

# Probabilistic Robotics Map Building in 2D

양정연

2020/12/10

# Map Update Strategy with Inverse measurement(or Sensor) model

$$l(m_i | x_{1:t}, z_{1:t}) = l(m_i | x_t, z_t) + l(m_i | x_{1:t-1}, z_{1:t-1}) - l(m_i)$$

$$l_t = l(m_i | x_t, z_t) + l_{t-1} - l_0(m_i)$$

$$= l(m_i | x_t, z_t) + l_{t-1} - l_0$$

$$t > 0: p(m_i | x_{1:t}, z_{1:t}) \rightarrow l_t$$

$$t = 0: p(m_i) \rightarrow l_0$$

- If we calculate  $l(m_i | x_t, z_t)$ , then we update a map
- It is called

$l(m_i | x_t, z_t) \rightarrow p(m_i | x_t, z_t) =$  Inverse measurement model

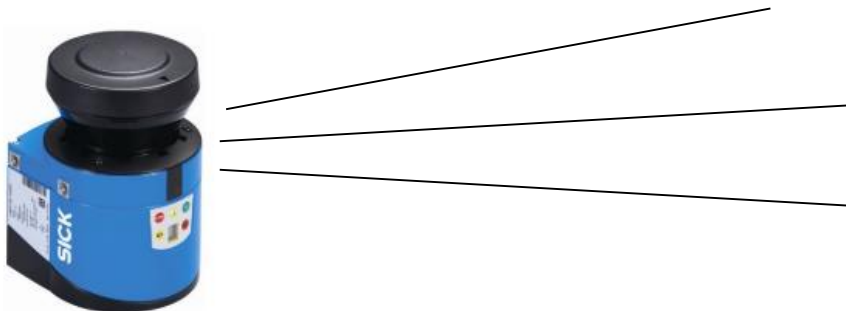
- From the current position,  $x$  and the current measurement,  $z$ , we estimate prob. Of whether a map  $m_i$  is empty or occupied.
- It is different with a classification probability.

# Extend Map Update into 2Dim

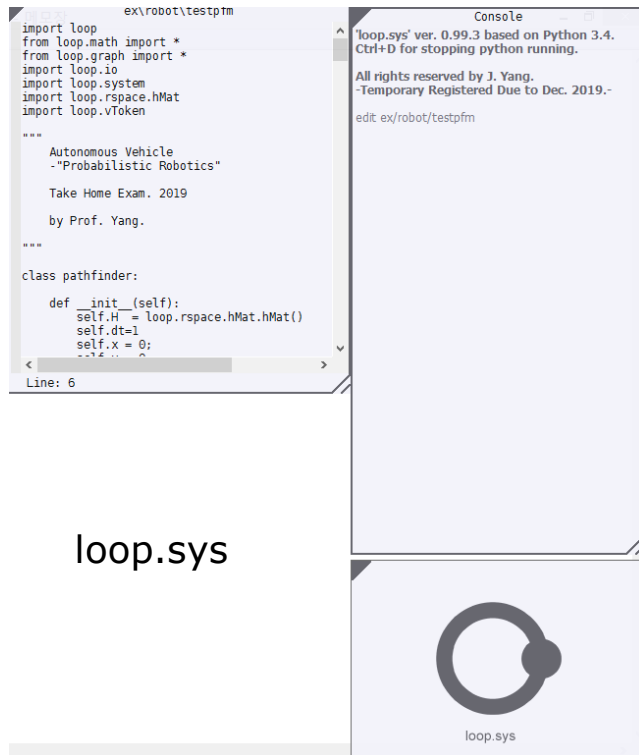
- Map update with Inverse Sensor Model

$$l_t = l(m_i | x_t, z_t) + l_{t-1} - l_0$$
$$= \text{InverseSensorModel}(m, x, z) + l_{t-1} - l_0$$

- SLAM uses Distance Metric Sensor like laser scanner
- Try to understand Distance Metric Sensor



# Two Programs are Used



The screenshot shows a Python IDE with the following code in the editor:

```

import loop
from loop.math import *
from loop.graph import *
import loop.io
import loop.system
import loop.rspace.hMat
import loop.vToken

"""
Autonomous Vehicle
- "Probabilistic Robotics"
Take Home Exam. 2019
by Prof. Yang.
"""

class pathfinder:
    def __init__(self):
        self.H = loop.rspace.hMat.hMat()
        self.dt=1
        self.x = 0;
        self.y = 0;

```

The console window displays the following output:

```

'loop.sys' ver. 0.99.3 based on Python 3.4.
Ctrl+D for stopping python running.

All rights reserved by J. Yang.
-Temporary Registered Due to Dec. 2019.-

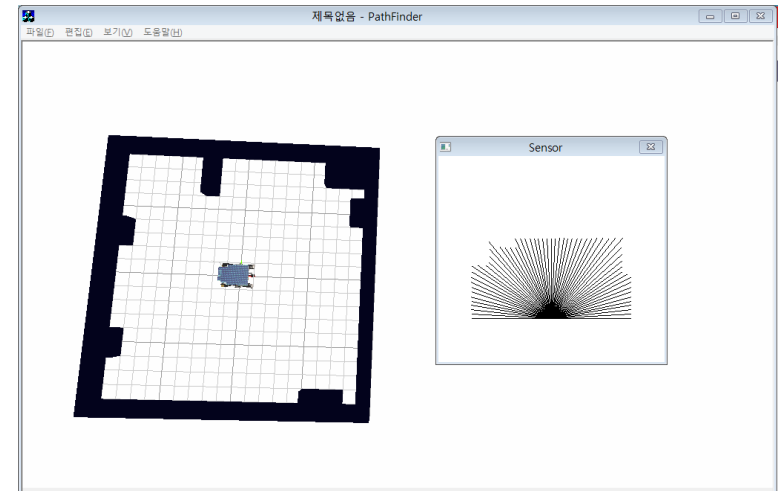
edit: ex/robot/testpfm

```

Below the code editor is a circular logo with a dot in the center, labeled "loop.sys".

