

Appliance Detection

Smart Home Automation Project
Embedded Systems

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Motivation

Background:

Smart homes are essential conveniences for consumers who want to make their lives just a little bit easier!

At the forefront of these advancing technologies are smart assistants:



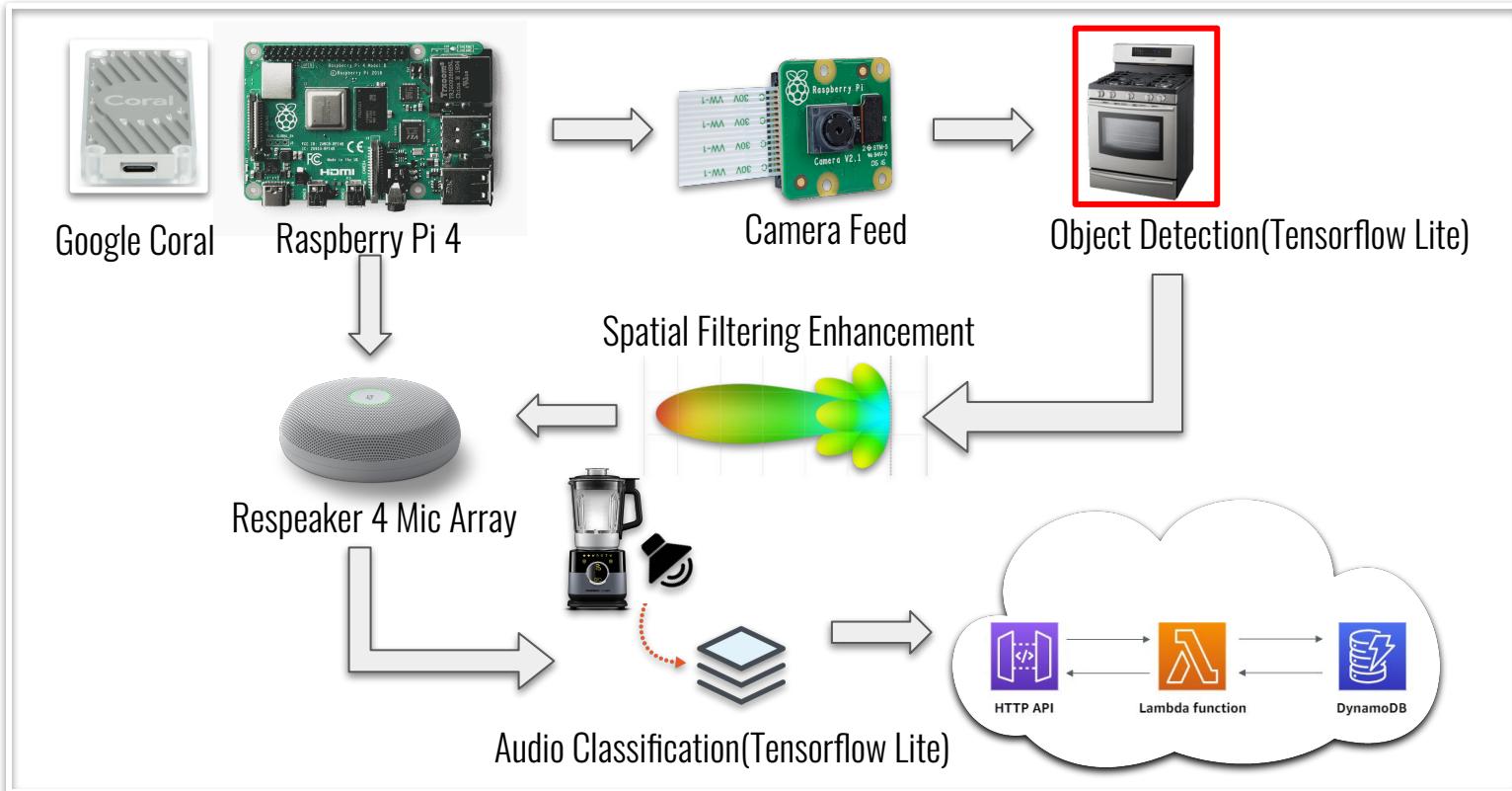
amazon alexa



Goals:

- Advance smart assistant functionality in noisy environments.
- Ease a consumer's life by identifying/classifying differing sounds and appliances.
- Manage useful events for each noise identified.

