



# SkyWatch

By Samuel, Alex & Marcus



A Day in the Life...

# What SkyWatch accomplishes

1. Collects weather data using Pi components
2. Active processing on Pi - No external computing
3. Plug and use functionality - Easy to set up
4. Access info from the app - Convenient



# Why Choose Us?

## Traditional Weather Systems

- ✗ Managed by large corporations
- ✗ Non transparent
- ✗ No real time data display

## Personal weather stations

- ✗ Expensive
- ✗ Hard to configure
- ✗ Semi transparent

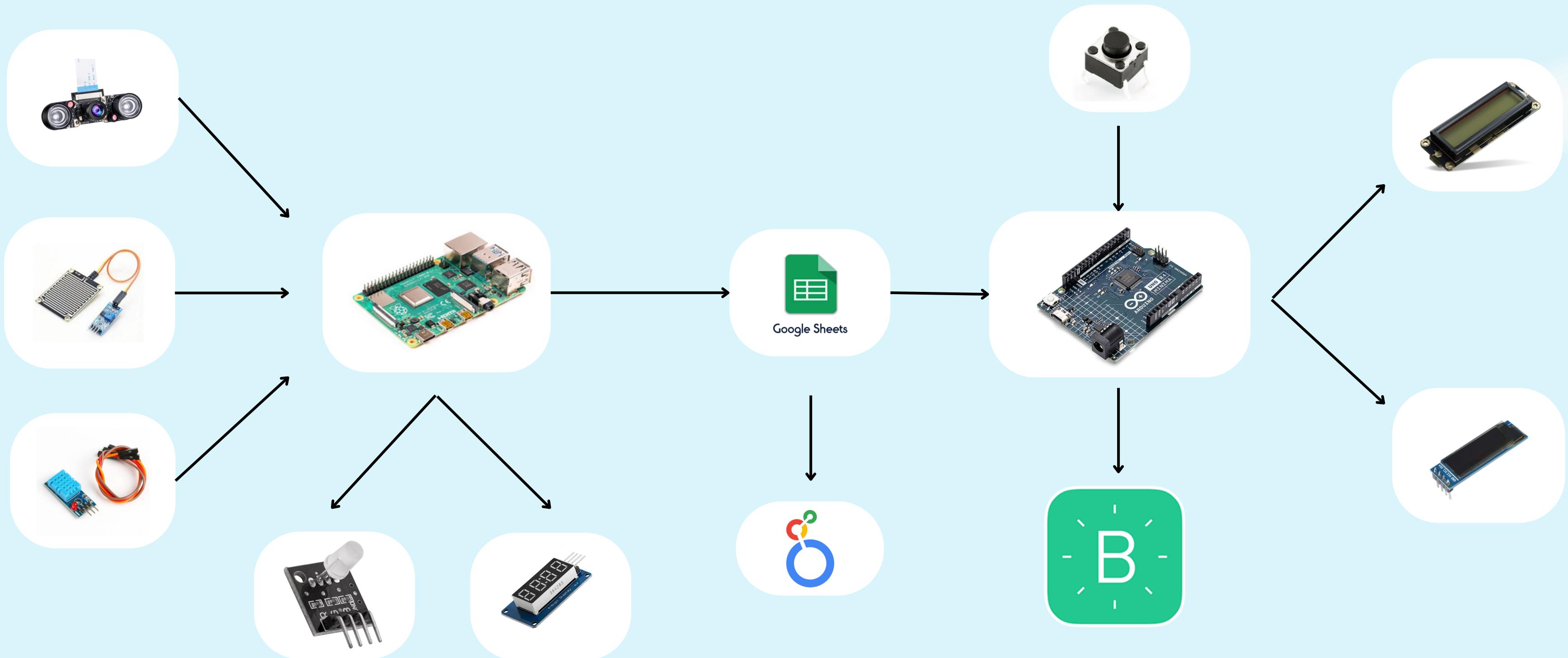
## SkyWatch

- ✓ Cheap
- ✓ Small company owned
- ✓ Completely open source
- ✓ Easy to set up
- ✓ Up-to-date data

# Making SkyWatch



# Block Diagram



# How it Works- Input

1.



Pi Camera

Takes images of the clouds.

2.