→  
Move(x,y,q)

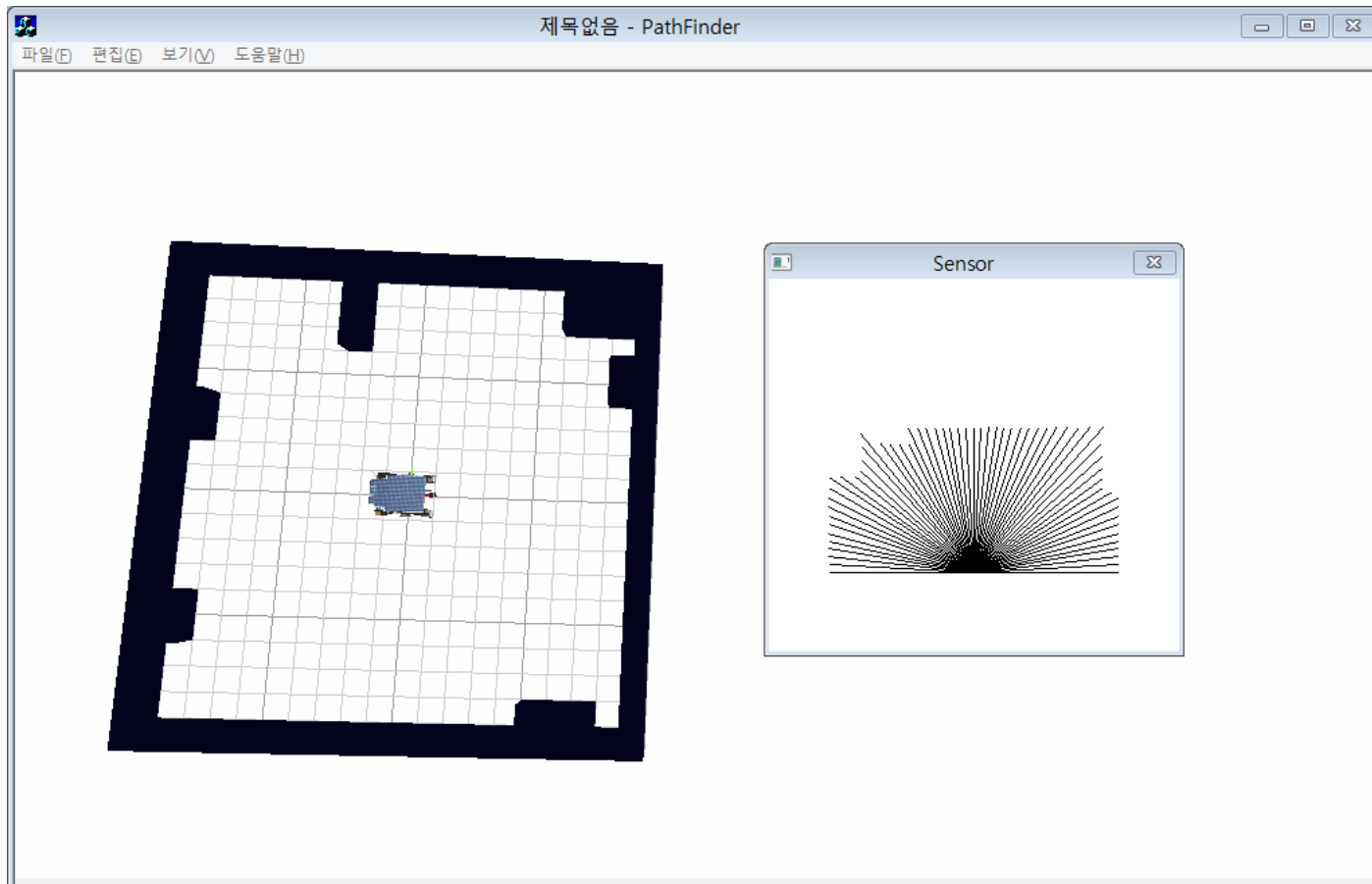
←  
Sensor  
Distance  
information



Path finder 3D environment  
Pathfinder.exe

- loop.sys is a Python-based program is communicated with Pathfinder.exe
  - Internal process communication (ICP)

# Example: Path Finder with Laser Scanner



- Add walls and Distance Sensor

# 1. Wall and Sensor Model from XML

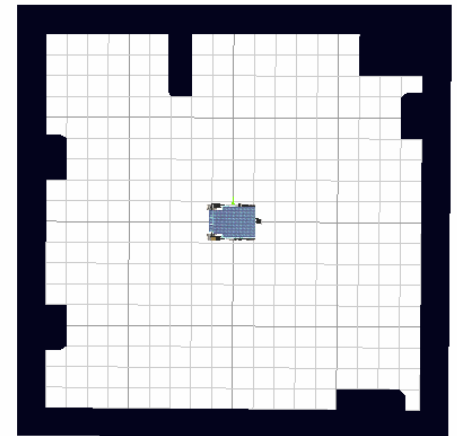
- Define wall, grid, and sensor model from “config.xml”
- Redefine Grid world, with 20x20

```

<config>
  <map path="map.bmp">
    <space w=20 h=20 z=20/>
    <grid w=21 h=21/>
  </map>

  <sensor start=-90 end=90 resolution=3>
  </sensor>
</config>

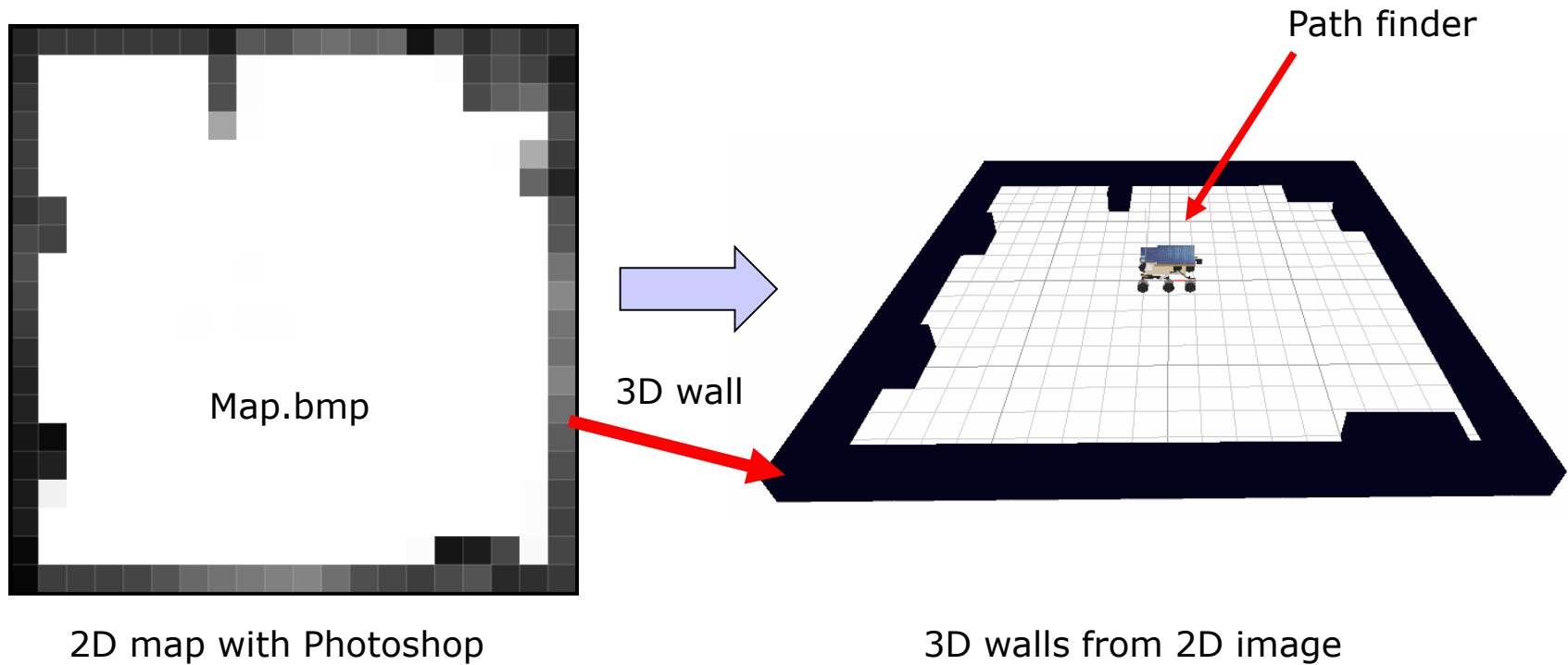
```