High-Level Approach



ODAS/ODAS_WEB



Features:

- Sound Source Separation
- Sound Source Localization
- Sound Source Tracking
- Post Filtering

Features of Interest:

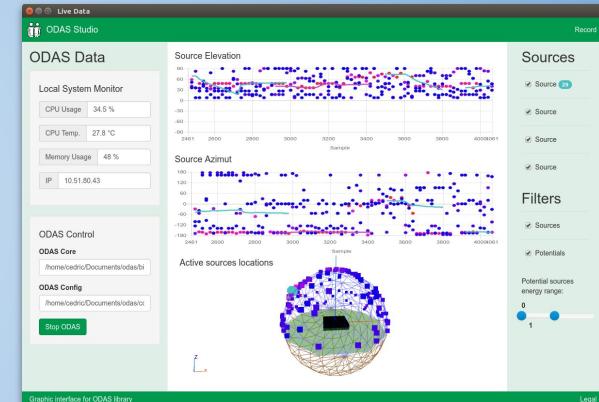
- Static Sound Source Tracking
- Spatial Filters

Initial Approach:

- Use the Static Sound Source tracking to beamform to locations of interest dynamically.

Secondary Approach:

- Use Spatial Filters



Certain features broken(Static Sound Source tracking does not always work)



Difficult to output/use data without ODAS_WEB; Faced many difficulties running ODAS_WEB



Dynamically configuring file at runtime required restructuring of project or scripting

Object Detection

MODEL

- Tensorflow Lite Pre-Trained Model
 - Based on COCO Common Objects Dataset
 - Streamlined:



ENHANCEMENT FEATURES

1) Centroid Extraction:

```
{'oven': (444.0, 400.0), 'sink': (137.5, 429.5),
{'oven': (444.0, 402.0), 'sink': (137.5, 429.5),
{'oven': (444.0, 402.0), 'sink': (137.5, 432.5),
```

2) Closeness Detection:

```
if "person" in centroid_dict.keys():
    for key in centroid_dict.keys():
        centroid_diff_dict[key] = (abs(centroid_dict[key][0] - centroid_dict["person"])[0],
        abs(centroid_dict[key][1] - centroid_dict["person"])[1])
        #print(centroid_diff_dict)
```

3) Angle Detection:

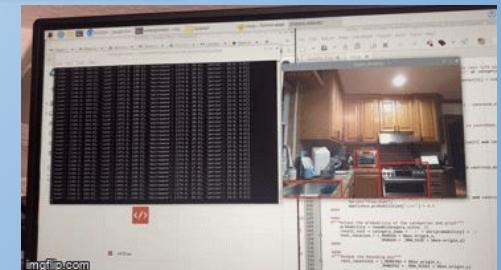
```
if category_name == "microwave" or category_name == "oven" or category_name == "sink":
    for x in centroid_dict.keys():
        angle_from_center[x] = (((((centroid_dict[x][0] - ccenter[0]) / ccenter[0])*(62.2/2)) ,
        ((centroid_dict[x][1] - ccenter[1]) / ccenter[1]) * (48.8/2))
        print(angle_from_center)
```

OUTPUTS

Object Detection/Centroids:

