

# NETAJI SUBHAS UNIVERSITY OF TECHNOLOGY



## COMPUTER HARDWARE SOFTWARE WORKSHOP COCSC19

SUBMITTED BY:

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# TINY ML

## PROJECT TITLE: AQI DETECTION

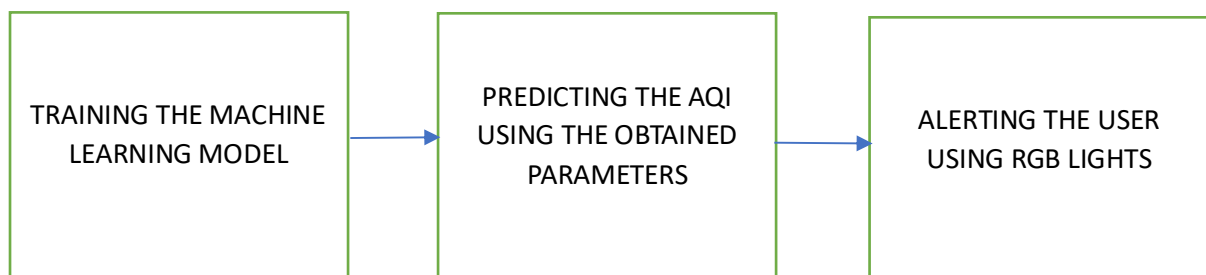
### IDEA:

Develop a machine learning model capable of forecasting the Air Quality Index (AQI) for a specific location by leveraging key atmospheric parameters such as PM2.5, PM10, NO2, and SO2 concentrations.

### TOOLS USED:

- Arduino uno simulator
- Jupyter notebook

### FLOW DIAGRAM:



## DATA:

To train the machine learning model we used a dataset consisting amounts of key atmospheric parameters like PM2.5, PM10, NO2, and SO2 concentrations from various cities across India since 2015 to 2020

	PM2.5	PM10	NO2	SO2	AQI
2123	81.40	124.50	20.50	15.24	184.0
2124	78.32	129.06	26.00	26.96	197.0
2125	88.76	135.32	30.85	33.59	198.0
2126	64.18	104.09	28.07	19.00	188.0
2127	72.47	114.84	23.20	10.55	173.0
...	...	...	...	...	...
29525	7.63	32.27	23.27	6.87	47.0
29526	15.02	50.94	25.06	8.55	41.0
29527	24.38	74.09	26.06	12.72	70.0
29528	22.91	65.73	29.53	8.42	68.0
29529	16.64	49.97	29.26	9.84	54.0

# PYTHON CODE:

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn import linear_model
```

```
combined_data = pd.read_csv('city_day.csv')
```

```
combined_data.head()
```

	City	Date	PM2.5	PM10	NO	NO2	NOx	NH3	CO	SO2	\
0	Ahmedabad	2015-01-01	NaN	NaN	0.92	18.22	17.15	NaN	0.92	27.64	
1	Ahmedabad	2015-01-02	NaN	NaN	0.97	15.69	16.46	NaN	0.97	24.55	
2	Ahmedabad	2015-01-03	NaN	NaN	17.40	19.30	29.70	NaN	17.40	29.07	
3	Ahmedabad	2015-01-04	NaN	NaN	1.70	18.48	17.97	NaN	1.70	18.59	
4	Ahmedabad	2015-01-05	NaN	NaN	22.10	21.42	37.76	NaN	22.10	39.33	

	O3	Benzene	Toluene	Xylene	AQI	AQI_Bucket
0	133.36	0.00	0.02	0.00	NaN	NaN
1	34.06	3.68	5.50	3.77	NaN	NaN
2	30.70	6.80	16.40	2.25	NaN	NaN
3	36.08	4.43	10.14	1.00	NaN	NaN
4	39.31	7.01	18.89	2.78	NaN	NaN

```
combined_data=combined_data.dropna()
```

```
X= combined_data[['PM2.5','PM10','NO2','SO2']]
Y= combined_data['AQI']
```

```
regr=linear_model.LinearRegression()
regr.fit(X,Y)
```

```
LinearRegression()
```

```
print(regr.coef_)
```

```
[ 0.89991892  0.48236037  0.0356627 -0.04208933]
```

```
print(regr.intercept_)
```

```
25.142174108944573
```

# ARDUINO CODE:

