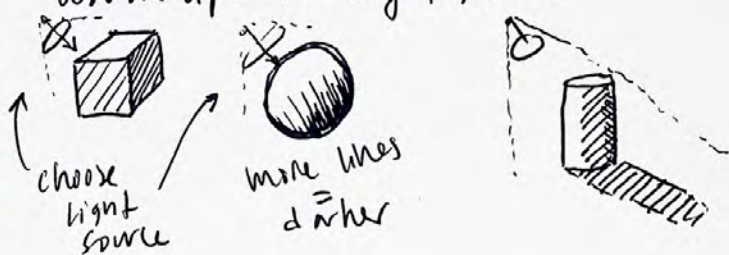
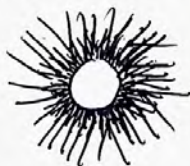


state 02

warm up: shading + blurs.



motion
blur



use many
lines
to indicate
blur or
shadow.

Last time: Bayes Theorem

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

$$p(\text{pos}|\text{sensor}) = P(\text{sensor}|\text{pos}) \cdot \frac{P(\text{pos})}{P(\text{sensor})}$$

sensor noise model
 sensor point on map
 map of sensor

$p(\text{pos}|s_1, s_2) \dots$ why not add history?

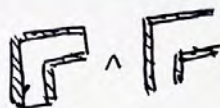
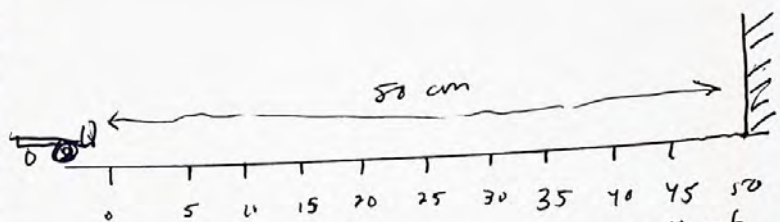
$p(\text{pos}|\text{sensors, previous})$

(2)

Intuition:



layers.

 x, y, d probability we
measured pulse
wrong is low.

$$p(\text{pos}) = 1/10$$

put in
array $N = \# \text{ of bins.}$
discretization

$$p(\text{pos}_i) = 1/10 \quad \text{belief}[N]$$

for (int $i = 0$; $i < N$; $i++$) {

$$\text{belief}[i] = 1/10;$$

}



0 1 2 3 4 5 6 7 8 9

← 50 cm →



update.

$$\uparrow p(d < 5 | s = 25 \text{ cm}) = \frac{p(s = 25 | d < 5) p(d < 5)}{p(s = 25)}$$

just
Bayes

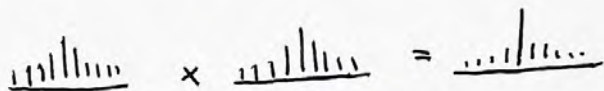
$$= \frac{0.01 \cdot 5/50}{1/50} = 0.05$$

Key insight: why not multiply? (3)

$$P(A \cap B) = P(A) \cdot P(B).$$

↑
beliefs
before

↑
beliefs
now.


$$\text{|||||} \times \text{|||||} = \text{|||||}$$

much more certain!

That's for a stationary point. But
what about motion?

change beliefs to predict.

array₁ = beliefs [N]

array₂ = predictions [N]

★ How can we make predictions?
(we measure later).

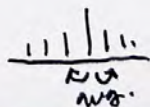
1. more correct belief: $\text{pred}[i] = \text{bel}[i-1]$



How much?

↳ guess based on output
↳ guess based on error.
...

2. blur.



④

$$\text{prediction}[i] = \text{pred}[i-1] + \text{pred}[i] + \text{pred}[i+1]$$

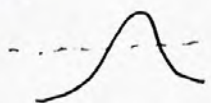
↑
very blurry.

kernel $\rightarrow \begin{bmatrix} 0.1 \\ 0.8 \\ 0.1 \end{bmatrix}$

$$\text{prediction}[i] = 0.2 \text{pred}[i-1] + 0.6 \text{pred}[i] + 0.2 \text{pred}[i+1]$$

Backward pass.

