# Bayesian ML Project

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## 1 State space formulation

Equations from the paper

$$y_t = \mu_t + \gamma_t + z_t^a o_t + (1 - z_t^a) \varepsilon_t$$

$$\mu_t = \mu_{t-1} + \delta_{t-1} + z_t^c r_t + (1 - z_t^c) u_t$$

$$\delta_t = \delta_{t-1} + v_t$$

$$\gamma_t = -\sum_{s=1}^{S} \gamma_{t-s} + w_t$$

State space equations

$$y_t = Z_t \alpha_t + A_t o_t + (1 - a_t) \varepsilon_t$$

$$\alpha_{t+1} = T_t \alpha_t + R_t C_t \eta_t + R_t (1 - c_t) \xi_t$$

$$o_t \sim \mathcal{N}(0, H_t^o)$$

$$\varepsilon_t \sim \mathcal{N}(0, H_t^{\varepsilon})$$

$$\eta_t \sim \mathcal{N}(0, Q_t^{\eta})$$

$$\xi_t \sim \mathcal{N}(0, Q_t^{\xi})$$

Where

$$\alpha_t = (\mu_t, \delta_t, \gamma_t, \gamma_{t-1}, \dots, \gamma_{t-S+2})^T$$

$$A_t = z_t^a \sim \text{Bernoulli}(p_a)$$

$$C_t = z_t^c \sim \text{Bernoulli}(p_c)$$

$$\eta_t = (r_t, v_t, w_t)$$

$$\xi_t = (u_t, v_t, w_t)$$

$$Z_t = (1, 0, 1, 0, \dots, 0)$$

$$T_t = \operatorname{diag}(T_\mu, T_\gamma)$$

$$T_{\mu} = \begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}$$

$$T_{\gamma} = \begin{pmatrix} -1 & -1 & \dots & -1 & -1 \\ 1 & 0 & \dots & 0 & 0 \\ 0 & 1 & \dots & 0 & 0 \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & 0 \end{pmatrix}$$

$$R = \operatorname{diag}(R_{\mu}, R_{\gamma})$$

$$R_{\mu} = I_2$$

$$R_{\gamma} = (1, 0, \dots, 0)^T$$

$$H_t^o = \sigma_o^2$$

$$H^\varepsilon=\sigma_\varepsilon^2$$

$$Q^{\eta} = \operatorname{diag}(\sigma_r^2, \sigma_v^2, \sigma_w^2)$$

$$Q^{\xi} = \operatorname{diag}(\sigma_u^2, \sigma_v^2, \sigma_w^2)$$

### 2 Kalman Filter

$$a_{t|t} = \mathbb{E}(\alpha_t|Y_t)$$

$$a_{t+1} = \mathbb{E}(\alpha_{t+1}|Y_t)$$

$$P_{t|t} = \operatorname{Var}(\alpha_{t+1}|Y_t)$$

$$P_{t+1} = \operatorname{Var}(\alpha_{t+1}|Y_t)$$

$$v_t = y_t - \mathbb{E}(y_t|Y_{t-1}) = y_t - \mathbb{E}(Z_t\alpha_t + A_to_t + (1 - A_t)\varepsilon_t|Y_{t-1}) = y_t - Z_ta_t$$

$$a_{t|t} = \mathbb{E}(\alpha_t|Y_t) = \mathbb{E}(\alpha_t|Y_{t-1}, v_t)$$

$$a_{t+1} = \mathbb{E}(\alpha_{t+1}|Y_t) = \mathbb{E}(\alpha_{t+1}|Y_{t-1}, v_t)$$

Lemma 1

If 
$$(x, y) = \mathcal{N}((\mu_x, \mu_y), \begin{pmatrix} \Sigma_{xx} & \Sigma_{xy} \\ \Sigma_{xy}^T & \Sigma_{yy} \end{pmatrix})$$

Then 
$$p(x|y) = \mathcal{N}(\mathbb{E}(x|y), \operatorname{Var}(x|y)),$$

$$\mathbb{E}(x|y) = \mu_x + \Sigma_{xy} \Sigma_{yy}^{-1} (y - \mu_y)$$

$$Var(x|y) = \Sigma_{xx} - \Sigma_{xy} \Sigma_{yy}^{-1} \Sigma_{xy}^{T}$$

Apply Lemma 1 for 
$$x = \alpha_t | Y_{t-1}$$
 and  $y = v_t | Y_{t-1}$ 

Then 
$$a_{t|t} = \mathbb{E}(\alpha_t | T_{t-1}) + \text{Cov}(\alpha_t, v_t) [\text{Var}(v_t)]^{-1} v_t = a_t + P_t Z_t^T F_t^{-1} v$$