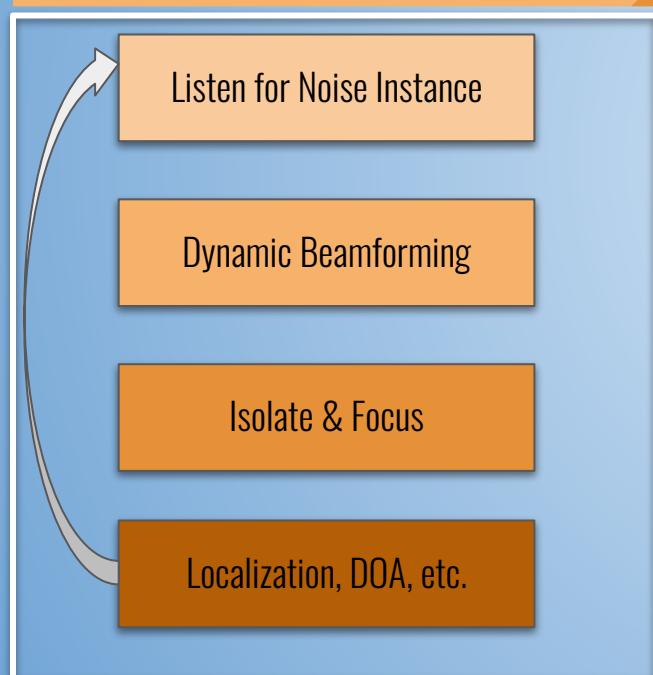


Closeness Detection:

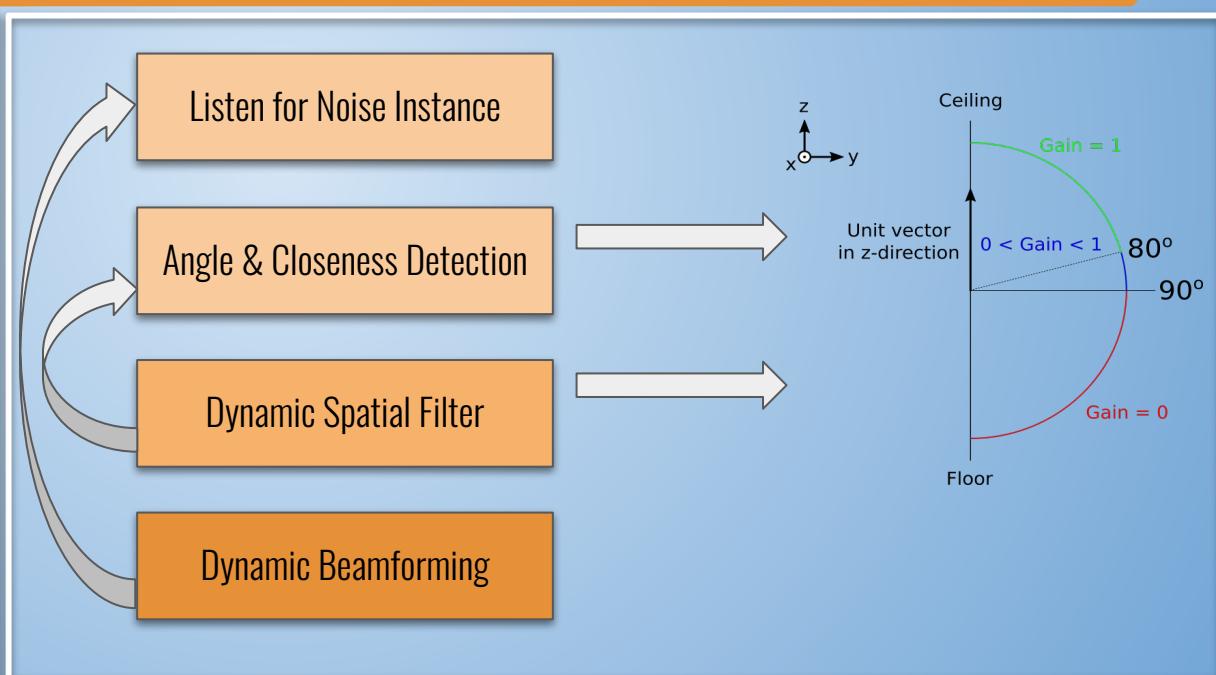


Microphone Array

STANDARD IMPLEMENTATION



PROPOSED IMPLEMENTATION



Microphone Array

STANDARD IMPLEMENTATION



PROPOSED IMPLEMENTATION

```
spatialfilters = (  
    {  
        direction = ( +0.000, +0.000, +1.000 );  
        angle = (80.0, 100.0);  
    }  
    {  
        direction = ( +1.000, +0.000, +0.000 );  
        angle = (angle_detection_x + 10.0 , angle_detection_x +20.0);  
    }  
    {  
        direction = ( +0.000, +1.000, +0.000 );  
        angle = (angle_detection_y + 10.0 , angle_detection_y +20.0);  
    }  
}
```

