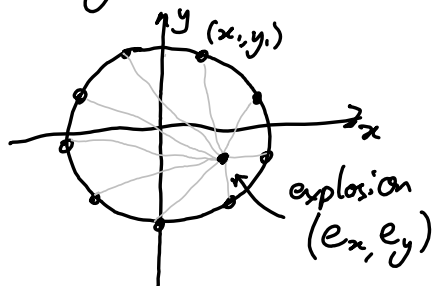


Q2. (1.22 BRML) Example 1: explosion:

This is extension of explosion example on p20.

N sensors evenly spread on Earth surface with locations $(x_i, y_i) \quad i \in \{1, 2, \dots, N\}$



Signal at each location

$$\left(\frac{1}{d_i^2 + 0.1} \right) \quad d_i \geq 0 \Rightarrow s \in (0, 1)$$

$$d_i^2 = (x_i - e_x)^2 + (y_i - e_y)^2$$

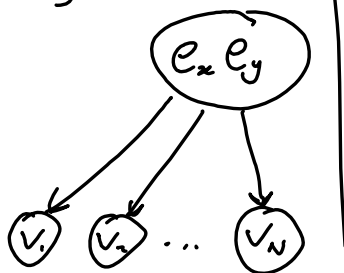
Sensors have noise $p(v_i | d_i) = \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2} \left(v_i - \frac{1}{d_i^2 + 0.1} \right)^2}$

↳ observed signal at (x_i, y_i)

Signals are independent so:

$$p(v_1, v_2, \dots, v_N, e_x, e_y) = \left[\prod_{i=1}^N p(v_i | d_i) \right] p(e_x, e_y)$$

Belief Network:



$$p(a|b) = \frac{p(a,b)}{p(b)} \propto p(a,b)$$

if $p(e_x, e_y) = c$

$$\Rightarrow p(e_x, e_y) \propto \prod_{i=1}^N p(v_i | d_i)$$

Q1.1.a)

$$P(v_1, \dots, v_N, s_1, s_2) = P(s_1)P(s_2) \prod_{i=1}^N P(v_i | d_i(1), d_i(2))$$

Note: $P(v_i | d_i(1), d_i(2)) = \frac{1}{d_i(1) + 0.1} + \frac{1}{d_i(2) + 0.1}$

and $d_i(1)$ is $\|s_1 - v_i\|$, $d_i(2) = \|s_2 - v_i\| \leftarrow L_2 \text{ norm}$

Now marginalize out s_2 :

$$P(v_1, \dots, v_N, s_1) = \sum_{s_2} P(s_1)P(s_2) \prod_{i=1}^N P(v_i | d_i(1), d_i(2))$$

$$P(s_1 | \underline{v}) = \frac{\sum_{s_2} P(s_1)P(s_2) \prod_{i=1}^N P(v_i | d_i(1), d_i(2))}{P(\underline{v})}$$

note: let priors be constant, and $P(\underline{v})$ doesn't depend on s_1 .

$$= \frac{P(s_1)P(s_2)}{P(\underline{v})} \sum_{s_2} \prod_{i=1}^N P(v_i | d_i(1), d_i(2))$$

$$P(s_1 | \underline{v}) = k \sum_{s_2} \prod_{i=1}^N P(v_i | d_i(1), d_i(2))$$

$$= k \sum_{s_2} \prod_{i=1}^N \frac{1}{\sqrt{2\pi\sigma^2}} \exp\left\{-\frac{1}{2\sigma^2} \left(v_i - \frac{1}{d_i(1)+0.1} - \frac{1}{d_i(2)+0.1}\right)^2\right\}$$

$$P(s_1 | \underline{v}) = \frac{k}{(\sqrt{2\pi\sigma^2})^N} \sum_{s_2} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^N \left[v_i - \frac{1}{d_i(1)+0.1} - \frac{1}{d_i(2)+0.1}\right]^2\right\}$$