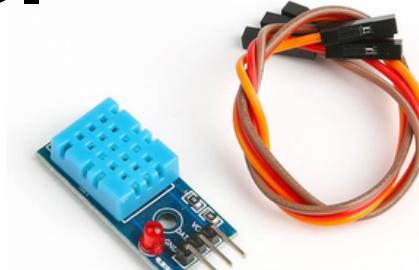


Raindrop Sensor

Detects:

- Moisture
- Precipitation

3.



DHT 11 Sensor

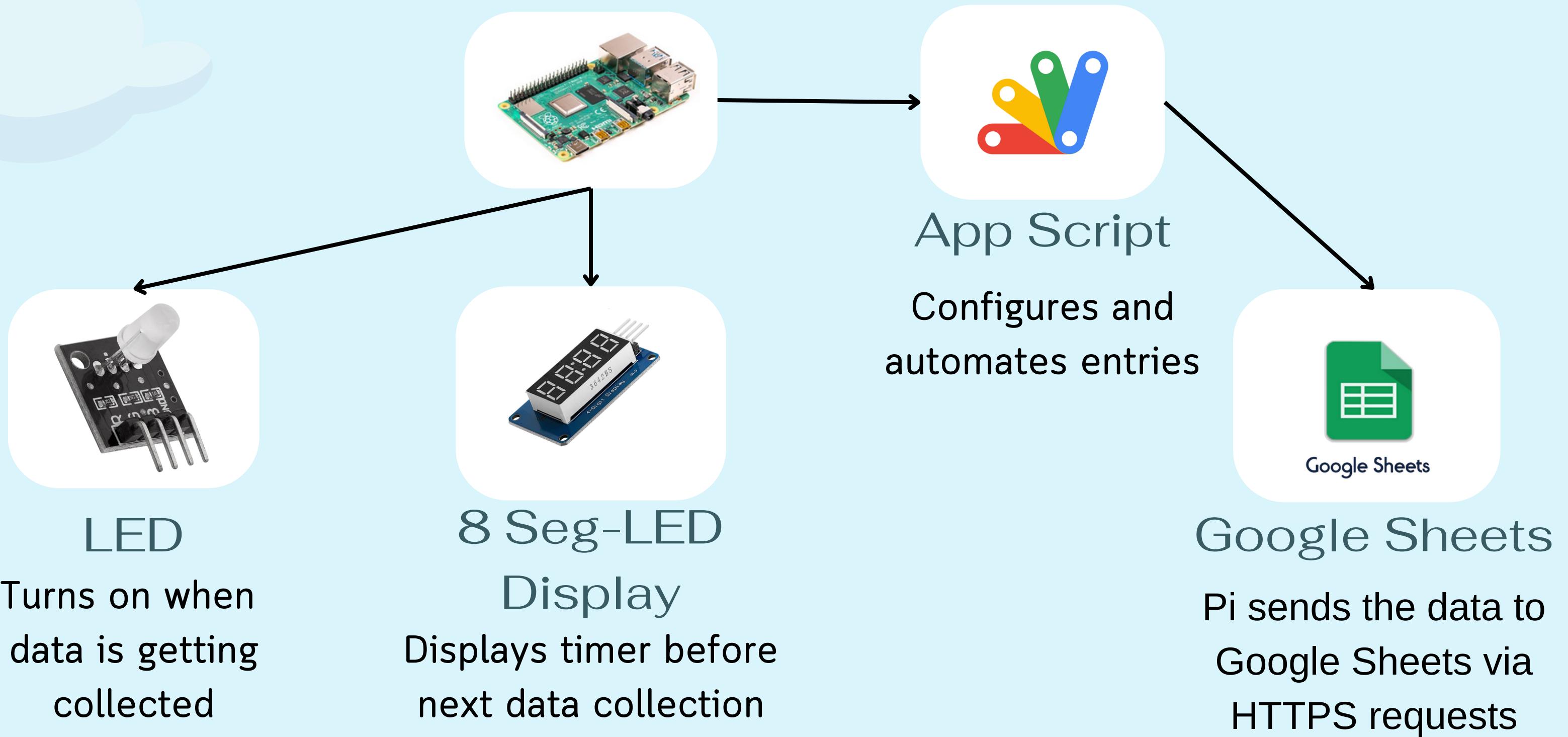
Detects:

