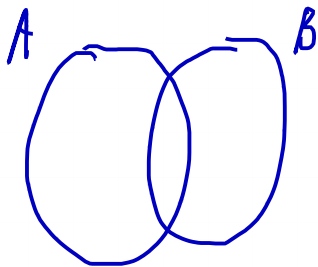


LOCALIZATION

BAYES THEOREM



$P(A|B) = P(A \cap B) / P(B)$
prob. of event A given that event B has happened.

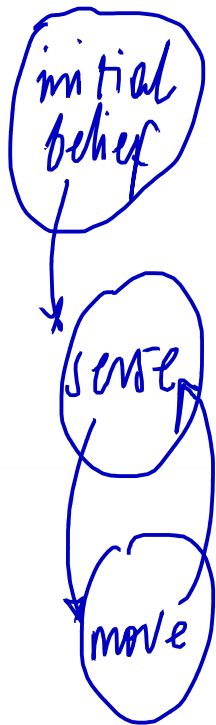
$$\left. \begin{aligned} \textcircled{1} P(A|B) &= P(A \cap B) / P(B) \\ \textcircled{2} P(B|A) &= P(A \cap B) / P(A) \end{aligned} \right\}$$

$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

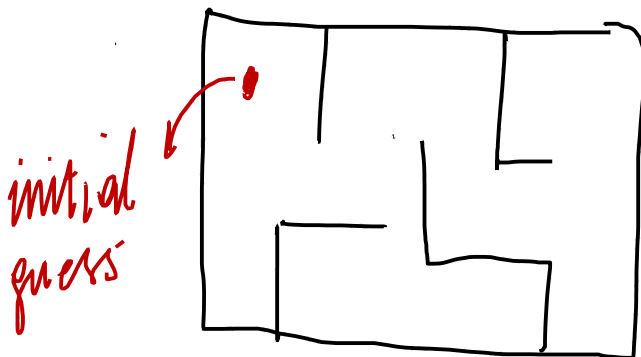
$$P(B) = \sum P(B|A) P(A)$$

normalization

BAYES FILTERS



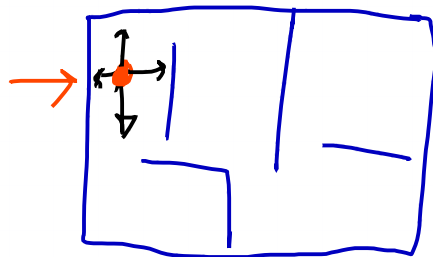
assume that we have a map



$x \rightarrow$ location

$z \rightarrow$ measurement

① sense: z_0
initial pos: x_0



② sense: z_1

$$P(x_1 | z_1) = \frac{P(z_1 | x_1) P(x_1 | z_0)}{P(z_1 | z_0)}$$

1 sensed z_1
and 1 am
at x_1

$P(z_1 | z_0)$
1 sensed $z_1 \leftarrow z_0$

1 am at x_1
and 1
sensed
 z_0 and z_1

POSTERIOR

Prior
from last
step

$$\frac{\overbrace{P(z_k | x_k)}^{\text{sense}} \overbrace{P(x_k | z_{1:k-1})}^{\text{prior from last step}}}{P(z_k | z_{1:k-1})}$$

Current Belief

z_k measured

$z_{1:k}$ and l_{am}
at location x_k

normalization

z_k, z_{k+1}, \dots, z_1

MOVEMENT

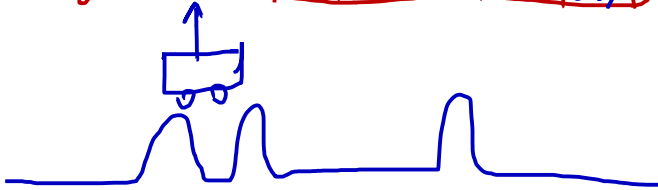
motion model

$$\underbrace{P(x_k | z_k)}_{\text{current}} = \int \underbrace{p(x_k | x_{k-1})}_{\text{motion model}} \underbrace{P(x_{k-1} | z_{1:k-1})}_{\text{from last step}} dx_{k-1}$$

EXAMPLE



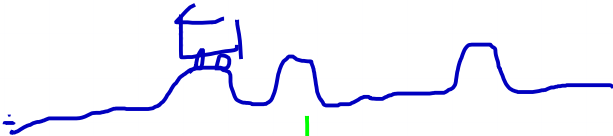
(We have a map)