Normalize



too high
 $\sum p(i) = 1.$

$$p(i) = \frac{p(i)}{\sum p(i)}$$



normalize.

update beliefs w/ ~~sensor~~ predictions.

$$\text{bel}[i] = \text{pred}[i]$$

→ Now.

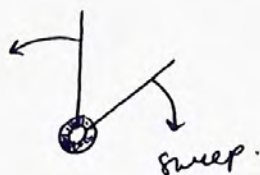
update w/ sensor:

$$\text{bel}[i] = p(\text{sensor} | i) \cdot \text{bel}[i]$$

* Error demo.

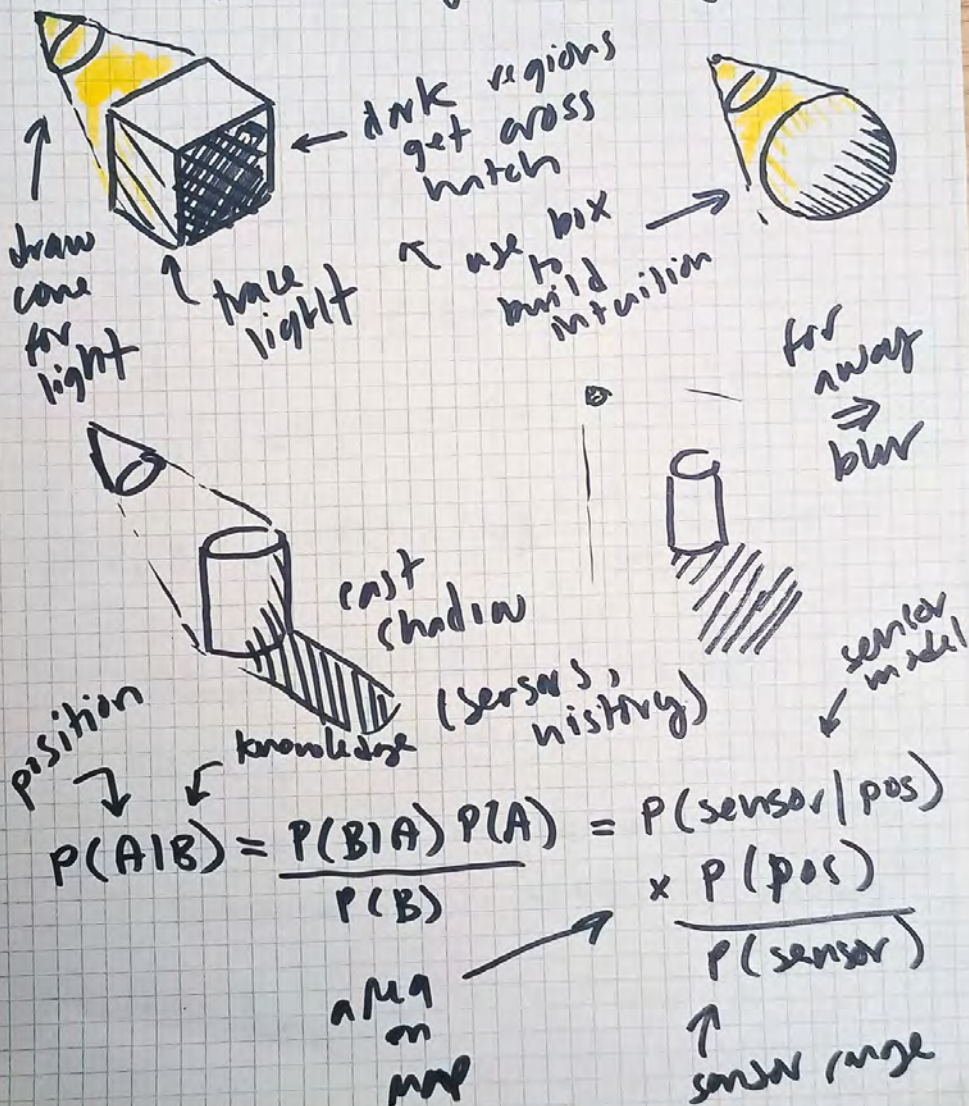
1f Hume: object detection.