W=20  
Grid =21

## 2. Add Walls

- Use a virtual map (“map.bmp”)



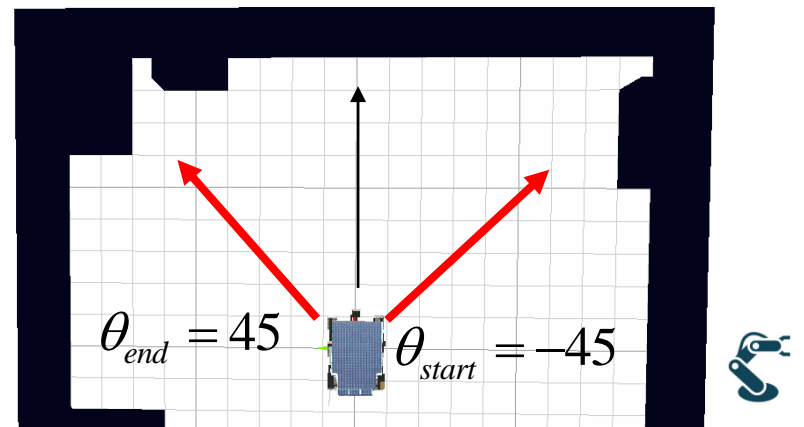
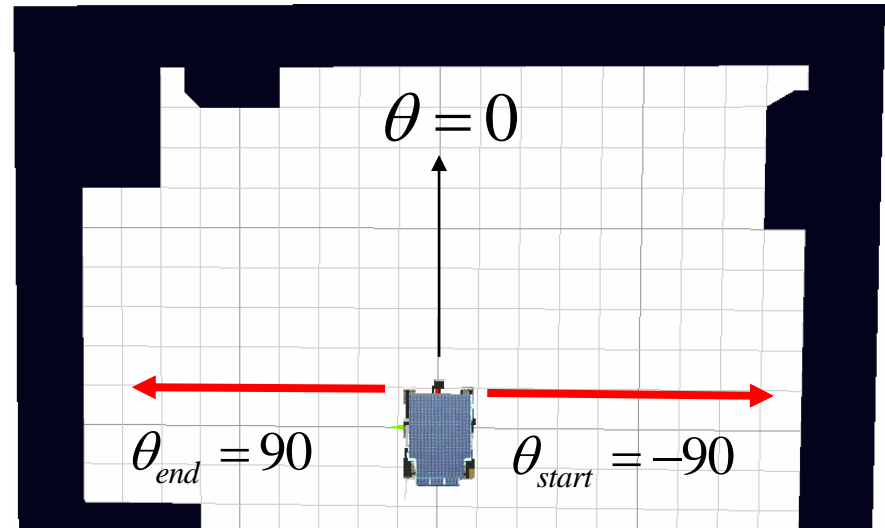
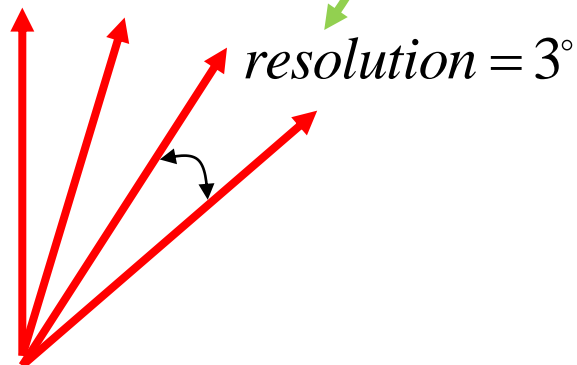
## 3. Virtual Sensor

- Define Sensor type from “config.xml”

```

<config>
  <map path="map.bmp">
    <space w=20 h=20 z=20/>
    <grid w=21 h=21/>
  </map>
  <sensor start=-90 end=90 resolution=3>
</sensor>
</config>