$$Cov(\alpha_t, v_t) = \mathbb{E}[\alpha_t(Z_t\alpha_t + A_to_t + (1 - A_t)\varepsilon_t - Z_ta_t)^T | Y_{t-1}] = \mathbb{E}[\alpha_t(\alpha_t - a_t)^T Z_t^T | Y_{t-1}] = P_t Z_t^T$$

$$P_{t} = \mathbb{E}[\alpha_{t}(\alpha_{t} - a_{t})^{T}|Y_{t-1}] = \mathbb{E}[(\alpha_{t} - a_{t})(\alpha_{t} - a_{t})^{T}|Y_{t-1}] + a_{t}\mathbb{E}[(\alpha_{t} - a_{t})^{T}] = \text{Var}(\alpha_{t}|Y_{t-1})$$

$$\begin{split} F_t &= \mathrm{Var}(v_t|Y_{t-1}) = \mathrm{Var}(Z_t\alpha_t + A_to_t + (1-A_t)\varepsilon_t - Z_ta_t|Y_{t-1}) \\ F_t &= \mathbb{E}(Z_t(\alpha_t - a_t)(\alpha_t - a_t)^T Z_t^T | Y_{t-1}) + \mathbb{E}[(A_to_t + (1-A_t)\varepsilon_t)(A_to_t + (1-A_t)\varepsilon_t)^T] \\ &= Z_t P_t Z_t^T + p_t^a H_t^o + [1-p_t^a] H_t^\varepsilon \\ a_{t|t} &= a_t + P_t Z^t F_t^{-1} v_t \\ P_{t|t} &= \mathrm{Var}(\alpha_t|Y_t) = \mathrm{Var}(\alpha_t|Y_{t-1}, v_t) = \mathrm{Var}(\alpha_t|Y_{t-1}) - \mathrm{Cov}(\alpha_t, v_t) [\mathrm{Var}(v_t)]^{-1} \mathrm{Cov}(\alpha_t, v_t)^T \\ P_t &= P_t Z_t^T F_t^{-1} Z_t P_t \\ a_{t+1} &= \mathbb{E}(T_t\alpha_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathbb{E}(\alpha_t|Y_t) = T_t a_{t|t} = T_t a_t + K_t v_t \\ K_t &= T_t P_t Z^t F_t^{-1} \\ P_{t+1} &= \mathrm{Var}(T_t\alpha_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t [p_t^c Q_t^\eta + T_t^2] \\ &= T_t \mathrm{Var}(T_t \alpha_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t [p_t^c Q_t^\eta + T_t^2] \\ &= T_t \mathrm{Var}(T_t \alpha_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t [p_t^c Q_t^\eta + T_t^2] \\ &= T_t \mathrm{Var}(T_t \alpha_t + R_t C_t \eta_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t [p_t^c Q_t^\eta + T_t^2] \\ &= T_t \mathrm{Var}(T_t \alpha_t + R_t C_t \eta_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t [p_t^c Q_t^\eta + T_t^2] \\ &= T_t \mathrm{Var}(T_t \alpha_t + R_t C_t \eta_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t [p_t^c Q_t^\eta + T_t^2] \\ &= T_t \mathrm{Var}(T_t \alpha_t + R_t C_t \eta_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t [p_t^c Q_t^\eta + T_t^2] \\ &= T_t \mathrm{Var}(T_t \alpha_t + R_t C_t \eta_t + R_t C_t \eta_t + R_t (1-C_t) \xi_t | Y_t) = T_t \mathrm{Var}(\alpha_t|Y_t) T_t^T + R_t \mathrm{Var}(\alpha_t|Y_t$$

#### KALMAN FILTER RECURSION

 $(1 - p_t^c)Q_t^{\xi}]R_t^T$ 

$$\begin{aligned} v_t &= y_t - Z_t a_t \\ a_{t|t} &= a_t + P_t Z^t F_t^{-1} v_t \\ a_{t+1} &= T_t a_t + K_t v_t \\ F_t &= Z_t P_t Z_t^T + p_t^a H_t^o + [1 - p_t^a] H_t^\varepsilon \\ P_{t|t} &= P_t - P_t Z_t^T F_t^{-1} Z_t P_t \\ P_{t+1} &= T_t \text{Var}(\alpha_t | Y_t) T_t^T + R_t [p_t^c Q_t^\eta + (1 - p_t^c) Q_t^\xi] R_t^T \end{aligned}$$