Q2.1.b) Find $\log[p(\underline{v}|H_2)] - \log[p(\underline{v}|H_1)]$

where H_2 : There are 2 explosions, H_1 : 1 explosion

$$p(\underline{s}, \underline{v}, H_2) = k \sum_{s_2} \exp \left\{ -\frac{1}{2\sigma^2} \sum_{i=1}^N \left[v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1} \right]^2 \right\}$$

$$p(\underline{v}, s_1, s_2 | H_2) = p(s_1) p(s_2) \prod_{i=1}^N p(v_i | s_1, s_2, H_2)$$

$$p(\underline{v}, s_1 | H_1) = p(s_1) \prod_{i=1}^N p(v_i | s_1, H_1)$$

Note: $p(v_i | s_1, H_1) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{1}{2\sigma^2} \left(v_i - \frac{1}{d_i^{(1)} + 0.1} \right)^2 \right\}$

$$p(v_i | s_1, s_2, H_2) = \frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{1}{2\sigma^2} \left(v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1} \right)^2 \right\}$$

marginalise out s_2 :

$$p(\underline{v} | H_1) = \sum_{s_1} p(\underline{v}, s_1 | H_1) \quad \text{Assuming we have a constant prior}$$

$$= \sum_{s_1} p(s_1) \prod_{i=1}^N p(v_i | s_1, H_1)$$

$$= p(s_1) \sum_{s_1} \prod_{i=1}^N p(v_i | s_1, H_1)$$

$$P(\underline{v}, s_1, s_2 | H_2) = P(s_1) P(s_2) \prod_{i=1}^N P(v_i | s_1, s_2, H_2)$$

$$\begin{aligned} P(\underline{v} | H_2) &= \sum_{s_1, s_2} P(s_1) P(s_2) \prod_{i=1}^N P(v_i | s_1, s_2, H_2) \\ &= P(s_1) P(s_2) \sum_{s_1, s_2} \prod_{i=1}^N P(v_i | s_1, s_2, H_2) \end{aligned}$$

note:

$$\log(P(\underline{v} | H_2)) - \log(P(\underline{v} | H_1)) = \log \left[\frac{P(\underline{v} | H_2)}{P(\underline{v} | H_1)} \right]$$

$$\frac{P(\underline{v} | H_2)}{P(\underline{v} | H_1)} = \frac{P(s_1) P(s_2) \sum_{s_1, s_2} \exp \left\{ \sum_{i=1}^N \frac{-1}{2\sigma^2} \left(v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1} \right)^2 \right\}}{P(s_1) \sum_{s_1} \exp \left\{ \sum_{i=1}^N \frac{-1}{2\sigma^2} \left(v_i - \frac{1}{d_i^{(1)} + 0.1} \right)^2 \right\}}$$

Note: $p(s_i)$ is independent and uniform $\Rightarrow p(s_i) = \frac{1}{S}$
(where S is number of discretized points in world space)

$$\frac{P(\underline{v} | H_2)}{P(\underline{v} | H_1)} = \frac{\sum_{s_1, s_2} \exp \left\{ \sum_{i=1}^N \frac{-1}{2\sigma^2} \left(v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1} \right)^2 \right\}}{S \sum_{s_1} \exp \left\{ \sum_{i=1}^N \frac{-1}{2\sigma^2} \left(v_i - \frac{1}{d_i^{(1)} + 0.1} \right)^2 \right\}}$$

H_2 : 2 explosions:

① ZZZ, L

$$P(s_1 | \underline{v}) = \sum_{s_2} P(s_1, s_2 | \underline{v}) = \sum_{s_2} \frac{P(s_1, s_2, \underline{v})}{P(\underline{v})}$$

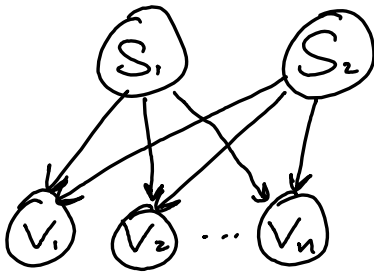
$$= \sum_{s_2} \left[\frac{P(s_1) P(s_2) \prod_{i=1}^n P(v_i | d_i^{(1)}, d_i^{(2)})}{P(\underline{v})} \right]$$

$$= \frac{P(s_1) P(s_2)}{P(\underline{v})} \sum_{s_2} P(v_i | d_i^{(1)}, d_i^{(2)})$$

$$= \frac{P(s_1) P(s_2)}{P(\underline{v})} \sum_{s_2} \prod_{i=1}^n \left[\frac{1}{\sqrt{2\pi\sigma^2}} \exp \left\{ -\frac{1}{2\sigma^2} \left[v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1} \right]^2 \right\} \right]$$

$$= \frac{P(s_1) P(s_2)}{P(\underline{v})} \left(\frac{1}{\sqrt{2\pi\sigma^2}} \right)^n \sum_{s_2} \prod_{i=1}^n \exp \left\{ -\frac{1}{2\sigma^2} \left[v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1} \right]^2 \right\}$$

$$P(s_1 | \underline{v}) = k_1 \sum_{s_2} \exp \left\{ -\frac{1}{2\sigma^2} \sum_{i=1}^n \left[v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1} \right]^2 \right\}$$



$$P(s_1, s_2, \underline{v}) = P(s_1) P(s_2) \prod_{i=1}^n P(v_i | s_1, s_2)$$

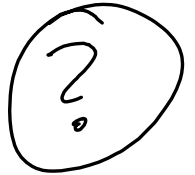
Given H_1 : 1 explosion only: Let $p(s_1) = p_1$, $p(s_2) = p_2$.