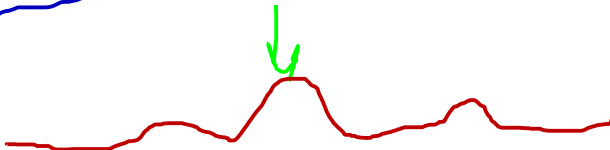
sense (green)








(move right)



(new prior)



sense (green)
posterior

0.2	0.2	0.2	0.2	0.2
				

} initial belief



↓
Assume that we sensed "green"

Our model:

$$P(z=g | x_i=f) = 0.6$$

$$P(z=b | x_i=f) = 0.4$$

$$P(z=b | x_i=b) = 0.8$$

$$P(z=r | x_i=b) = 0.2$$

→ we know these values from our previous measurements

$P(z|x)$

z_1	z_2	z_3	z_4	z_5
-------	-------	-------	-------	-------

0.2 0.2 0.2 0.2 0.2

$P(x)$

initial belief

0.2 0.4 0.6 0.2 0.2 $P(z=g|x_i)$

0.04 0.12 0.12 0.04 0.04 $\rightarrow P(z=g|x) P(x)$

$$\Rightarrow P(x_k | z_{1:k}) = \frac{P(z_k | x_k) P(x_k | z_{1:k-1})}{P(z_k | z_{1:k-1})}$$

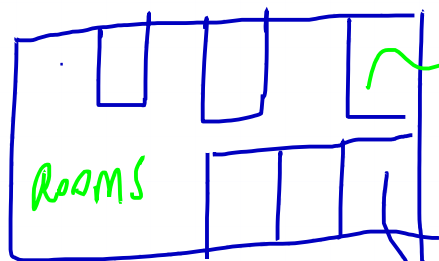
$P(z_k) = 0.36$

~~NORMALIZE~~

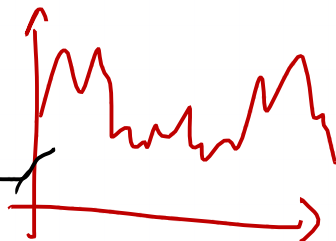
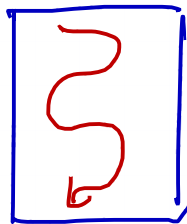
$\frac{0.04}{0.36}$	$\frac{0.12}{0.36}$	$\frac{0.12}{0.36}$	$\frac{0.04}{0.36}$	$\frac{0.04}{0.36}$
$1/9$	$1/3$	$1/3$	$1/9$	$1/9$

POSTERIOR

A REALWORLD EXAMPLE



For ⁽¹⁹⁾ each room
gather signal
strength (WIFI signal)

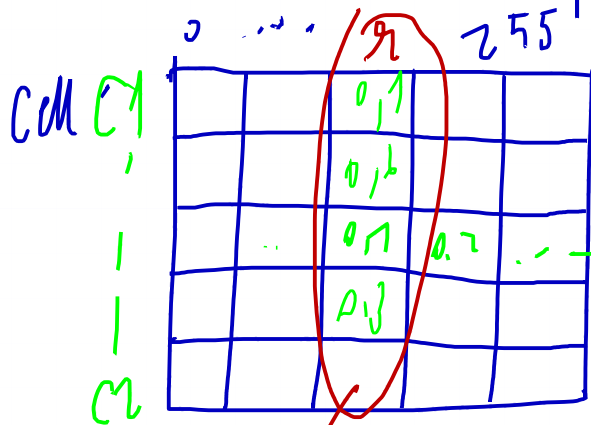


Filter and clean
data



each room
has a different
RSSI
histogram

We have a radio map for each cell.



probability of each
RSSI value
 $\sum p_i = 1$

* If we measure an RSSI value of x ; we can reason about what's the location more likely.

Probability of I am in cell i given
I got a measurement z from sensor point j

posterior = prior

c_1
c_2
\vdots
c_N

=

c_1
c_2
\vdots
c_N

\times

z	\dots	z	\dots	z
\vdots		\vdots		\vdots
$0,1$	$0,2$	\dots		

RSS

probability
for
each
cell

(NORMALIZE !!!)

Summary

- ① Get an RSSI map for each location
- ② get an initial belief
- ③ sense the environment
- ④ update your belief using the sensed value and the map.

$$\left. \begin{aligned} P(A|B) &= P(A \cap B) / P(B) \\ P(B|A) &= P(A \cap B) / P(A) \end{aligned} \right\} P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

denote $x \rightarrow$ location, $z \rightarrow$ measurement

$$P(x_i|z) = \frac{P(z|x_i) P(x_i)}{P(z)}$$

BAYES FILTER