Audio Classification

MODEL

ENHANCEMENT FEATURE

OUTPUT

Tensorflow Lite Pre-Trained Model

- Based on Google AudioSet Data
- Extracting Required Objects

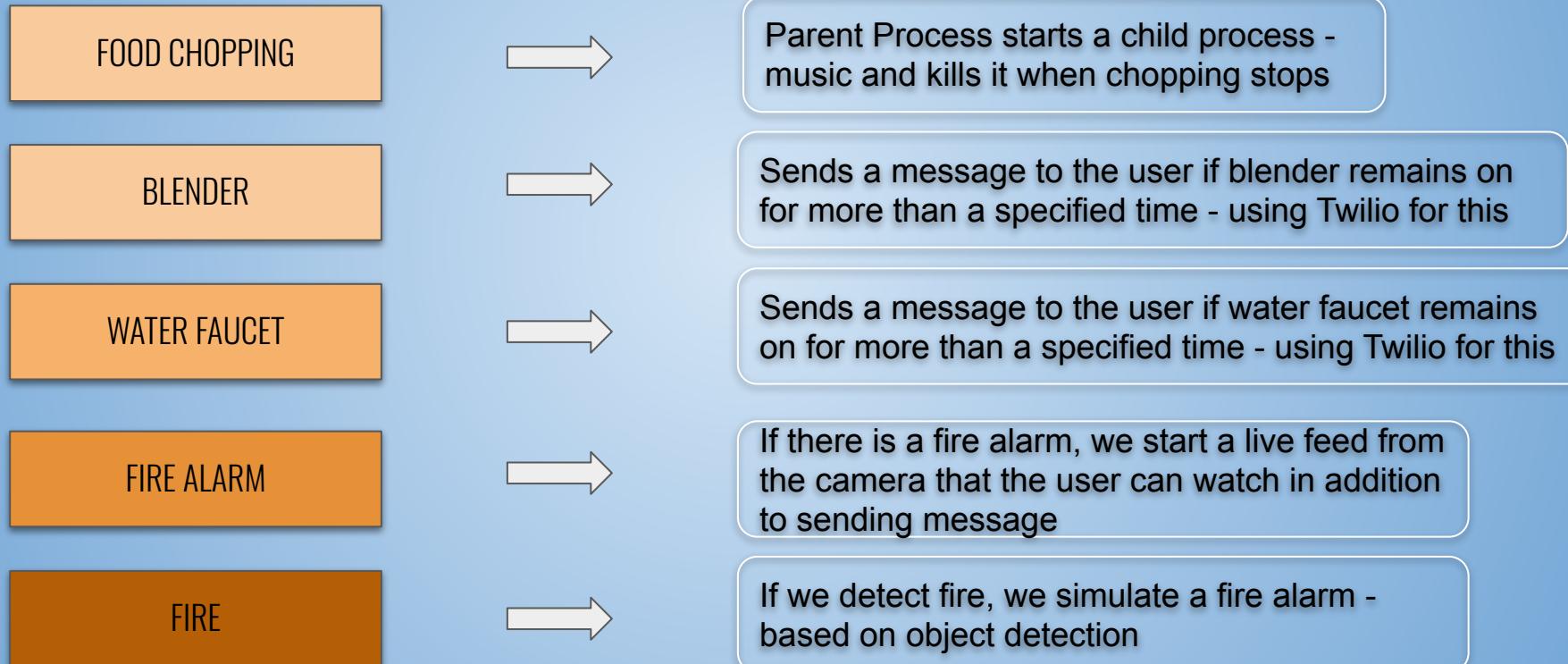
Confidence Level Normalization

Hysteresis

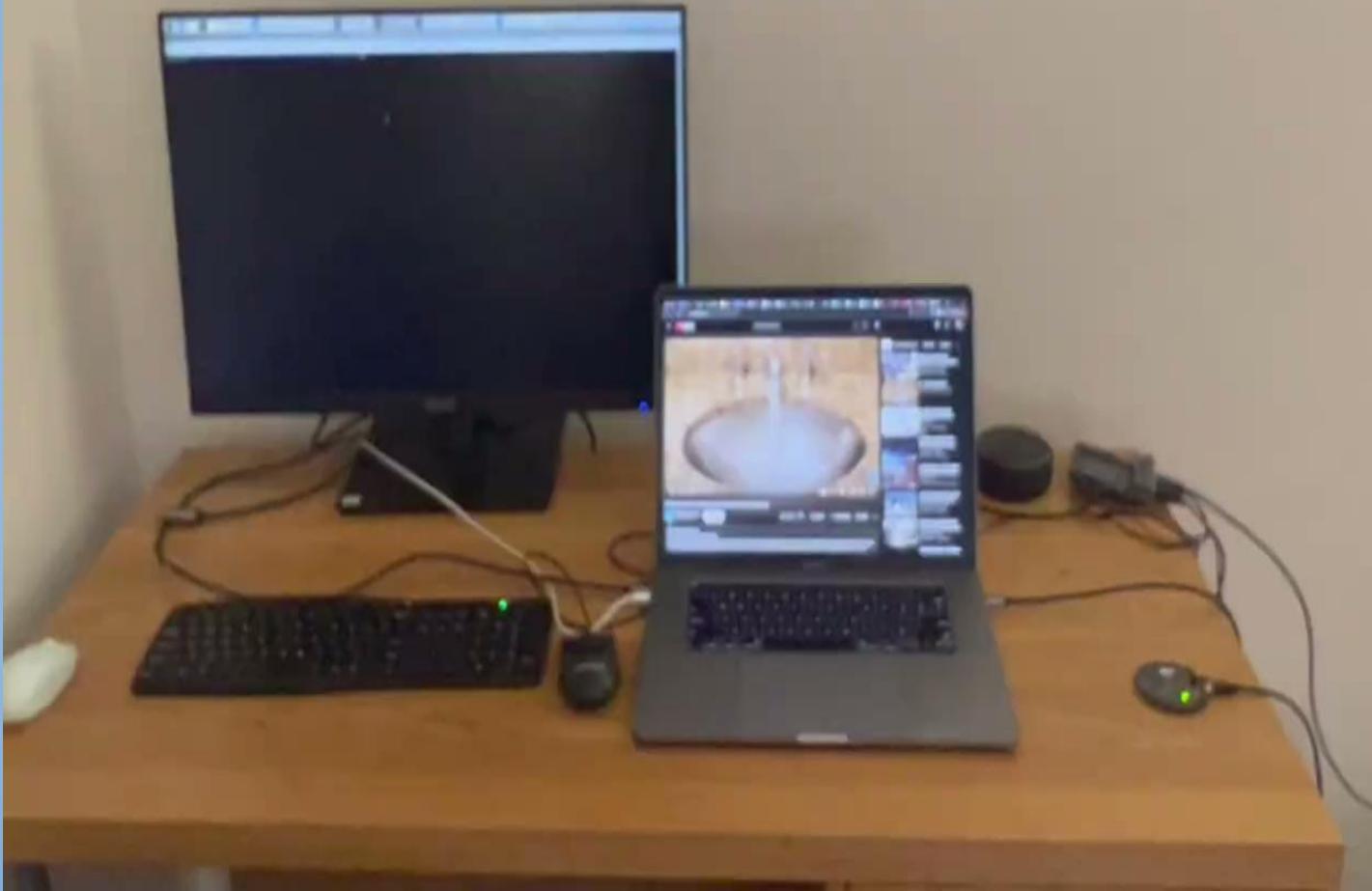
Identification of the Devices

Event Based Decisions

Event Based Actions



Music in Action



Message Notification in Action



User Interface

EVENT

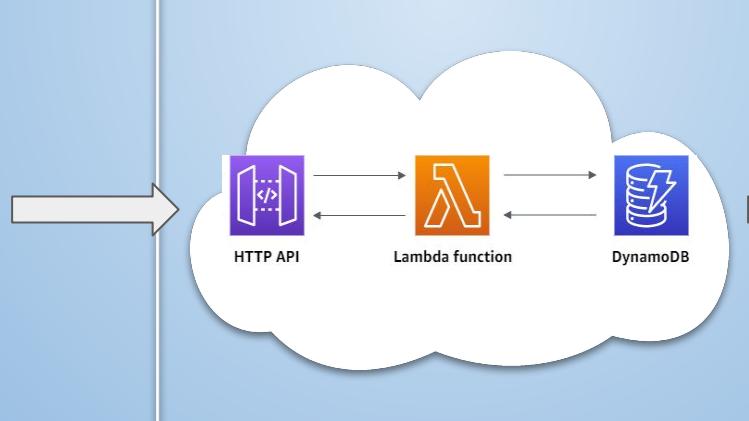
AWS

UI

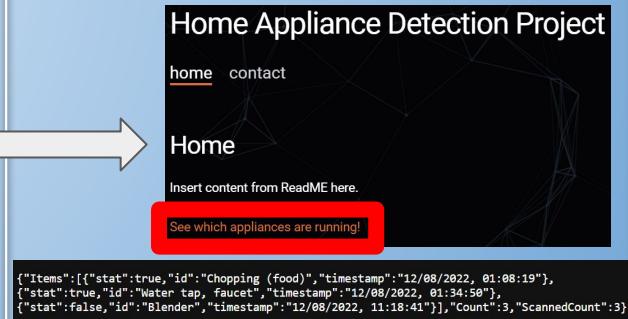
When an appliance is detected:
{
appliance: "sink",
status: True,
timestamp: datetime.datetime.now()
}

When an appliance stops running:
{
appliance: "sink",
status: False,
timestamp: datetime.datetime.now()
}

HTTP PUT



HTTP GET



Challenges

- Lack of flexible pre-existing localization techniques
 - Integration of ODAS with Python/Re-Speaker Configuration File w/ Dynamic Variables.
- Integration of Individual Components
 - Finding features to take advantage of and implement.
