3. (35) Consider the model, described on the lecture:

$$x_t = x_{t-1} + u_{t-1} + \epsilon_{t-1}, \quad z_t = x_t + \eta_t$$
 (4)

where  $z_t$  is observed state,  $x_t$  is hidden state and, for simplicity,  $u_t \equiv u$ . Assume that the measurement and the control noise have the following distributions:

$$P(\epsilon_t) = \mathcal{N}(0, \sigma), \quad P(\eta_t) = \mathcal{N}(0, \rho = 2\sigma).$$
 (5)

Compute, analytically, the following distributions:

(a) (5)  $P(x_0|z_0)$ : distribution of  $x_0$  after the first 1 observation

$$P(Z_{o}|X_{o}) = \frac{1}{p\sqrt{2\pi}} exp\left(-\frac{(x_{o}-z_{o})^{2}}{2p^{2}}\right)$$

$$P(X_{o}|Z_{o}) = \frac{P(Z_{o}|X_{o})P(X_{o})}{P(Z_{o})} = P(Z_{o}|X_{o})$$

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$$= \frac{1}{p\sqrt{2\pi}} exp\left(-\frac{(x_{o}-z_{o})^{2}}{2p^{2}}\right)$$

(b) (10)  $P(x_0|z_0,z_1)$ : distribution of  $x_0$  after the first 2 observations

$$Z_1 = X_0 + U_0 + \epsilon_0 + \ell_1 \sim \mathcal{N}(X_0 + U_0, \sigma^2 + p^2)$$

$$P\left(x_{0} \mid Z_{0} \mid Z_{1}\right) = \frac{P(z_{1} \mid x_{0}) P(z_{0} \mid x_{0}) P(x_{0})}{P(z_{1} \mid z_{0})}$$

Если забит о константах и априорной

вероятност Р(хо), то:

$$P\left(\left(\frac{1}{2}\right) - \left(\frac{1}{2}\right) \sim exp\left(-\frac{\left(\frac{1}{2}\right) - \left(\frac{1}{2}\right)^{2}}{2\left(\sigma^{2} + p^{2}\right)}\right) exp\left(-\frac{\left(\frac{1}{2}\right) - \left(\frac{1}{2}\right)^{2}}{2\left(\sigma^{2}\right)^{2}}\right)$$

$$= exp \left[ -\frac{1}{2} \left( \frac{1}{\sigma_0^2} + \frac{1}{\sigma_1^2} \right) \left( x_0^2 - 2x_0 \left( \frac{\sigma_*^2}{\sigma_1^2} (z_1 - u_0) + \frac{\sigma_*^2}{\sigma_0^2} z_0 \right) + \frac{\sigma_*^2}{\sigma_0^2} z_0 \right) + \frac{\sigma_*^2}{\sigma_0^2} z_0 \right]$$

$$\chi_{o} \mid Z_{o}, Z_{1} \sim \mathcal{N} \left( \frac{\sigma_{*}^{2}}{\sigma_{1}^{2}} (z_{1} - u_{o}) + \frac{\sigma_{*}^{2}}{\sigma_{o}^{2}} z_{o} \right)$$

$$ge = (\sigma_0^{-2} + \sigma_1^{-2})^{-1}, \quad \sigma_0^2 = p^2, \quad \sigma_1^2 = p^2 + \sigma^2$$

(c) (20)  $P(x_1|z_0, z_2)$ : distribution of  $x_1$  after the first 3 observations, when the second observation is missing.

Note that each of these distributions is Gaussian, so its enough to specify mean and variance in each case. Feel free to use Eq. (3) (Mathematica or sympy may also be useful for algebraic manipulations.)

$$Z_{o} = X_{o} + h_{o} = (X_{1} - u_{o} - \epsilon_{o}) + h_{o} \sim N(x_{1} - u_{o}, \sigma^{2} + p^{2})$$

$$Z_{2} = X_{2} + h_{2} = (x_{1} + u_{1} + \epsilon_{1}) + h_{2} \sim N(x_{1} - u_{1}, \sigma^{2} + p^{2})$$
Ananom and upon peg n gy use my my nx ry,
$$\mathcal{E}_{cm} \quad \mathcal{B}_{b} = \mathcal{G}_{1}^{2} = \mathcal{G}_{1}^{2} + p^{2}$$

$$\mathcal{G}_{0}^{2} = \mathcal{G}_{1}^{2} = \mathcal{G}_{1}^{2} + p^{2}$$

$$\mathcal{G}_{1}^{2} = (\mathcal{G}_{0}^{-2} + \mathcal{G}_{2}^{-2})^{-1} = \frac{1}{2} (\mathcal{G}_{1}^{2} + p^{2})$$
To other M by get
$$X_{1} | Z_{2} Z_{0} \sim N\left(\frac{\mathcal{G}_{2}^{2}}{\mathcal{G}_{0}^{2}}(Z_{0} + u) + \frac{\mathcal{G}_{2}^{2}}{\mathcal{G}_{2}^{2}}(Z_{2} - u), \mathcal{G}_{2}^{2}\right)$$

$$\int \left(\frac{Z_{0} + Z_{1}}{2}, \mathcal{G}_{2}^{2}\right)$$