# IRIS

early Identification of Respiratory Illness System via Al and edge computing

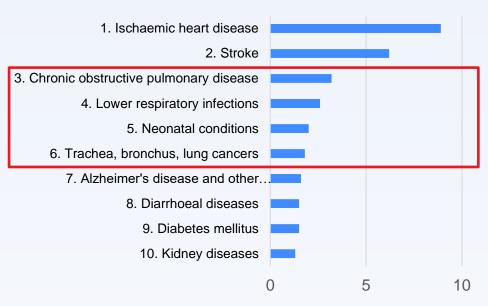
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# Research Background & Problem Statement

#### Disease Burdon

### 2019 world's leading causes of death



number of deaths (in millions)
Source: WHO



- 45.7 million adult asthma patients
- >100 million COPD patients

### Challenges

- Missed or delayed diagnosis results in the delayed start of effective treatment
- Early detection of respiratory illnesses is vital, but diagnostic tests are difficult to perform or difficult to obtain.
- Respiratory disease cannot be managed effectively without continuous monitoring data.







# **Proposed Solution**

A system for early identification of respiratory diseases based on edge computing and fast integration of convolutional analysis models

Blood oxygen sensor

Cough sound and blood oxygen data analyzer with algorithmic model for cough sound detection and classification

web-based software to collect baseline data and analysis result on early identification of respiratory disease



# **Selected Parameters**



### **COPD-SQ**

Mostly used In China for screening. It comprises 7 questions and the highest total score is 38 points.

At risk: score > 16



# blood oxygen

Blood oxygen saturation is the amount of oxygen circulating in the blood. It is a crucial measure of lung function.

At risk: < 95%



# **Cough Sound**

Cough is a common symptom.10-38% of respiratory diseases are accompanied by chronic cough.

At risk: Type of cough sound

# **Innovation**

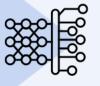
### Multidimensional data to determine risk

Integrative analysis of blood oxygen levels, cough sounds and questionnaire values



### **Cough sound model based on MobileNet**

Lightweight convolutional network backbone deployed efficiently on a single CPU while obtaining high accuracy



### **Access and availability**

### Fast ensemble approach

Improves convolutional network accuracy without adding significant computational complexity



### blood oxygen



Questionnaire

Stethoscope



X-Ray Imagine

CT Scan

### **User-friendliness and usability**

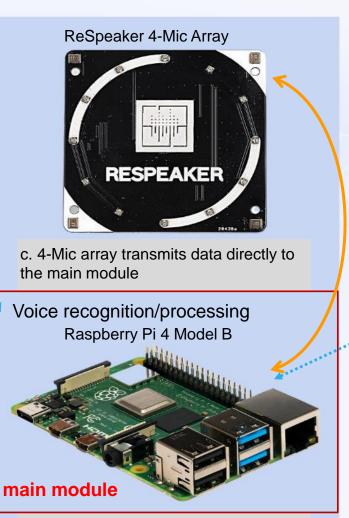
Portable integrated devices allow continuously ( monitoring physiological data

# **Hardware Prototyping**

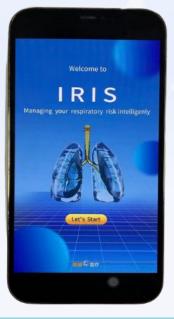
b. M5Stack C+ sends data via WLAN to the main module



a. Upload the pulse oximeter data from MAX30100 to M5Stack C+



d. Main module processes data, analyzes data model, and provides web backend services



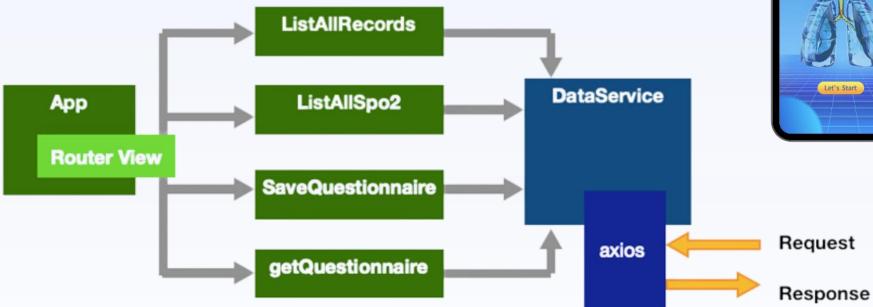
b. Connect the web client to Wi-Fi for data checkup

Job Position





The django-Vue web application framework







# **Outlook Design**









- Portable
- Effective heat dissipation
- ABS material



# **Open Dataset for Model Training**

Cough detection model (A + B dataset)

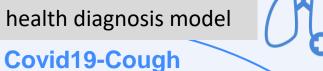
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### A URBANSOUND8K

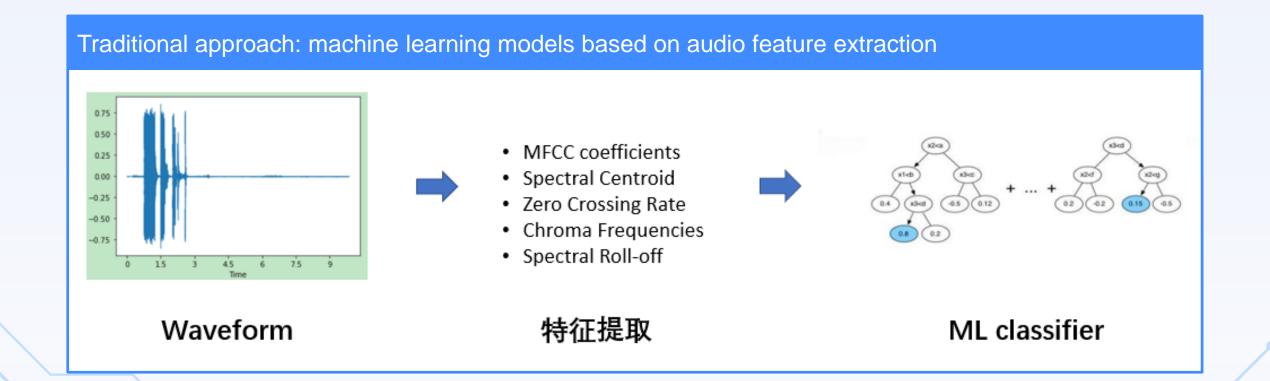
- Sample without cough sounds
- 8,732 annotated audio clips
- Sounds covering air conditioning, car horns, music, construction sounds, and children playing etc.

