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VI-CVA

VI-CVA here denotes the CVA model trained by VI, and the whole training process is mainly established according to the reference:

Yu J, Ye L, Zhou L, et al. Dynamic process monitoring based on variational Bayesian canonical variate analysis[J]. IEEE Transactions on Systems, Man, and Cybernetics: Systems, 2021, 52(4): 2412-2422.

Wishart distribution ${\mathcal W}$ is used as the prior of the precision in Multivariate Gaussian.

Model Structure

input:

$$\mathbf{x}_t \in \mathbb{R}^{M imes 1}$$

$$\mathbf{P}^t = [\mathbf{x}_{t-1}^T, \mathbf{x}_{t-2}^T, ..., \mathbf{x}_{t-l}^T]^T \in \mathbb{R}^{Ml \times 1}$$

$$\mathbf{F}^t = [\mathbf{x}_t^T, \mathbf{x}_{t+1}^T, ..., \mathbf{x}_{t+s}^T]^T \in \mathbb{R}^{M(s+1) \times 1}$$

$$\mathbf{P}^t = [p_1^t, p_2^t, ..., p_M^t]^T \in \mathbb{R}^{M imes 1}$$

$$\mathbf{F}^t = [f_1^t, f_2^t, ..., f_{M(s+1)}^t]^T \in \mathbb{R}^{M(s+1) imes 1}$$

model mapping:

$$\mathbf{P} = \mathbf{W}^T \mathbf{z} + \epsilon$$

$$\mathbf{F} = \mathbf{H}^T \mathbf{z} + \delta$$

$$\mathbf{z}_t \in \mathbb{R}^{D \times 1}$$

$$\mathbf{W}^T \in \mathbb{R}^{Ml \times D}$$

$$\mathbf{H}^T \in \mathbb{R}^{M(s+1) imes D}$$

$$\mathbf{W}^T = egin{pmatrix} -\mathbf{W}_1^T - \ -\mathbf{W}_2^T - \ dots \ -\mathbf{W}_M^T - \end{pmatrix} = egin{pmatrix} (\mathbf{W}_1, \mathbf{W}_2, \dots, \mathbf{W}_M) \ , \mathbf{W}_m \in \mathbb{R}^{D imes 1} \end{pmatrix}$$

$$\mathbf{H}^T = egin{pmatrix} -\mathbf{H}_1^T - \ -\mathbf{H}_2^T - \ dots \ -\mathbf{H}_{M(s+1)}^T - \end{pmatrix} = egin{pmatrix} \mathbf{H}_1, \mathbf{H}_2, \dots, \mathbf{H}_{M(s+1)} \end{pmatrix}, \mathbf{H}_m \in \mathbb{R}^{D imes 1}$$

prior distributions:

$$\mathcal{P}(\mathbf{z}) = \prod_{t=1}^{N} \mathcal{N}(\mathbf{z}_t|0,\mathbf{I})$$

$$\mathcal{P}(\epsilon) = \prod_{m=1}^{Ml} \mathcal{N}(\epsilon_m|0, au_m^{-1}); \mathcal{P}(\delta) = \prod_{m=1}^{M(s+1)} \mathcal{N}(\delta_m|0,\psi_m^{-1})$$

$$\mathcal{P}(au) = \prod_{m=1}^{Ml} \mathcal{G}(au_m|j_m^ au, k_m^ au); \mathcal{P}(\psi) = \prod_{m=1}^{M(s+1)} \mathcal{G}(\psi_m|j_m^\psi, k_m^\psi)$$

$$\mathcal{P}(lpha) = \prod_{m=1}^{Ml} \mathcal{W}(lpha_m | a_m^lpha, \mathbf{B}_m^lpha); \mathcal{P}(eta) = \prod_{m=1}^{M(s+1)} \mathcal{W}(eta_m | a_m^eta, \mathbf{B}_m^eta)$$

$$\mathcal{P}(\mathbf{W}|lpha) = \prod_{m=1}^{Ml} \mathcal{N}(\mathbf{W}_m|0,lpha_m^{-1}); \mathcal{P}(\mathbf{H}|eta) = \prod_{m=1}^{M(s+1)} \mathcal{N}(\mathbf{H}_m|0,eta_m^{-1})$$

