68
69 wave_audio
70
71 wave_sample_rate
72
73
74 plt.figure(figsize=(12,4))
75 plt.plot(wave_audio)
76 plt.show()
77
78 # Here Scipy converts the signal to stereo, meaning the channels will be 2
79 #
80
81
82
83 # Metadata
84
85
86
87 metadata = pd.read_csv('UrbanSound8K.csv')
88 metadata.head()
89
90 metadata.shape
91
92
93 metadata.isnull().sum()
94
95
96 metadata.duplicated().sum()
97
```



THANK YOU