### Lung health diagnosis model



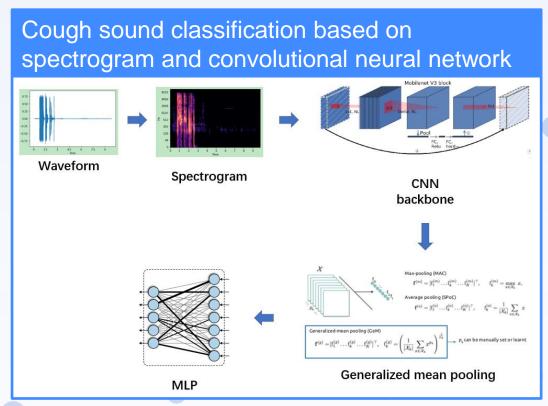
- Anonymize&
- 1,324 subjects
- Label: COVID19-positive/respiratory symptoms
- 682 positive(382 PCR results, 295 symptomatic)
- Recordings: total 58 minutes of cough sound, average 2.6s each
- Symptomatic samples as positive and other samples as negative

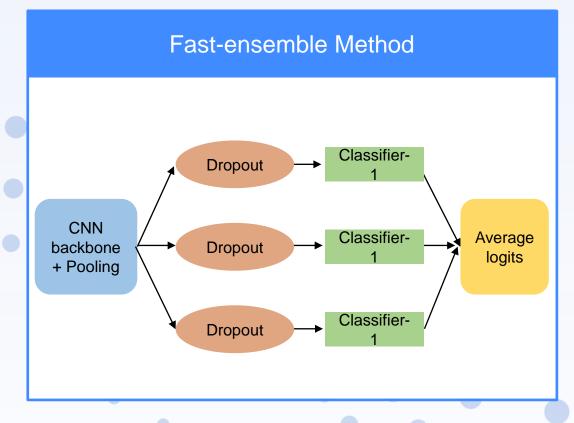
# **Model Framework - 1**



# **Model Framework - 2**

- Deep convolutional networks offer better performance on a variety of audio tasks due to:
  - Powerful automatic feature extraction capability;
  - Pre-training on ImageNet: better parameter initialization;
- Fast Ensemble method: provide model ensemble during inference without much loss of efficiency





# **Model Performances**

Cough detection model						
	Model	AUC(%)	TPR(%)	TNR(%)		
Machine learning	LightGBM	78.5	73.2	80.1		
Deep learning	MobileNet-v3 (not pretrained)	86.8	80.6	85.7		
	MobileNet-v3	97.1	91.5	94.2		
	MobileNet-v3 + Fast Ensemble	97.9	92.7	94.5		

Lung health diagnosis model						
	Model	AUC(%)	TPR(%)	TNR(%)		
Machine learning	LightGBM	65.3	56.7	61.1		
Deep learning	MobileNet-v3 (not pretrained)	73.1	58.7	72.7		
	MobileNet-v3	81.1	70.6	80.3		
	MobileNet-v3 + Fast Ensemble	82.5	71.3	81.6		

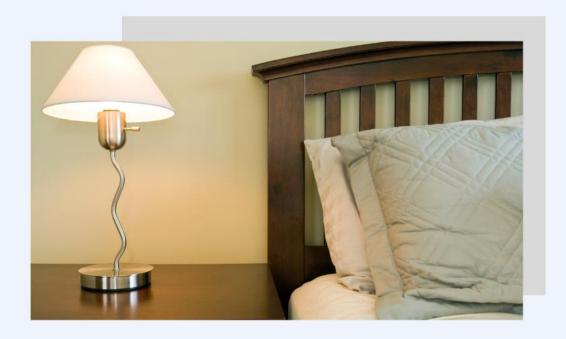
Adopted the pre-trained lightweight network *MobileNet-v3* provided by the Pytorch-Image-Models framework as our convolution backbone.

#### Performances:

- The AUC of the cough detection model reached 97.9%, and the AUC of the lung health diagnosis model reached 82.5%
- Fine-tuning pre-trained backbone models significantly improves the results;
- The fast ensemble method further improves fine-tuned models effectively

### **Product Scenario**

- Personalized continuous respiratory health monitoring and decision making.
- Use algorithmic models to detect health risks and ensure early intervention
- Ease the burden on healthcare systems and ensure that medical resources are allocated efficiently, especially during epidemics.





# **Future Works**

Now Future

### Calculate separately

Analyze the risk of COPD-SQ, blood oxygen saturation, and cough sounds separately



### Binary

Results are classified as healthy and under risk



### Individual

User's own personal data







#### Cross-calculate

Combining multiple indicators to determine respiratory disease risk

### **CDSS**

Support the diagnosis of different respiratory diseases



#### China Dataset

An open database for cough sounds and physiological indicators