$$\begin{aligned}
 p(\underline{v}|H_1) &= \sum_{s_1} p(\underline{v}, s_1) = \sum_{s_1} p(s_1) \prod_{i=1}^n p(v_i | s_1) \\
 &= p_1 \sum_{s_1} \prod_{i=1}^n p(v_i | s_1) = p_1 \sum_{s_1} \prod_{i=1}^n \frac{1}{2\pi\sigma^2} \exp\left\{-\frac{1}{2\sigma^2} \left(v_i - \frac{1}{d_i^{(1)} + 0.1}\right)^2\right\} \\
 &= p_1 \left(\frac{1}{2\pi\sigma^2}\right)^n \sum_{s_1} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n \left(v_i - \frac{1}{d_i^{(1)} + 0.1}\right)^2\right\}
 \end{aligned}$$

under H_2 : 2 explosions:

$$p(\underline{v}|H_2) = p_1 p_2 \left(\frac{1}{2\pi\sigma^2}\right)^n \sum_{s_1, s_2} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n \left(v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1}\right)^2\right\}$$

$$\log\left[\frac{p(\underline{v}|H_2)}{p(\underline{v}|H_1)}\right] = \log\left[\frac{p_2 \sum_{s_1, s_2} \exp\left\{-\frac{1}{2\sigma^2} \sum_{i=1}^n \left(v_i - \frac{1}{d_i^{(1)} + 0.1} - \frac{1}{d_i^{(2)} + 0.1}\right)^2\right\}}{\sum_{s_1} \exp\left\{\frac{1}{2\sigma^2} \sum_{i=1}^n \left(v_i - \frac{1}{d_i^{(1)} + 0.1}\right)^2\right\}}\right]$$



Q1.22.3

if $P(H_1) = P(H_2) = k \in \mathbb{R}^+$ why does $\log P(\underline{v}|H_1) - \log P(\underline{v}|H_2)$ relate to prob of 2 explosions compared to 1?

$$P(\underline{v}|H_1) = \frac{P(\underline{v}, H_1)}{P(H_1)} \quad \text{and} \quad P(\underline{v}|H_2) = \frac{P(\underline{v}, H_2)}{P(H_2)}$$

$$\text{then } \log P(\underline{v}|H_2) - \log P(\underline{v}|H_1) = \log \left[\frac{P(\underline{v}|H_2)}{P(\underline{v}|H_1)} \right]$$

Let E be random variable for number of explosions.

$$\Rightarrow P(\underline{v}|H_1) = P(\underline{v}|E=1), \quad P(\underline{v}|H_2) = P(\underline{v}|E=2).$$

$$\begin{aligned} \Rightarrow \log \left[\frac{P(\underline{v}|H_2)}{P(\underline{v}|H_1)} \right] &= \log \left[\frac{P(H_1) P(\underline{v}, E=1)}{P(H_2) P(\underline{v}, E=2)} \right] = \log \left[\frac{k P(\underline{v}, E=1)}{k P(\underline{v}, E=2)} \right] \\ &= \log \left(\frac{P(\underline{v}, E=1)}{P(\underline{v}, E=2)} \right) = \log \left(\frac{P(E=1|\underline{v}) P(\underline{v})}{P(E=2|\underline{v}) P(\underline{v})} \right) = \log \left(\frac{P(E=1|\underline{v})}{P(E=2|\underline{v})} \right) \end{aligned}$$



1.22.4/

Assume k explosions, what is computational complexity of $P(\underline{V} | H_k)$?

• Using some methods as before we get:

$$P(\underline{V} | H_k) = \prod_{i=1}^k [P(S_i)] \left(\frac{1}{2\pi\sigma^2} \right)^n \sum_{S_1, \dots, S_k} \exp \left\{ -\frac{1}{2\sigma^2} \sum_{i=1}^n \left[V_i - \sum_{j=1}^k \frac{1}{d_i(j) + 0.1} \right]^2 \right\}$$

Therefore we have to marginalise over all k possible explosion locations. This means calculations grow $O(S^k)$ where S is the number of discretized world points we are iterating over for each S_i .