- Humidity
- Temperature

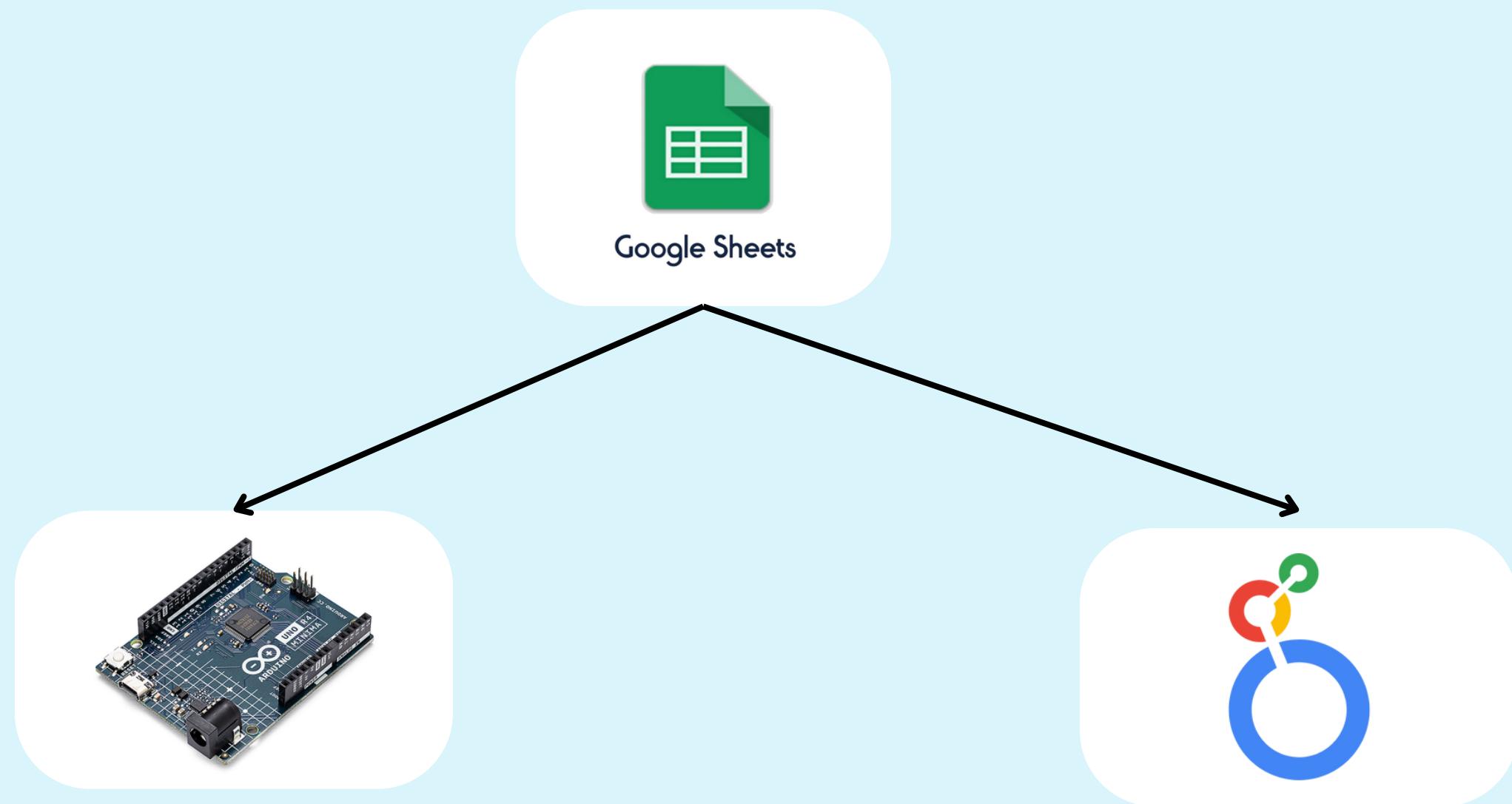
**Data collection takes place every 5 minutes.**