```

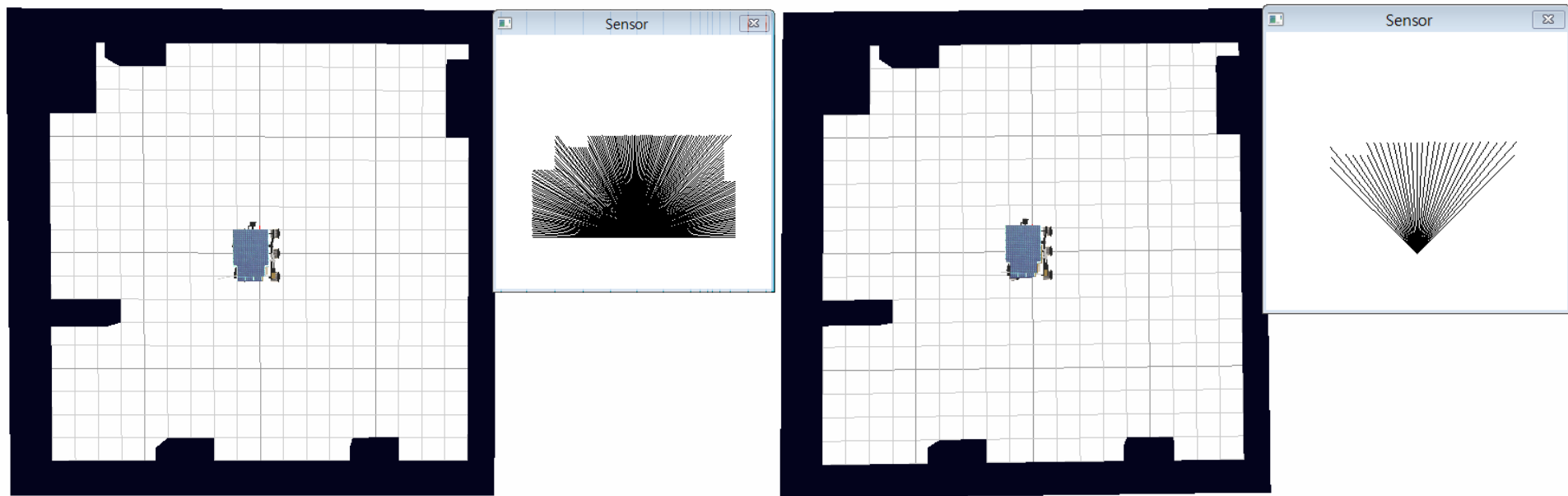




### 3. Virtual Sensor Data

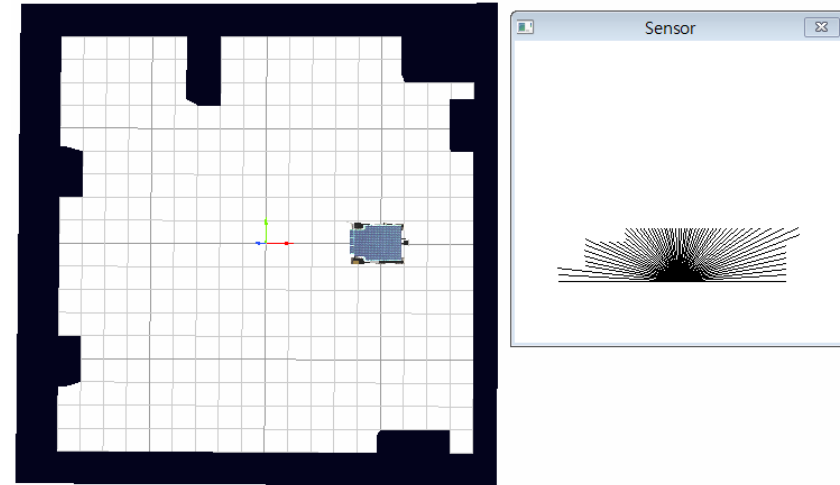
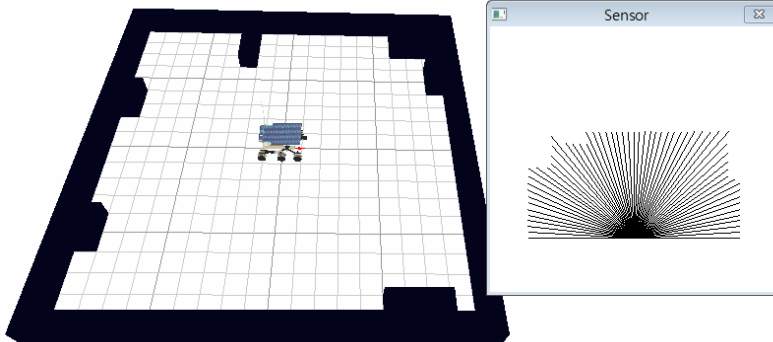
```
<sensor start=-90 end=90 resolution=1>  
</sensor>
```

```
<sensor start=-45 end=45 resolution=3>  
</sensor>
```

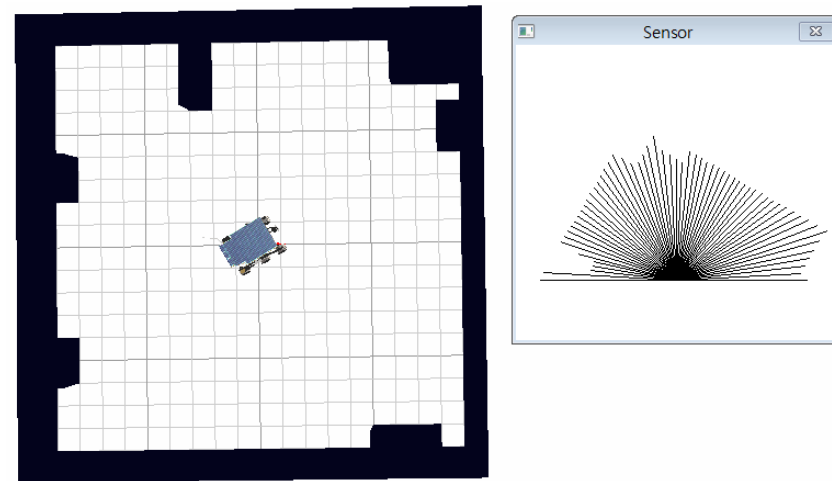
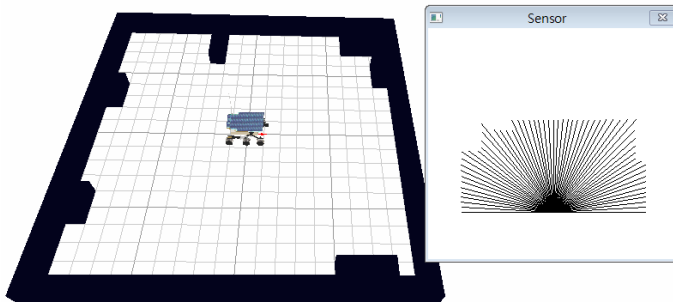



- High Quality Sensor has 0.5 degree resolution
  - Distance error is about 1mm.