```
int redPin = A3;
int greenPin = A4;
int bluePin = A5;
float output=0;
float lrCoef[5] =
{25.142174108944573, 0.89991892, 0.48236037, 0.0356627, -0.04208933};
void setup() {

  pinMode(A0, INPUT);
  pinMode(A1, INPUT);
  pinMode(A2, INPUT);
  pinMode(A3, INPUT);

  pinMode(redPin, OUTPUT);
  pinMode(greenPin, OUTPUT);
  pinMode(bluePin, OUTPUT);

  Serial.begin(9600);
  // delay(1000);

  float val1 = 81.40;
  float val2 = 124.50;
  float
val3 = 20.50;
  float val4 = 15.24;

  output = multiLinReg(val1, val2, val3, val4);

  Serial.print(output, 10);

}

void loop() {
  if(output<=100){
    analogWrite(redPin,
255);
    analogWrite(bluePin, 255);
  }
  else if(output<=200){
    analogWrite(bluePin,
255);
  }
  else{
    analogWrite(greenPin, 255);
    analogWrite(bluePin, 255);
  }

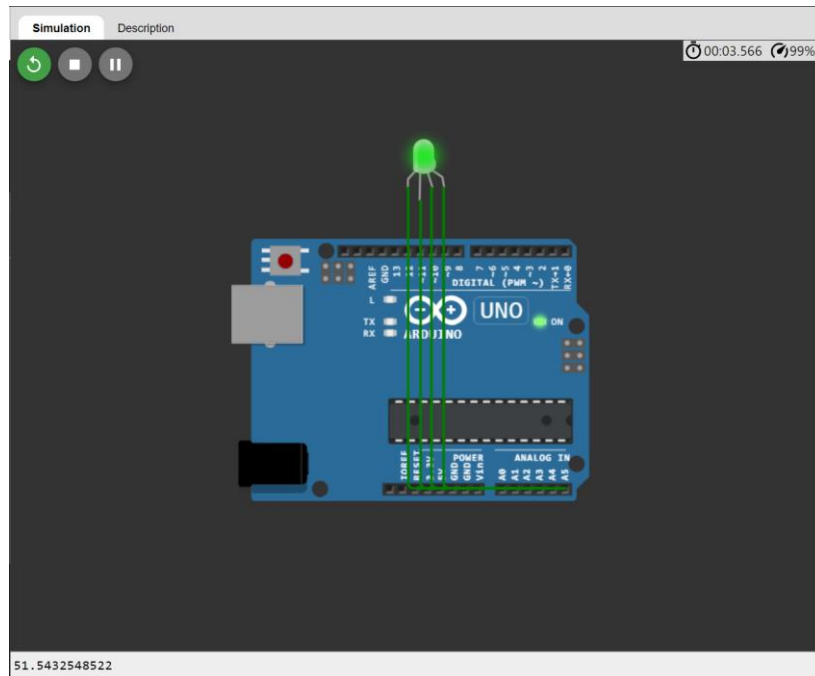
  delay(250);
  analogWrite(redPin, 0);
  analogWrite(bluePin, 0);
  analogWrite(greenPin, 0);

  delay(250);