**The data collected is then uploaded onto Google Sheets via HTTPS.**

# How it works- Raspberry Pi



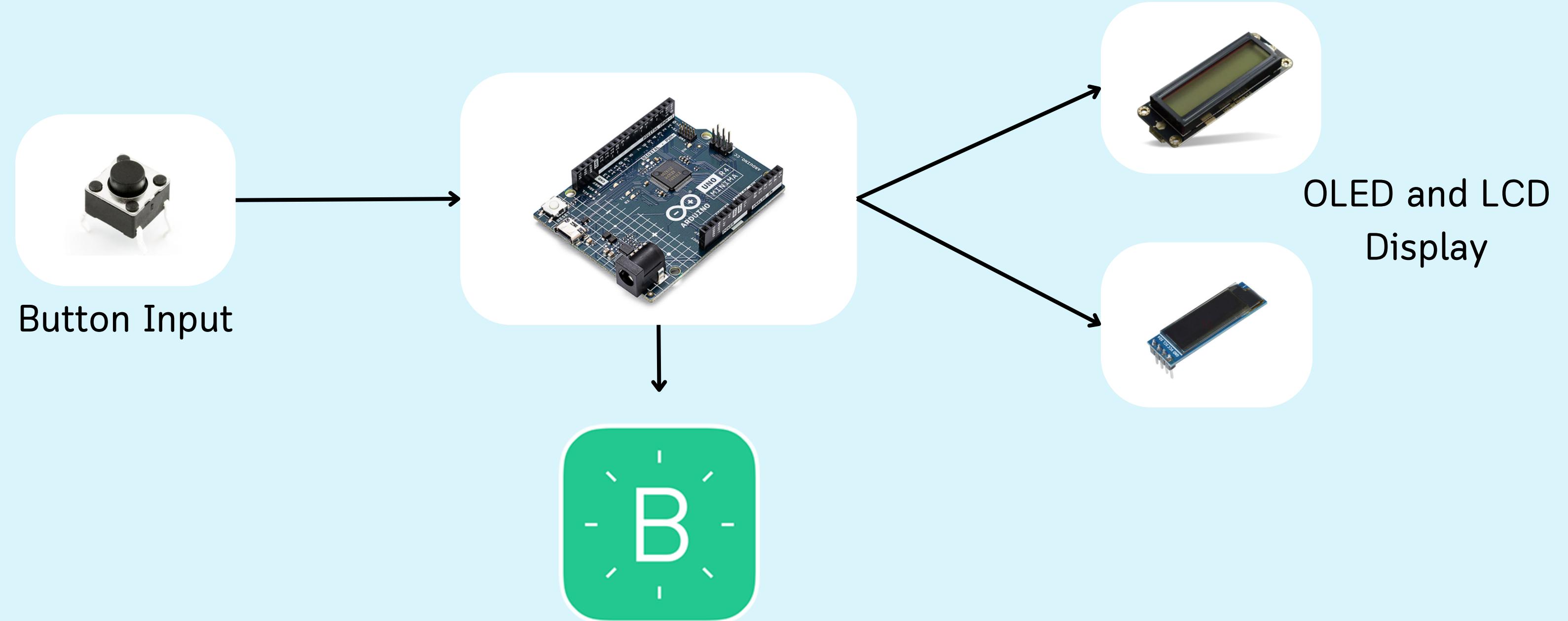
# How it works- Google Sheets



Arduino pulls data from database,  
sending data to Blynk studio  
for visualization

