

ASSUME

\* LOCATION OF LANDMARK AT TIME  $t$ ,  $l_t$ , IS KNOWN

\* OBSERVE  $Z_t$  WHICH IS THE DELTA FROM

\* OBSERVATION MODEL:  $P(Z_t | X_t) \propto \exp\left(-\frac{\|Z_t - (l_t - X_t)\|^2}{\sum_t^{obs}}\right)$

\* PRIOR AT  $t=0$ :  $P(X_0) \propto \exp\left(-\frac{\|X_0 - \mu_0\|^2}{\Sigma_0}\right)$



T=0

GIVEN OBSERVATION  $Z_0$ , WHAT IS THE MOST LIKELY STATE  $X_0$ ?

$$\operatorname{argmax}_{X_0} \log P(X_0 | Z_0)$$

$$= \operatorname{argmax}_{X_0} \log \frac{P(Z_0 | X_0) P(X_0)}{P(Z_0)}$$

[  $P(Z_0)$  DOES NOT DEPEND ON  $X_0$  ]

$$= \operatorname{argmax}_{X_0} \log P(Z_0 | X_0) + \log P(X_0)$$

SUBSTITUTE PRIOR & OBSERVATION MODEL

$$= \operatorname{argmax}_{x_0} \log \left[ \frac{1}{\eta} \exp \left( - \frac{\|x_0 - \mu_0\|^2}{\Sigma_0} \right) \right] +$$

$$\log \left[ \frac{1}{\eta} \exp \left( - \frac{\|z_0 - (\ell_0 - x_0)\|^2}{\Sigma_0^{\text{obs}}} \right) \right]$$

$$= \operatorname{argmin}_{x_0} \left[ \frac{\|x_0 - \mu_0\|^2}{\Sigma_0} + \frac{\|z_0 - (\ell_0 - x_0)\|^2}{\Sigma_0^{\text{obs}}} \right]$$

TAKE GRAD  $\frac{\partial}{\partial x_0} (\cdot) = 0$  AND SET TO ZERO

$$\frac{\partial}{\partial x_0} \left( \frac{\|x_0 - \mu_0\|^2}{\Sigma_0} + \frac{\|z_0 - (\ell_0 - x_0)\|^2}{\Sigma_0^{\text{obs}}} \right) = 0$$

$$x_0 \left( \frac{1}{\Sigma_0} + \frac{1}{\Sigma_0^{\text{obs}}} \right) = \frac{\mu_0}{\Sigma_0} + \frac{(\ell_0 - z_0)}{\Sigma_0^{\text{obs}}}$$

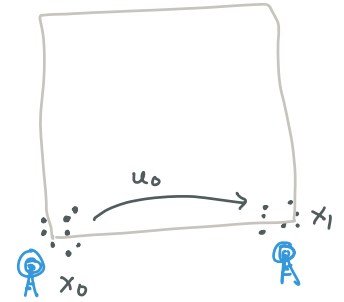
$$x_0 = \frac{\frac{\mu_0}{\Sigma_0} + \frac{(\ell_0 - z_0)}{\Sigma_0^{\text{obs}}}}{\left( \frac{1}{\Sigma_0} + \frac{1}{\Sigma_0^{\text{obs}}} \right)}$$

CONTRIBUTION BY PRIOR CONTRIBUTION BY OBSERVATION

GIVEN OBSERVATION  $Z_0, Z_1$  AND ACTION  $u_0$  FIND THE MOST LIKELY  $X_0, X_1, \dots$

ASSUME DYNAMICS MODEL:  $P(X_1 | X_0, u_0) \propto \exp\left(-\frac{\|X_1 - (X_0 + u_0)\|^2}{\sum_0 a_{gn}}\right)$

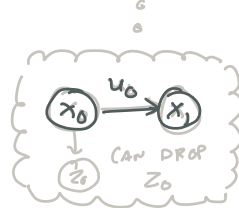
arg max  $P(X_1, X_0 | Z_1, u_0, Z_0)$   
 $X_0, X_1$   
 (APPLY BAYES RULE)



arg max  $P(Z_1 | X_1, X_0, u_0, Z_0) P(X_1, X_0 | u_0, Z_0)$   
 $X_0, X_1$   
 $P(Z_1) =$  DOES NOT DEPEND ON  $X_0, X_1$

arg max  $P(Z_1 | X_1) P(X_1, X_0 | u_0, Z_0)$   
 $X_0, X_1$   
 $P(A, B) = P(A|B)P(B)$

arg max  $P(Z_1 | X_1) P(X_1 | X_0, u_0, Z_0) P(X_0 | u_0, Z_0)$   
 $X_0, X_1$



arg max  $P(Z_1 | X_1) P(X_1 | X_0, u_0) P(X_0 | Z_0)$   
 $X_0, X_1$   
 BAYES THEOREM

$$\arg \max_{x_0, x_1} P(z_1 | x_1) P(x_1 | x_0, u_0) P(z_0 | x_0) P(x_0)$$

$$\arg \max_{x_0, x_1} \log P(z_1 | x_1) + \log P(x_1 | x_0, u_0) + \log P(z_0 | x_0) + \log P(x_0)$$