 - Sequential vs Parallel Processing
- Chip Shortage - Lack of PI available

Future Steps

- Implement parallel execution of models.
- Implement Stereo Camera! Can easily take into account depth/XYZ real-world coordinates.
- Create dynamically configurable localization:
 - Write a script that accesses configuration file and modifies as variables change.
 - Create spatial filtering from scratch.
- Add multiple users with credentials to view personal appliances
- Create specialized Audio and Computer Visions models from scratch
- Implement Live Feed Feature
- Develop a more intuitive and attractive UI

Contributions

- **Bhanu**
 - Audio Classification Model research
 - Model selection, tuning and normalization
 - Creating a separate process for parallel task
 - Twilio text updates
- **Nathan**
 - Model research
 - Localization and beamforming
 - Computer vision detection
- **Disha**
 - Hardware research
 - Localization and beamforming
 - Computer vision detection
 - AWS integration

References

- <https://github.com/yinkalario/General-Purpose-Sound-Recognition-Demo>
- [https://github.com/tensorflow/examples/tree/master/lite/examples/object detection/raspberry_pi](https://github.com/tensorflow/examples/tree/master/lite/examples/object_detection/raspberry_pi)
- https://github.com/tensorflow/examples/tree/master/lite/examples/audio_classification/raspberry_pi
- [introlab/odas: ODAS: Open embeddeD Audition System \(github.com\)](#)
- [Spatial Filter configuration · Issue #19 · introlab/odas \(github.com\)](#)
- <https://docs.aws.amazon.com/apigateway/latest/developerguide/http-api-dynamo-db.html>
- <https://wiki.seeedstudio.com/ReSpeaker-USB-Mic-Array/>
- https://www.w3schools.com/python/module_requests.asp

Audio Classification Research

- ❖ Different Approaches tried
 - Training our own models - Edge Impulse
 - Fine Tuning Pre-trained models - Using AWS SaaS
 - Fully Trained Models
 - [General-Purpose-Sound-Recognition-Demo](#)
 - [Tensor_flow_lite_audio_classification](#)