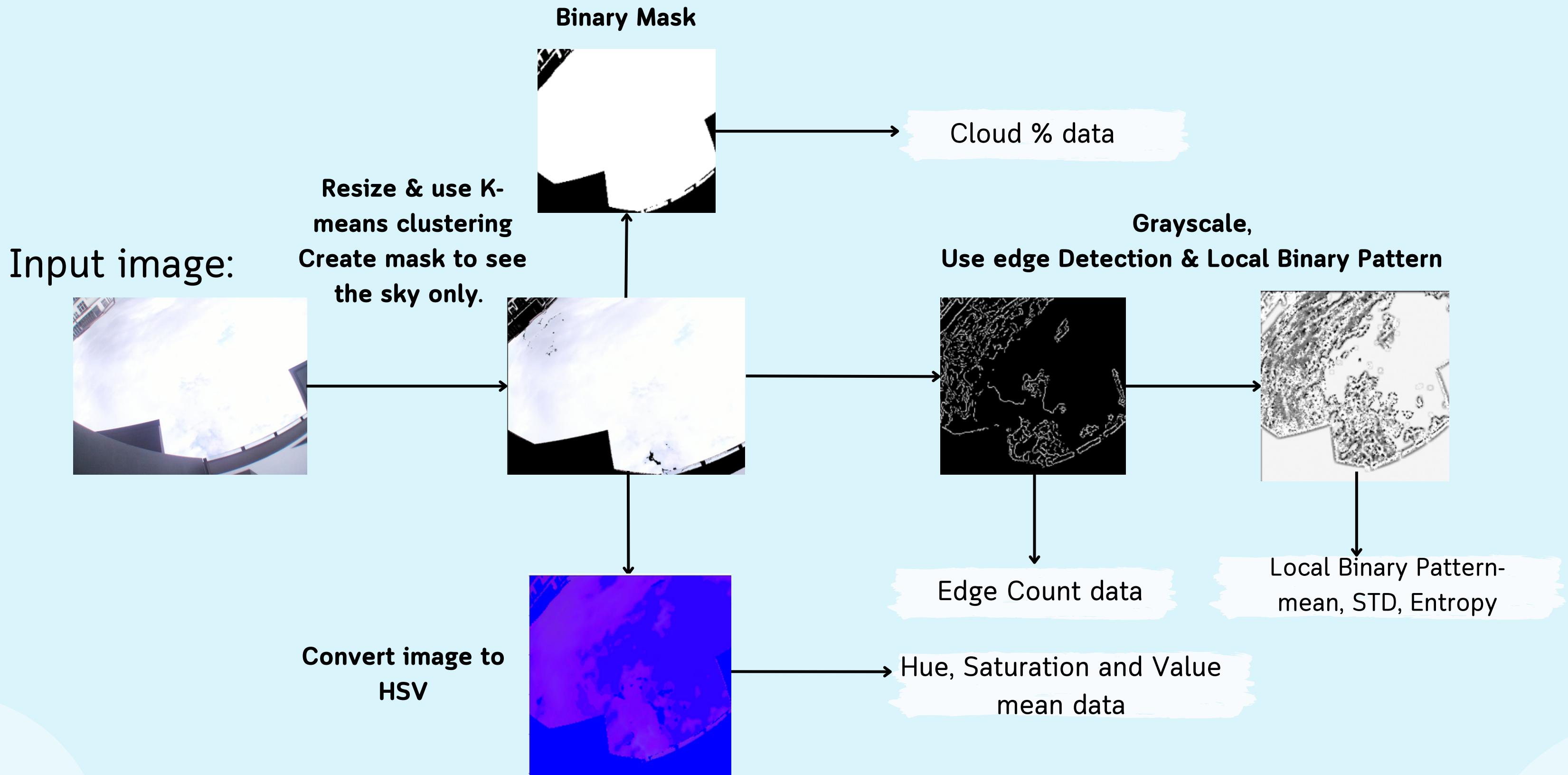
Looker Dashboard pulls  
data from database

# How it works- Arduino



Button allows for instantaneous data collection!