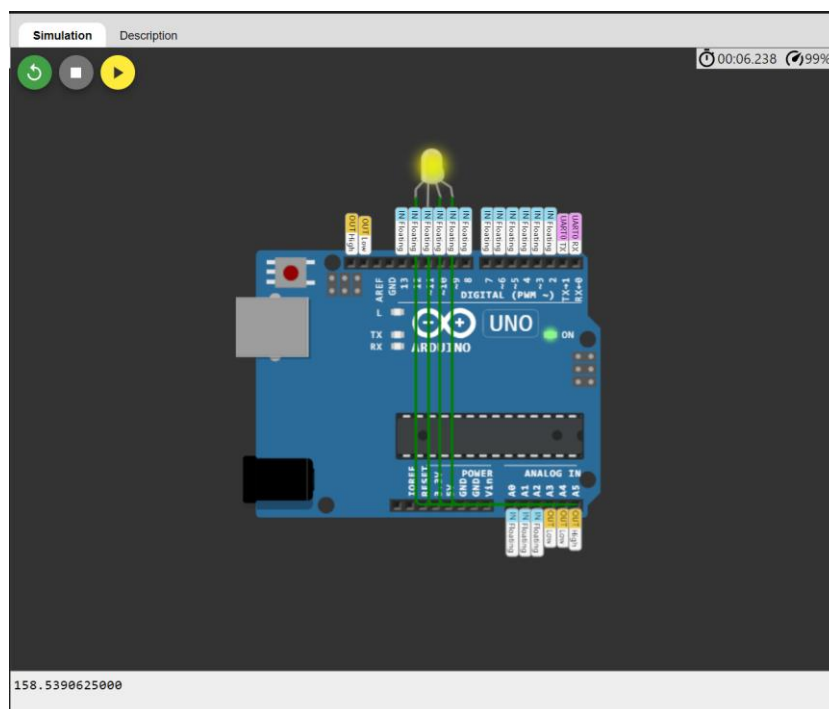
}

float multiLinReg(float a, float b, float c, float d) {
  return lrCoef[0] +
a * lrCoef[1] + b * lrCoef[2] + c * lrCoef[3] + d * lrCoef[4];
}
```

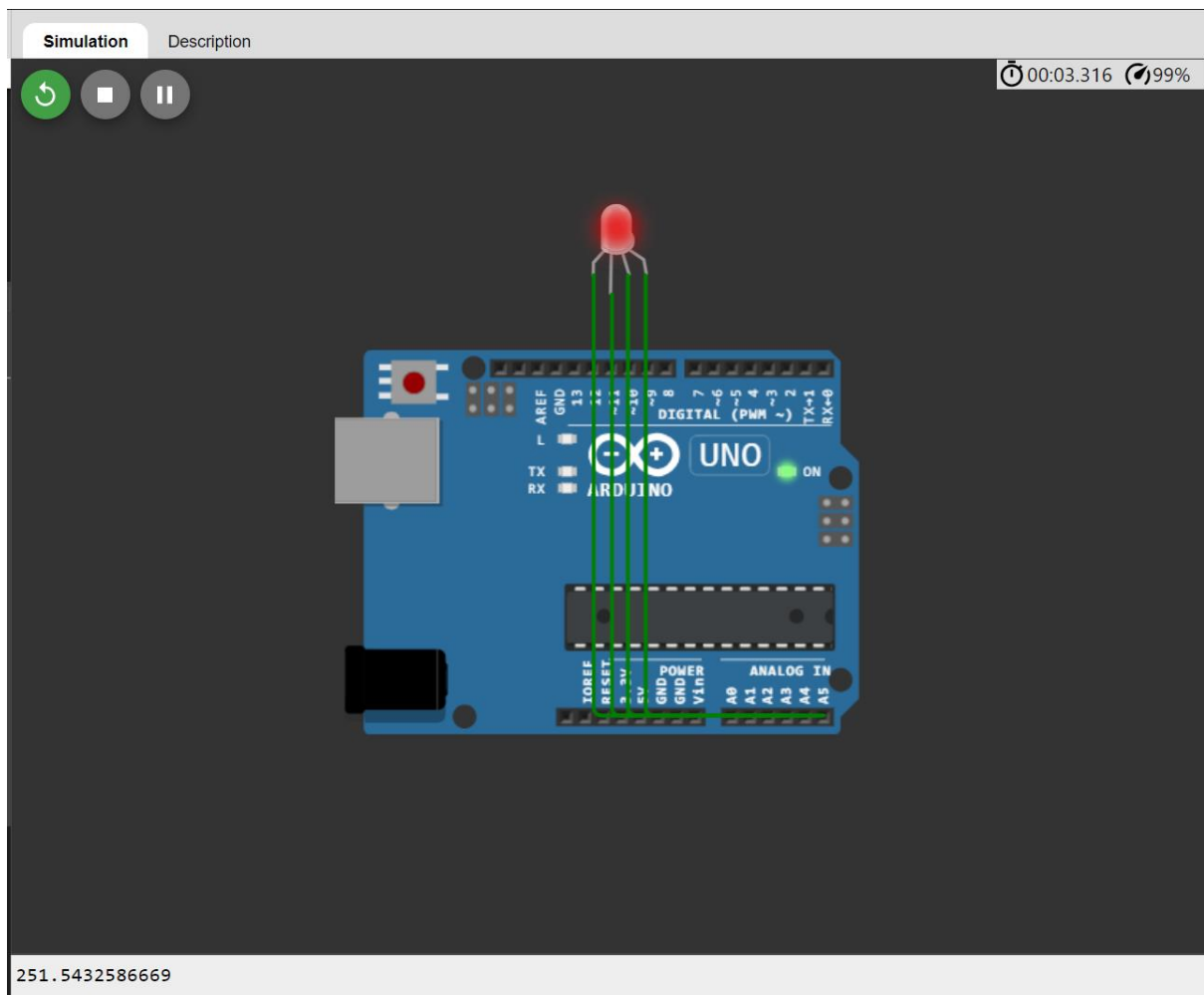
# ARDUINO SIMULATION:



GREEN AQI RANGE: 0-100



YELLOW AQI RANGE: 100-200



RED AQI RANGE: >200