

Code

```
#define TRIGGER_PIN_0 2
```

```
#define ECHO_PIN_0 3
```

```
#define TRIGGER_PIN_90 4
```

```
#define ECHO_PIN_90 5
```

```
#define TRIGGER_PIN_180 6
```

```
#define ECHO_PIN_180 7
```

```
#define TRIGGER_PIN_270 8
```

```
#define ECHO_PIN_270 9
```

```
#define CHAMBER_WIDTH 500
```

```
#define CHAMBER_HEIGHT 600
```

```
float x = 0;
```

```
float y = 0;
```

```
void setup() {
```

```
    Serial.begin(9600);
```

```
    pinMode(TRIGGER_PIN_0, OUTPUT);
```

```
    pinMode(ECHO_PIN_0, INPUT);
```

```
    pinMode(TRIGGER_PIN_90, OUTPUT);
```

```
    pinMode(ECHO_PIN_90, INPUT);
```

```
    pinMode(TRIGGER_PIN_180, OUTPUT);
```

```
    pinMode(ECHO_PIN_180, INPUT);
```

```
    pinMode(TRIGGER_PIN_270, OUTPUT);
```

```
    pinMode(ECHO_PIN_270, INPUT);
```

```
}
```

```
void loop() {
```

```
    int distance_0 = measureDistance(TRIGGER_PIN_0, ECHO_PIN_0);
```

```
    int distance_90 = measureDistance(TRIGGER_PIN_90, ECHO_PIN_90);
```

```
    int distance_180 = measureDistance(TRIGGER_PIN_180, ECHO_PIN_180);
```

```
    int distance_270 = measureDistance(TRIGGER_PIN_270, ECHO_PIN_270);
```

```
    x += distance_0 - distance_180;
```

```
    y += distance_90 - distance_270;
```

```
    if (x < 0) {
```

```
        x = 0;
```

```
    }
```

```
    if (x > CHAMBER_WIDTH) {
```

```
        x = CHAMBER_WIDTH;
```

```
    }
```

```
    if (y < 0) {
```

```
        y = 0;
```

```
    }
```

```
    if (y > CHAMBER_HEIGHT) {
```

```
        y = CHAMBER_HEIGHT;
```

```
    }
```

```
    Serial.print("X: ");
```

```
    Serial.print(x);
```

```
    Serial.print(" cm, Y: ");
```