$$\mathcal{P}(\mathbf{P}|\mathbf{z},\mathbf{W},lpha, au) = \prod_{t=1}^{N}\prod_{m=1}^{Ml}\mathcal{N}(p_{m}^{t}|\mathbf{W}_{m}^{T}\mathbf{z}_{t}, au_{m}^{-1})$$

$$\mathcal{P}(\mathbf{F}|\mathbf{z},\mathbf{H},eta,\psi) = \prod_{t=1}^{N} \prod_{m=1}^{M(s+1)} \mathcal{N}(f_m^t|\mathbf{H}_m^T\mathbf{z}_t,\psi_m^{-1})$$

variational distributions:

$$\mathcal{Q}(\mathbf{z}) = \prod_{t=1}^{N} \mathcal{N}(\mathbf{z}_t | \mu^z_t, (\Lambda^z_t)^{-1})$$

$$\mathcal{Q}(\mathbf{W}) = \prod_{m=1}^{Ml} \mathcal{N}(\mathbf{W}_m | \mu_m^W, (\Lambda_m^W)^{-1}); \, \mathcal{Q}(\mathbf{H}) = \prod_{m=1}^{M(s+1)} \mathcal{N}(\mathbf{H}_m | \mu_m^H, (\Lambda_m^H)^{-1})$$

$$\mathcal{Q}(au) = \prod_{m=1}^{Ml} \mathcal{G}(au_m | \lambda_m^ au,
u_m^ au); \, \mathcal{Q}(\psi) = \prod_{m=1}^{M(s+1)} \mathcal{G}(\psi_m | \lambda_m^\psi,
u_m^\psi)$$

$$\mathcal{Q}(lpha) = \prod_{m=1}^{Ml} \mathcal{W}(lpha_m |
u_m^lpha, \mathbf{V}_m^lpha); \, \mathcal{Q}(eta) = \prod_{m=1}^{M(s+1)} \mathcal{W}(eta_m |
u_m^eta, \mathbf{V}_m^eta)$$

Upate Strategy

joint distribution

$$\mathcal{P}(\mathbf{P}, \mathbf{F}, \mathbf{H}, \mathbf{W}, \mathbf{z}, \tau, \psi, \alpha, \beta)$$

$$= \mathcal{P}(\mathbf{P}|\mathbf{W}, \mathbf{z}, \tau)\mathcal{P}(\mathbf{W}|\alpha)\mathcal{P}(\alpha)\mathcal{P}(\tau)$$

$$\mathcal{P}(\mathbf{F}|\mathbf{H}, \mathbf{z}, \psi)\mathcal{P}(\mathbf{H}|\beta)\mathcal{P}(\beta)\mathcal{P}(\psi)\mathcal{P}(\mathbf{z})$$

mean-field distribution

$$\mathcal{Q}(\mathbf{H}, \mathbf{W}, \mathbf{z}, \tau, \psi, \alpha, \beta)$$

$$= \mathcal{Q}(\mathbf{W}) \mathcal{Q}(\alpha) \mathcal{Q}(\tau) \mathcal{Q}(\mathbf{H}) \mathcal{Q}(\beta) \mathcal{Q}(\psi) \mathcal{Q}(\mathbf{z})$$

Update $\mathcal{Q}(\mathbf{z}_t)$

$$egin{aligned} \Lambda_t^z &= \sum_{m=1}^{Ml} \mathbb{E}_{Q(au_m)}(au_m) \mathbb{E}_{Q(\mathbf{W}_m)}(\mathbf{W}_m \mathbf{W}_m^T) + \ \sum_{m=1}^{M(s+1)} \mathbb{E}_{Q(\psi_m)}(\psi_m) \mathbb{E}_{Q(\mathbf{H}_m)}(\mathbf{H}_m \mathbf{H}_m^T) + \mathbf{I} \end{aligned}$$

$$\begin{array}{l} \mu_t^z = (\Lambda_t^z)^{-1} \{ \sum_{m=1}^{Ml} \mathbb{E}_{Q(\tau_m)}(\tau_m) \mathbb{E}_{Q(\mathbf{W}_m)}(\mathbf{W}_m) p_m^t + \\ \sum_{m=1}^{M(s+1)} \mathbb{E}_{Q(\psi_m)}(\psi_m) \mathbb{E}_{Q(\mathbf{H}_m)}(\mathbf{H}_m) f_m^t \} \end{array}$$