$$\arg \max_{x_0, x_1} \log \left( \frac{1}{\eta} \exp \left( - \frac{\|z_1 - (l_1 - x_1)\|^2}{\sum_1^{obs}} \right) \right) + \log \left( \frac{1}{\eta} \exp \left( - \frac{\|x_1 - (x_0 + u_0)\|^2}{\sum_1} \right) \right) \\ + \log \left( \frac{1}{\eta} \exp \left( - \frac{\|z_0 - (l_0 - x_0)\|^2}{\sum_0^{obs}} \right) \right) + \log \left( \frac{1}{\eta} \exp \left( - \frac{\|x_0 - \mu_0\|^2}{\sum_0} \right) \right)$$

$$\arg \min_{x_0, x_1} \frac{\|z_1 - (l_1 - x_1)\|^2}{\sum_1^{obs}} + \frac{\|x_1 - (x_0 + u_0)\|^2}{\sum_1} + \frac{\|z_0 - (l_0 - x_0)\|^2}{\sum_0^{obs}} + \frac{\|x_0 - \mu_0\|^2}{\sum_0}$$

$$\nabla_{x_0} (\cdot) = 0 \Rightarrow - \frac{x_1 - (x_0 + u_0)}{\sum_1} + \frac{z_0 - (l_0 - x_0)}{\sum_0^{obs}} + \frac{x_0 - \mu_0}{\sum_0} = 0$$

$$\nabla_{x_1} (\cdot) = 0 \Rightarrow \frac{z_1 - (l_1 - x_1)}{\sum_1^{obs}} + \frac{x_1 - (x_0 + u_0)}{\sum_1} = 0$$

AT THIS POINT, IT IS HARD TO ANALYTICALLY MAKE PROGRESS  
 AND WE SOLVE NUMERICALLY

$$[A] \begin{bmatrix} x_0 \\ x_1 \end{bmatrix} = [B]$$

$\theta$

$$\theta = A^{-1} b$$

$x_0$ :	BALANCE	PRIOR ,	CBS ,	MOTION
		$\frac{\mu_0}{\Sigma}$	$\frac{(l_0 - z_0)}{\Sigma^{obs}}$	$\frac{(x_1 - u_0)}{\Sigma^{dyn}}$

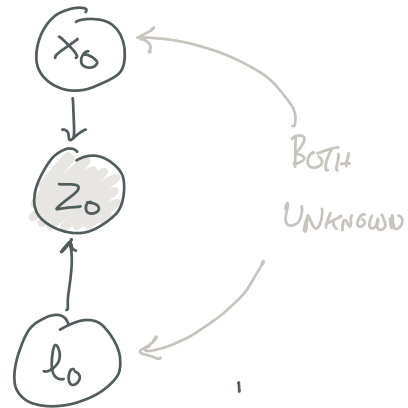
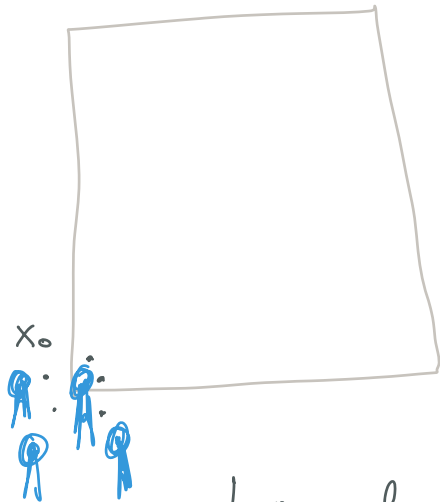
IN GENERAL

$$P(x_0, x_1, \dots \mid z_0, u_0, z_1, \dots)$$

IS A LEAST SQUARES OPTIMIZATION

Now WHAT IF LANDMARKS ARE UNKNOWN

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LET  $l_0$  BE LANDMARK WHICH IS UNKNOWN.

$$\arg \max_{x_0, l_0} P(x_0, l_0 | z_0)$$

$$\arg \max_{x_0, l_0} \frac{P(z_0 | x_0, l_0) P(x_0, l_0)}{P(z_0)}$$

$$\arg \max_{x_0, l_0} \log P(z_0 | x_0, l_0) + \log P(x_0) + \log P(l_0)$$

// SAME OBS MODEL. = PRIOR OVER  $x_0$   
= PRIOR OVER LANDMARK

$$\arg \min_{x_0, l_0} \frac{\|z_0 - (x_0 - l_0)\|^2}{\sum_0^{\text{obs}}} + \frac{(x_0 - t_0)^2}{\sum_0} + \frac{(l_0 - k_0^t)^2}{\sum_0}$$

$$\frac{\partial}{\partial} (\cdot) = 0 \Rightarrow x_0 = \dots$$

$$l_0 = \dots$$

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LAND MARKS BECOME EXTRA VARIABLES YOU SOLVE FOR!