# Ex1) Distance Metric with Robot Movements



- `testpfm.pfm.move(0,0,0) → testpfm.pfm.move(5,0,0)`



- `testpf.pfm.move(0,0,0) → testpf.pfm.move(0,0,30)`<sup>10</sup> 

# 2D Map Update

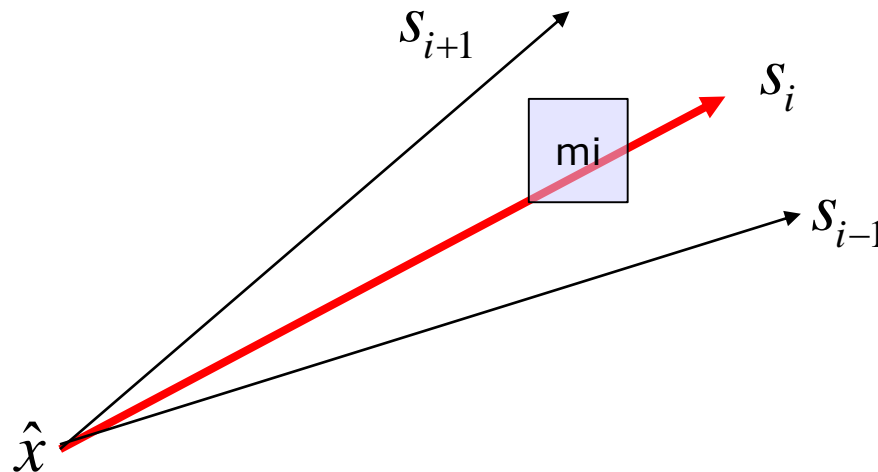
- Log Odds-based Map Update

$$l(m_i | x_{1:t}, z_{1:t}) = l(m_i | x_t, z_t) + l(m_i | x_{1:t-1}, z_{1:t-1}) - l(m_i)$$

$$\therefore l_t = l(m_i | x_t, z_t) + l_{t-1} - l_0$$

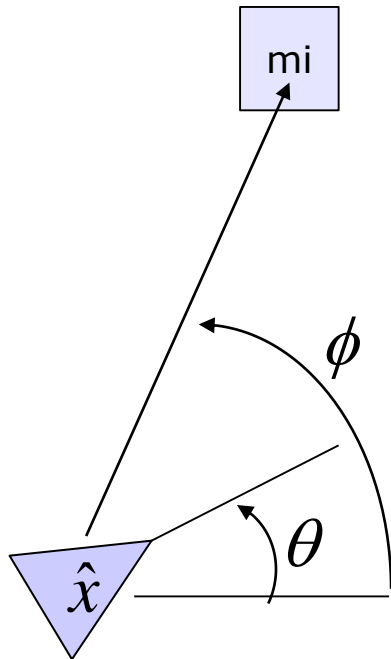
$$= \text{InverseSensorModel} + l_{t-1} - l_0$$

- Suppose that One Map,  $m_i$  with one Distance Sensor,  $s_i$



# Check Condition $m_i$ and $s_i$

- 1. Find the angle,  $\phi$        $\phi = \text{atan2}(m_{ix} - x, m_{iy} - y)$



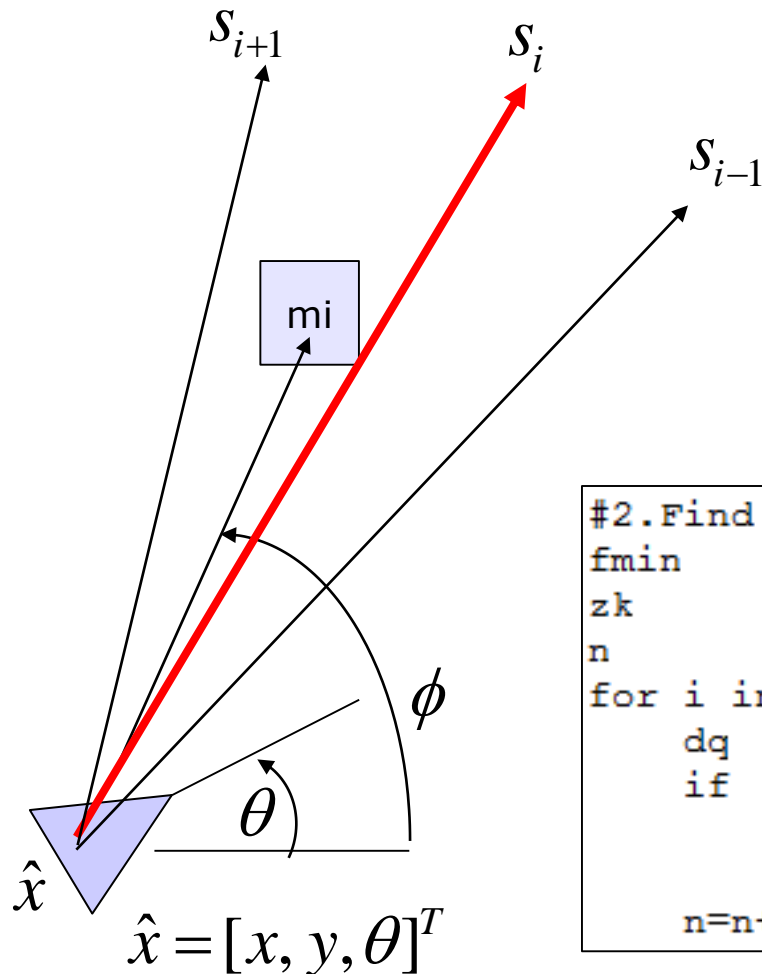
$$\hat{x} = [x, y, \theta]^T \quad \theta : \text{heading angle}$$

```
#1. Get angle from a robot to map, mi
dx = mi.x-self.x
dy = mi.y-self.y
r = sqrt(dx*dx+dy*dy)
pi = DEG(math.atan2(dy,dx))
pi = DPI(pi)
```