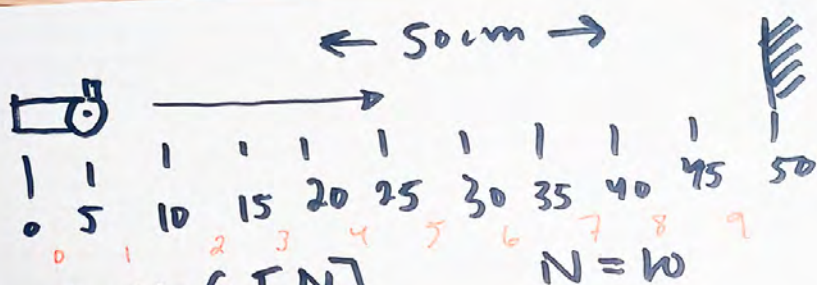
Same idea as robot but
imagine a sensor tower.



1. set up initial bugs. eg. to model
2. initialize deficits.
3. update prediction..
 - ↳ what moves?

Warm up: shading + blurring





$$p(\text{pos}) = 1/10 = \frac{1}{50}$$

for (i=0; i<N; i++)

belief[i] = 1/10;

3

$$p(d < 5 | s = 30) = \frac{p(s = 30 | d < 5) p(d < 5)}{p(s = 30)}$$

$$p(E_1 \cap E_2) = p(E_1) \cdot p(E_2) = \frac{0.01 \cdot \frac{1}{10}}{\frac{1}{50}} = 0.05$$

$$\text{beliefs} \times \text{predictions} = \text{...}$$

beliefs predictions.

How can we make a prediction?

↳ sensors? ↳ change array?

↳ speed

$$\text{pred}[i] = \text{bel}[i-1]$$

.....|.....
 ↑↑↑↑↑

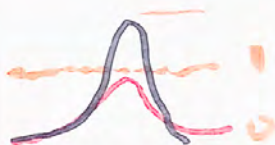
↑ amount we move

③

blur our predictions. → bc sensors are bad

$$\text{pred}[i] = 0.1 \cdot \text{pred}[i-1] + 0.8 \cdot \text{pred}[i] + 0.1 \cdot \text{pred}[i+1]$$

↑ weighted average



$$\text{bel}[i] = \text{pred}[i]$$

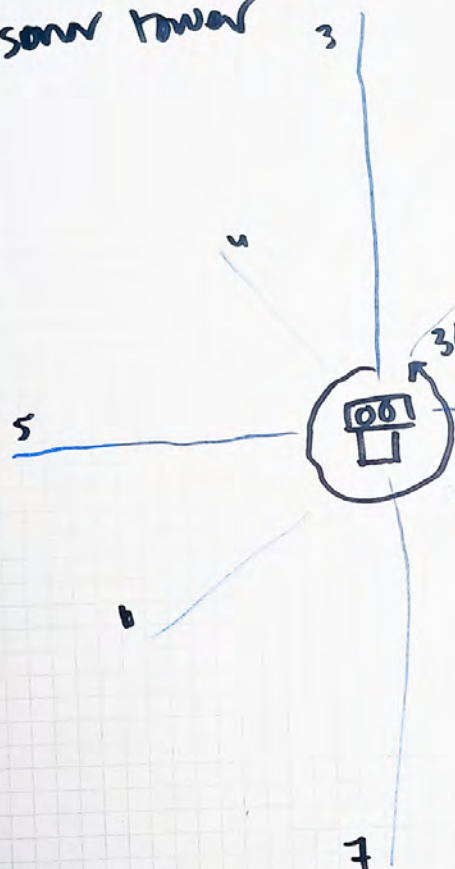
normalizing

$$\text{bel}[i] = \frac{\text{bel}[i]}{\sum \text{bel}[i]}$$

update

$$\text{bel}[i] = p(\text{sensor} | i) \cdot \text{bel}[i]$$

sonar tower



pus

(4)

$$P(A) = 1/8$$

$$P(B) = 1/360$$

A = pus of object

B = sensor dir

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

1. Set up initial bayes
2. Initial bel [i]
3. update w/ predictions