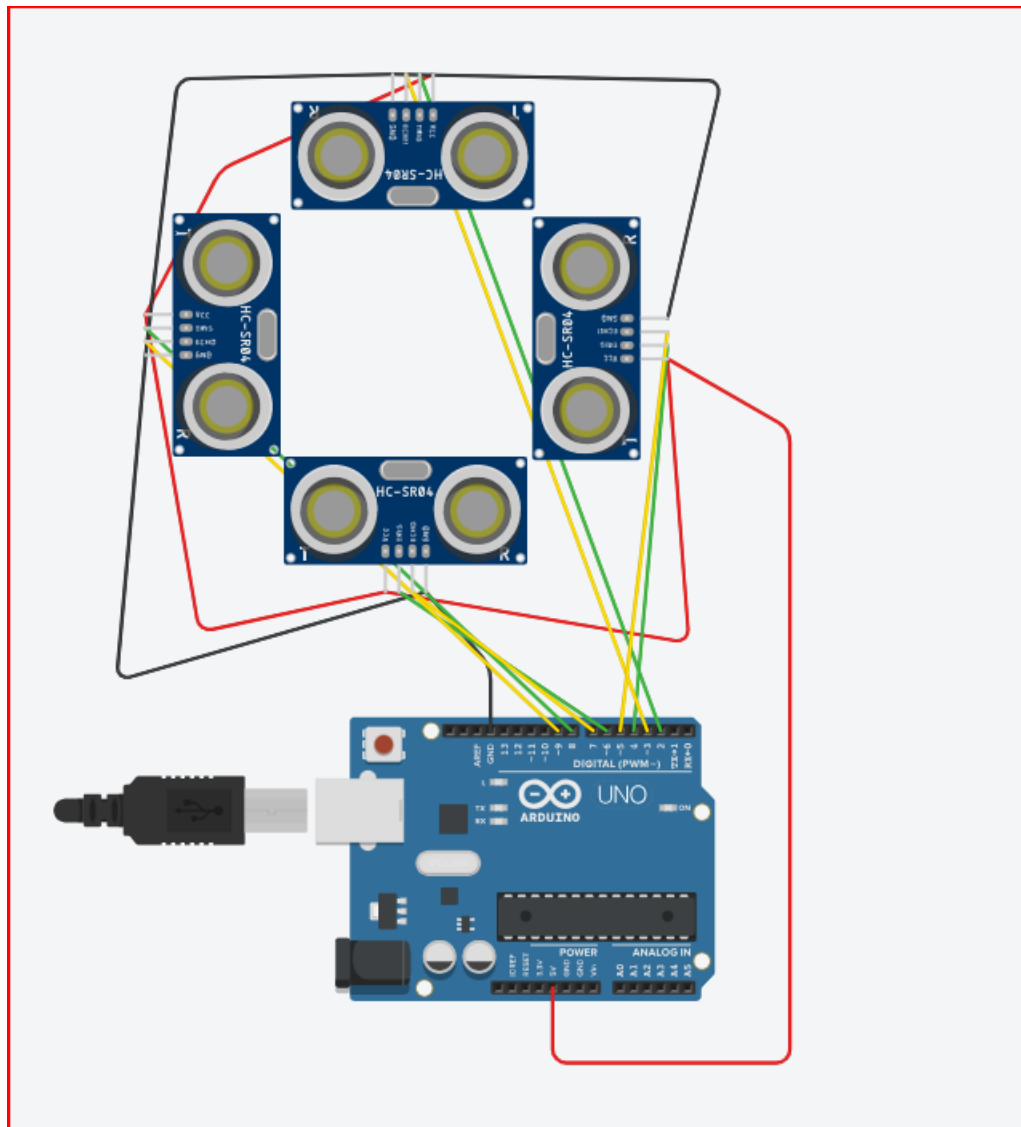
```
    Serial.print(y);
```

```
    Serial.println(" cm");
```

```
    delay(1000);  
}
```

```
int measureDistance(int triggerPin, int echoPin) {  
    digitalWrite(triggerPin, LOW);  
    delayMicroseconds(2);  
    digitalWrite(triggerPin, HIGH);  
    delayMicroseconds(10);  
    digitalWrite(triggerPin, LOW);  
  
    long duration = pulseIn(echoPin, HIGH);  
  
    int distance = duration * 0.034 / 2;  
  
    return distance;  
}
```

Schematic



Bonus

Here's an outline of what sensors may be needed and the algorithm to implement:

Sensors:

1. **Lidar or Laser Range Finder:** Lidar sensors emit laser beams and measure the time it takes for the beams to bounce back. This data is used to create a 2D or 3D map of the surroundings, including obstacles and features.
2. **IMU (Inertial Measurement Unit):** IMU sensors provide information about the robot's acceleration, angular velocity, and orientation. This data helps in estimating the robot's movement.
3. **Wheel Encoders:** These sensors measure the rotation of the robot's wheels, which can be used to estimate the robot's movement over time.
4. **Camera:** Cameras can be used for visual SLAM, where the robot captures images of its environment and extracts features to create a map and localize itself.
5. **Ultrasonic Sensors:** Ultrasonic sensors, like the ones you mentioned earlier, can provide additional distance information for close-range obstacles.

Algorithm:

1. **Extended Kalman Filter (EKF) or Particle Filter:** These are common algorithms used for SLAM. They combine sensor measurements with motion models to estimate the robot's position and simultaneously update the map.
2. **Feature-Based SLAM:** This approach involves detecting and tracking distinct features in the environment, such as corners or edges, using sensors like Lidar or cameras. The robot then builds a map based on the observed features and estimates its position relative to them.
3. **Graph-Based SLAM:** In this method, the environment is represented as a graph, with nodes representing robot poses and edges representing measurements between poses. Optimization techniques like the Gauss-Newton method or Levenberg-Marquardt algorithm are used to find the best estimates of robot poses and map features that minimize measurement errors.
4. **Visual SLAM:** If you're using cameras, Visual SLAM algorithms like ORB-SLAM or PTAM (Parallel Tracking and Mapping) can be employed. These algorithms track visual features in real-time to estimate both the robot's pose and the 3D map.
5. **Monte Carlo Localization (Particle Filter):** This is a popular approach for mobile robot localization. It involves maintaining a set of particles that represent possible positions of the robot. The particles are updated based on sensor measurements and motion models.
6. **Mapping Techniques:** For map creation, techniques such as occupancy grid mapping or grid-based mapping can be used to build a representation of the environment based on sensor data.