$$r = \|m_i - \hat{x}\| = \sqrt{(m_{i,x} - x)^2 + (m_{i,y} - y)^2}$$

# Check Condition $m_i$ and $s_i$

- 2. Find the best closest Sensor



$$i = \arg \min(|\theta + s_i - \phi|)$$

$$\rightarrow s_i$$

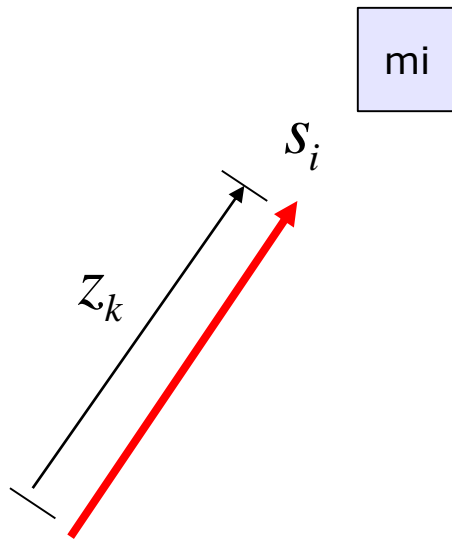
$$z_k = \text{norm} \left| \begin{array}{c} \nearrow \\ | \end{array} \right|$$

```
#2.Find the closest Distance
fmin    = 1e10
zk      = 0;
n       = 0;
for i in range(self.s,self.e+self.dq,self.dq):
    dq   = ABS(pi-i)
    if (fmin>dq):
        fmin    = dq
        zk      = dist[n]
    n=n+1
```

# Use Sensor Model

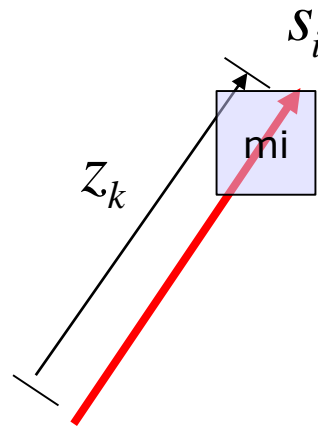
## 3. Check Distance

**Case 1) We Don't Know it**



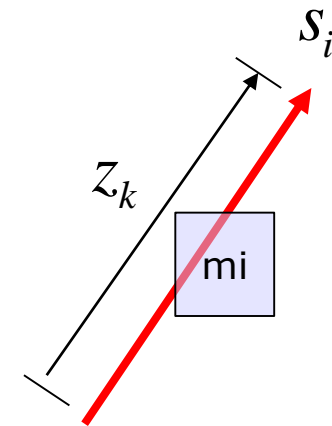
Map, mi is out of sensor area

**Case 2) Occupancy Update**



Map, mi is similar to sensor data

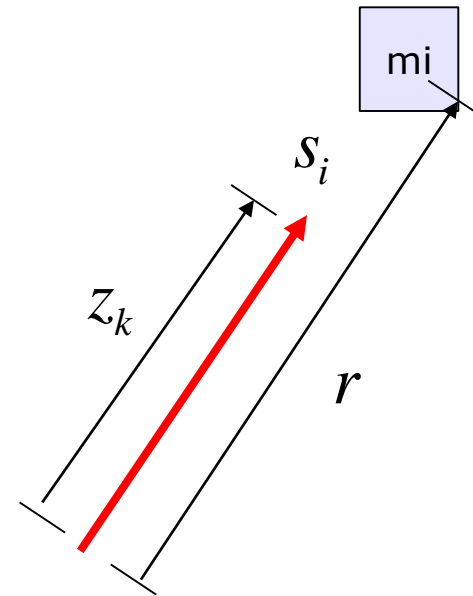
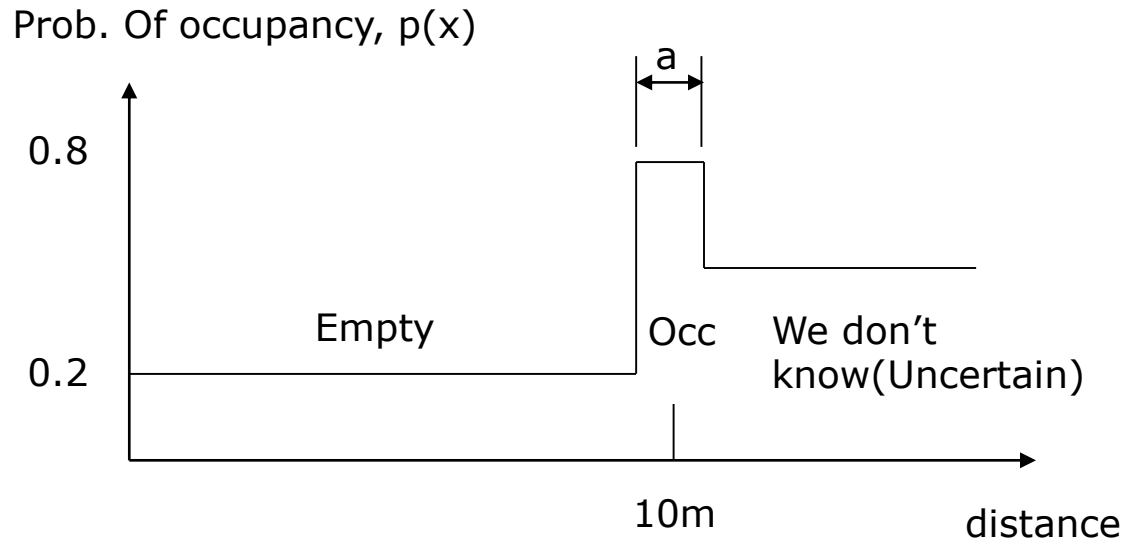
**Case 3) Emptiness Update**



Map, mi is inside of sensor area



# Sensor Probability Model



```
#3. check distance=(o-mi) < zk
if (r>zk+self.a/2):
    return self.lo    → Uncertain case

if (ABS(zk-r)<self.a/2):
    return self.lo    → Occupant case

if (r<zk):
    return self.le;   → Empty case
```

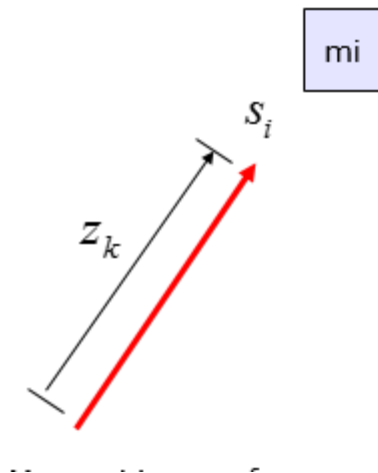


# Use Sensor Model

## 3. Check Distance

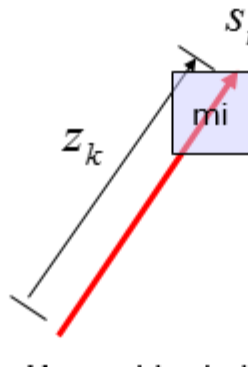
$$\log \text{ Odds} : l_0(m) = \log \frac{p(m)}{1-p(m)}$$