# Making More Data



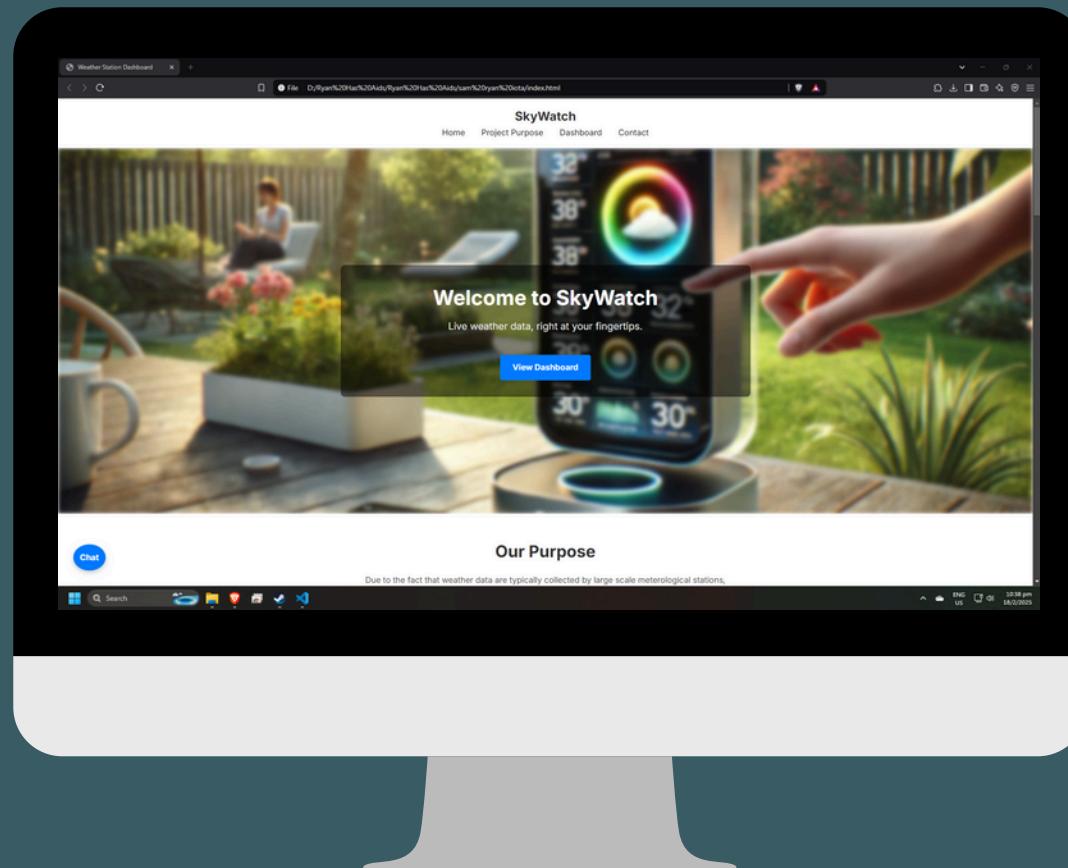
# How We Predict Rain?

```
✓ function checkRainPrediction(rowData) {  
    // Define weights  
    ✓ var weights = {  
        temperature: 0.15,  
        humidity: 0.25,  
        edgeCount: 0.15,  
        lbpMean: 0.1,  
        lbpStdDev: 0.05,  
        lbpEntropy: 0.05,  
        hueMean: 0.05,  
        saturationMean: 0.05,  
        valueMean: 0.05,  
        cloudPercentage: 0.2  
    };
```

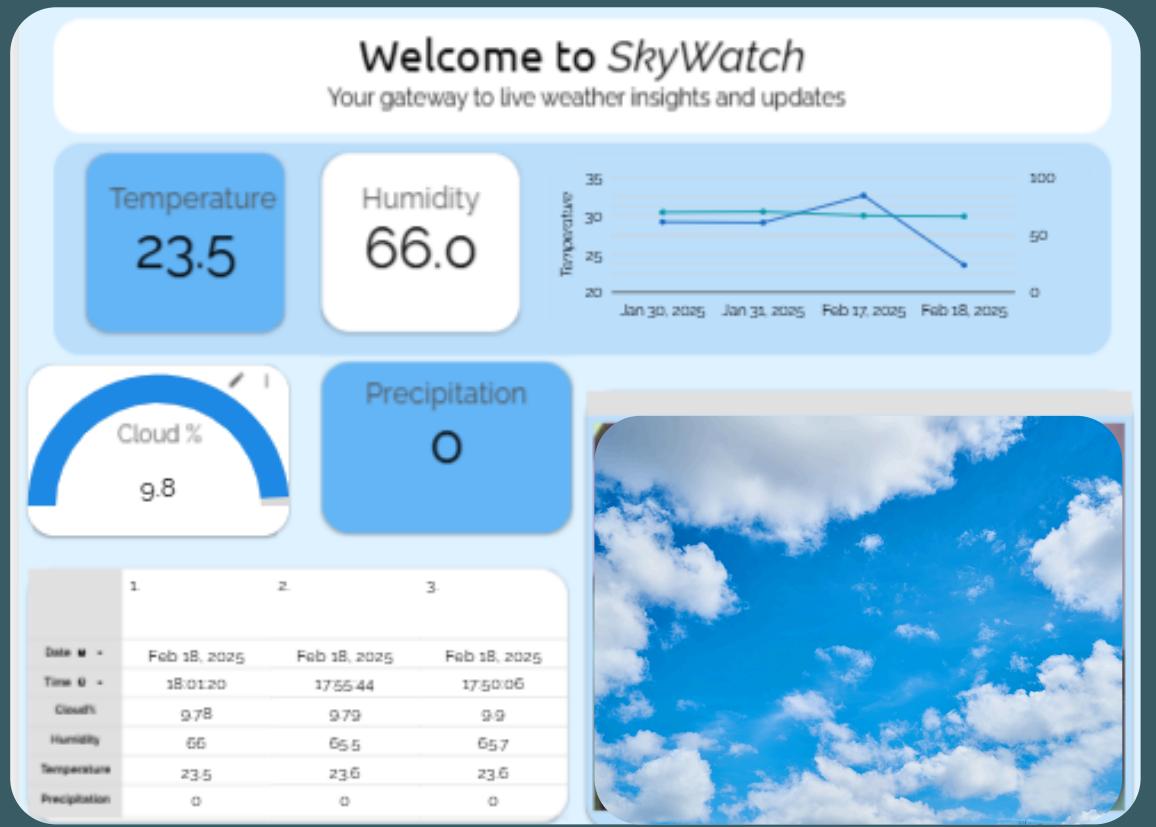
By summing the weighted values and comparing it with the threshold, we can predict if it will rain.