**Case 1)  
Uncertain**



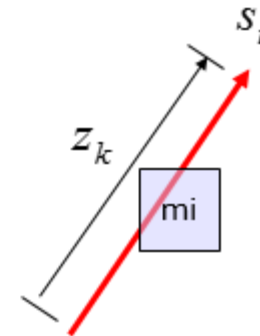
$$l_0 = \log \frac{0.5}{1-0.5} = 0$$

**Case 2)  
Occupancy Update**



$$l_{occ} = \log \frac{0.8}{1-0.8}$$

**Case 3)  
Emptiness Update**



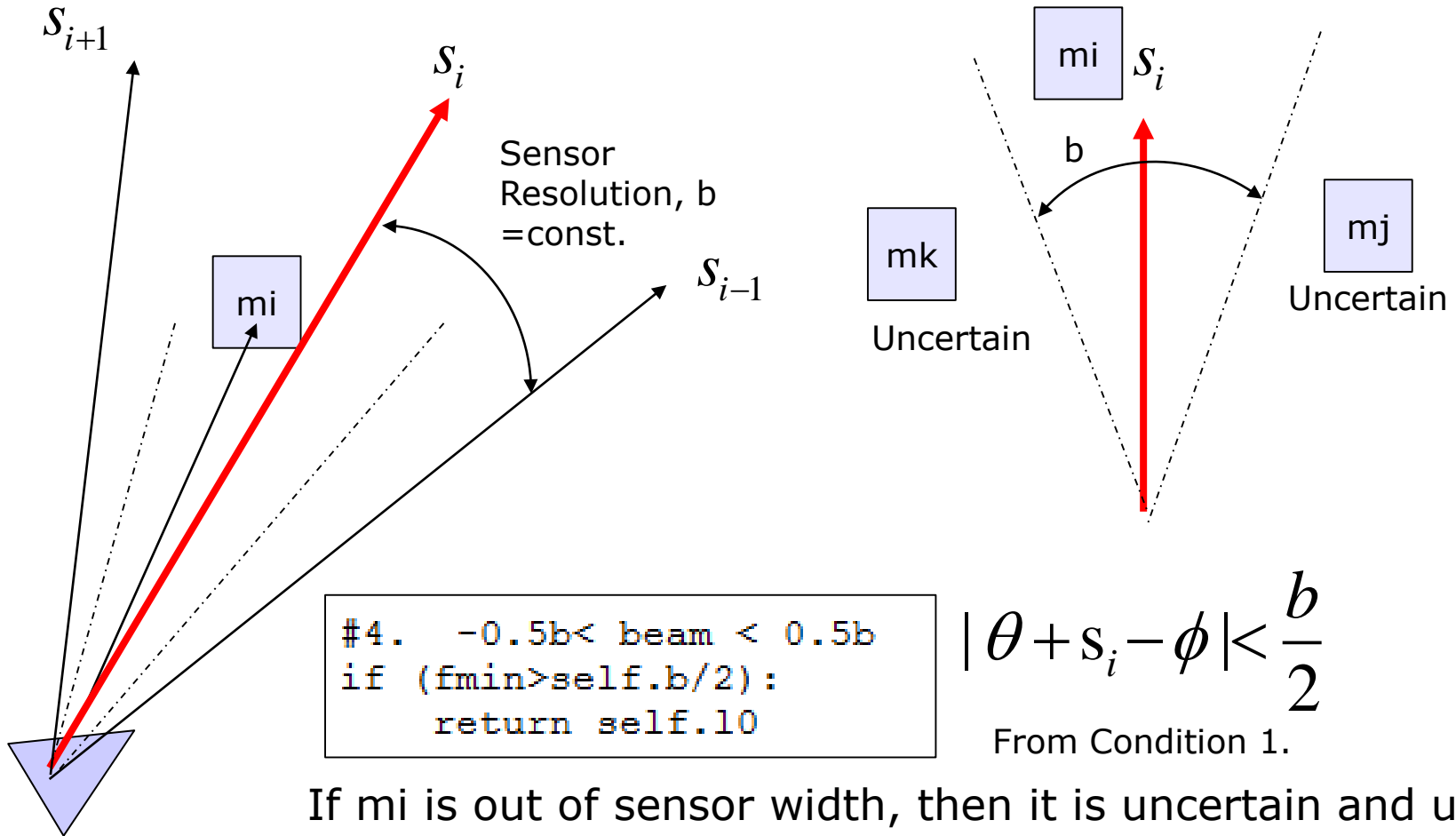
$$l_{empty} = \log \frac{0.2}{1-0.2}$$





# Use Sensor Model

## 4. Check Sensor Width



## Ex2) Map Update with Linear Motion

- Map Update for every  $m_i$  ( $0 < i < 20$  and  $0 < j < 20$ )

```
def update(self, x, y, q, dist):
    self.x = x
    self.y = y
    self.q = q

    for j in range(0, self.h):
        for i in range(0, self.w):
            mi = self.m[j*self.w+i];
            mi.l = mi.l + self.invm(m_i, dist) - self.l_0;
            mi.p = 1 - 1 / (1 + exp(mi.l));
            self.pm[j+1, i+1] = mi.p;
    surf(self.pm)
```

$$l_t = l(m_i | x_t, z_t) + l_{t-1} - l_0 = \text{InverseSensorModel} + l_{t-1} - l_0$$



## Ex2) testpfm.py

```

"""
    Example
"""
def init():
    pf.reset();

def getdist():
    st = loop.sys.recv("loop.sys")
    tok= loop.vToken.vToken(st)
    tok.Separator("\n")
    ret = []
    for i in range(tok.GetSize()):
        noise = 0.5*randn()
        ret.append(tok.f(i)+noise)
    return ret;

def updatemap():
    d = getdist()
    ms.update(pf.x,pf.y,pf.q, d)

```

Get sensor data from pathfinder

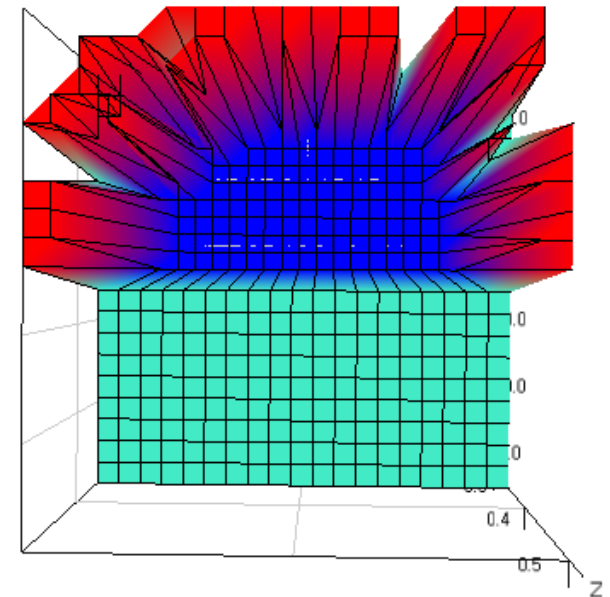
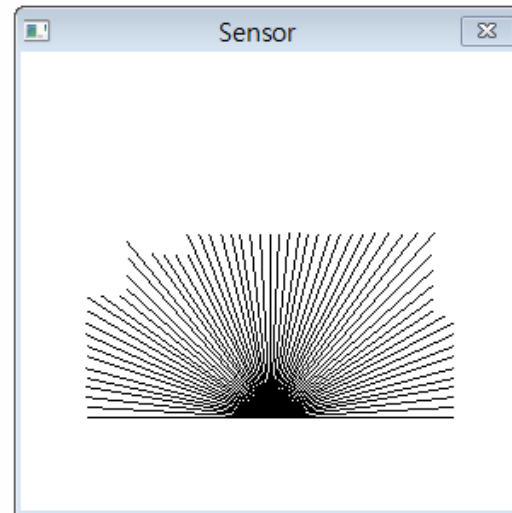
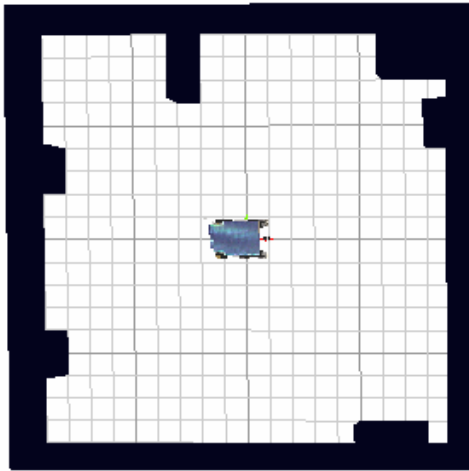
Sensor Noise with 0.5 Gaussian

Start angle=-90  
End angle=90  
Resolution=3 deg.  
Thus,  $(90 - (-90)) / 3$   
→ 60 + 1 (end angle)  
ret=61 Distance Array

Whenever pf moves,  
Call updatemap()



## Ex2) testpfm.py

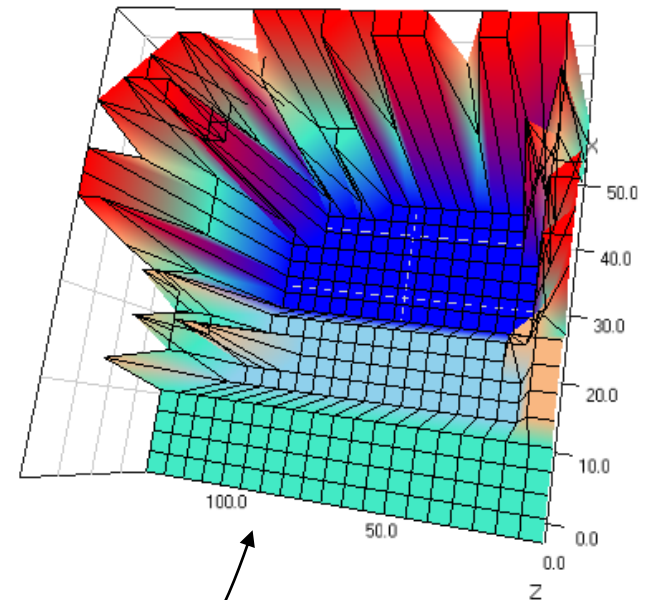
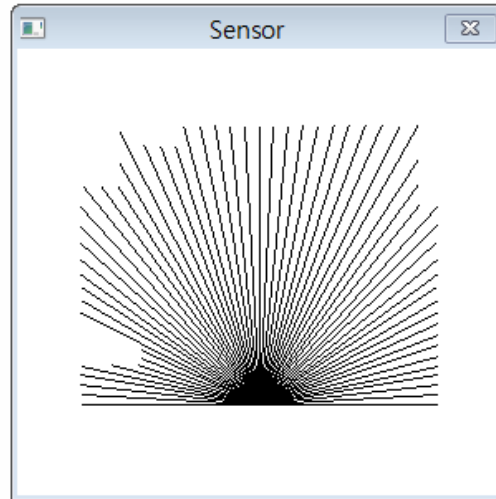
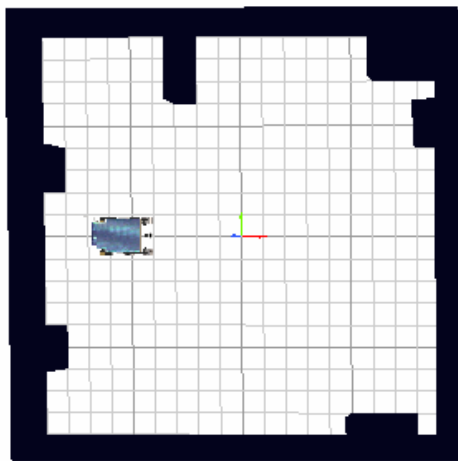


```
import testpfm as t
reload(t)
('Connected to Pathfinder.',)
<module 'testpfm' from 'C:\WWWL\WWWDR\WWWI
```

```
t.pf.move(0,0,0)
t.updatemap()
```



## Ex2) testpfm.py move(-5,0,0)



```
t.pf.move(-5,0,0)  
t.updatemap()
```

# HW) Complete Your Map Building

- Testpfm.py DOES NOT consider heading angle
  - Rotation is NOT considered.
- Complete your mapping with regarding to Rotation.
  - Think heading angle..



# HW 2) Complete Your Map Building

- Testpfm.py DOES NOT consider heading angle
  - Rotation is NOT considered.
- Complete your mapping with regard to Rotation.
  - Think heading angle..