# The Final Results...

Website



Looker Studio Dashboard



Blynk App



# Future Implementations



# Groundwork for Machine Learning

Use machine learning to:

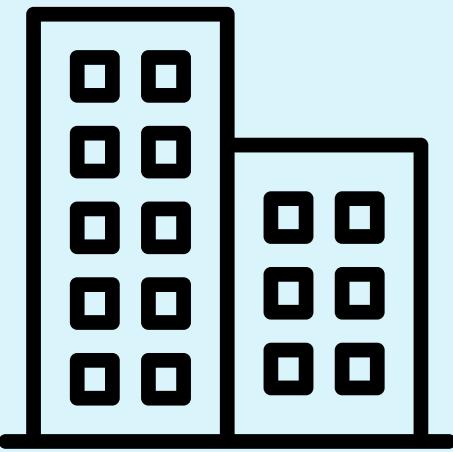
- Make predictions more reliable
- Make app more insightful
- Make more critical observations

```
1 #HSV Image
2 masked_hsv_image = cv2.cvtColor(masking, cv2.COLOR_BGR2HSV)
3 #Grayscale Image
4 masked_gray_image = cv2.cvtColor(masking, cv2.COLOR_BGR2GRAY)
5 plt.figure(figsize=(20, 10))
6
7 plt.subplot(1, 2, 1)
8 plt.title("HSV")
9 plt.imshow(masked_hsv_image)
10
11 plt.subplot(1, 2, 2)
12 plt.title("Gray")
13 plt.imshow(masked_gray_image, cmap="gray")
14 #Smooth out:
15 masked_blurred_image= cv2.GaussianBlur(masked_gray_image, (11, 11), 0)
16 plt.imshow(masked_blurred_image, cmap ="gray")
17 #Edge data:
18 masked_edges = cv2.Canny(masked_blurred_image, 100, 200)
19 plt.imshow(masked_edges, cmap = 'gray')
20 edge_count = np.sum(masked_edges > 0)
21 #LBP Data:
22 from skimage.feature import local_binary_pattern
23 radius = 3 # Radius of the circular LBP
24 n_points = 8 * radius # Number of points to consider for the circular LBP
25 lbp_image = local_binary_pattern(masked_gray_image, n_points, radius, method='uniform')
26 lbp_hist, lbp_bins = np.histogram(lbp_image.ravel(), bins=np.arange(0, n_points + 3), range=(0, n_points + 2))
27 lbp_hist_normalized = lbp_hist / lbp_hist.sum()
28 #HSV Data
29 cloud_mask = np.where((masked_hsv_image[..., 2] > 100), 1, 0)
30 cloud_pixels = masked_hsv_image[cloud_mask == 1]
31 avg_hue = np.mean(cloud_pixels[..., 0]) # Hue
32 avg_saturation = np.mean(cloud_pixels[..., 1]) # Saturation
33 avg_value = np.mean(cloud_pixels[..., 2]) # Brightness
34 print(avg_hue, avg_saturation, avg_value)
35 #Cloud% data:
36 cloud_pixels = mask.size # Count cloud pixels
37 total_pixels = masking.flat.size # Total pixels in the image
```

# Demo



# How Our Project Can Help You



## 01 Personal Use

Get real time weather updates before heading out

## 02 Companies and Corporations

Enables efficient and reliable tracking for planning outdoor events

## 03 Community & Smart Cities

Provide real-time local weather data for neighborhood awareness.