$$\mathbb{E}_{Q(au_m)}(au_m) = rac{\lambda_m^ au}{
u_m^ au}$$

$$\mathbb{E}_{Q(\mathbf{W}_m)}(\mathbf{W}_m\mathbf{W}_m^T) = (\Lambda_m^W)^{-1} + \mu_m\mu_m^T$$

update $Q(\mathbf{H})$, $Q(\mathbf{W})$

$$\Lambda_m^W = \sum_{t=1}^N \mathbb{E}_{Q(au_m)}(au_m) \mathbb{E}_{Q(\mathbf{z}_t)}(\mathbf{z}_t \mathbf{z}_t^T) + \mathbb{E}_{Q(lpha_m)}(lpha_m)$$

$$egin{aligned} \mu_m^{\mathbf{W}} &= (\Lambda_m^W)^{-1} \mathbb{E}_{Q(au_m)}(au_m) \sum_{t=1}^N \mathbb{E}_{Q(\mathbf{z}_t)}(\mathbf{z}_t) p_m^t \ \Lambda_m^H &= \sum_{t=1}^N \mathbb{E}_{Q(\psi_m)}(\psi_m) \mathbb{E}_{Q(\mathbf{z}_t)}(\mathbf{z}_t \mathbf{z}_t^T) + \mathbb{E}_{Q(eta_m)}(eta_m) \ \mu_m^{\mathbf{H}} &= (\Lambda_m^H)^{-1} \mathbb{E}_{Q(\psi_m)}(\psi_m) \sum_{t=1}^N \mathbb{E}_{Q(\mathbf{z}_t)}(\mathbf{z}_t) f_m^t \ \mathbb{E}_{Q(lpha_m)}(lpha_m) &=
u_m^{lpha} \mathbf{V}_m^{lpha} \ \mathbb{E}_{Q(\mathbf{z}_t)}(\mathbf{z}_t \mathbf{z}_t^T) &= (\Lambda_t^z)^{-1} + \mu_t^z (\mu_t^z)^T \end{aligned}$$

update $\mathcal{Q}(au)$, $\mathcal{Q}(\psi)$

$$egin{aligned} \lambda_m^{ au} &= j_m^{ au} + rac{1}{2}N \
u_m^{ au} &= k_m^{ au} + rac{1}{2}\sum_{t=1}^N \mathbb{E}_{Q(\mathbf{z}_t)Q(\mathbf{W}_m)}\{(p_m^t - \mathbf{W}_m^T\mathbf{z}_t)^T(p_m^t - \mathbf{W}_m^T\mathbf{z}_t)\} \ \lambda_m^{\psi} &= j_m^{\psi} + rac{1}{2}N \
u_m^{\psi} &= k_m^{\psi} + rac{1}{2}\sum_{t=1}^N \mathbb{E}_{Q(\mathbf{z}_t)Q(\mathbf{H}_m)}\{(f_m^t - \mathbf{H}_m^T\mathbf{z}_t)^T(f_m^t - \mathbf{H}_m^T\mathbf{z}_t)\} \
onumber \ \mathbb{E}_{Q(\mathbf{z})}(\mathbf{z}^T\mathbf{z}) &= \mathrm{Tr}(\Lambda_z^{-1}) + \mu_z^T\mu_z \end{aligned}$$

update $\mathcal{Q}(\alpha)$, $\mathcal{Q}(\beta)$

$$egin{aligned}
u_m^lpha &= a_m^lpha + 1 \ (\mathbf{V}_m^lpha)^{-1} &= (\mathbf{B}_m^lpha)^{-1} + \mathbb{E}_{Q(\mathbf{W}_m)}\{\mathbf{W}_m\mathbf{W}_m^T\} \
u_m^eta &= a_m^eta + 1 \ (\mathbf{V}_m^eta)^{-1} &= (\mathbf{B}_m^eta)^{-1} + \mathbb{E}_{Q(\mathbf{H}_m)}\{\mathbf{H}_m\mathbf{H}_m^T\} \end{aligned}$$

Monitoring Strategy

Monitoring Strategy is completely established according to the aforementioned reference, where SPE and T^2 are constructed. The specific calculations are given as follows:

$\operatorname{index} T^2$

$$\begin{split} T^2 &= \mathbf{z}_{new}^T \mathbf{z}_{new} \\ \text{where} \\ \mathbf{z}_{new} &= \Lambda_t^{\mathbf{z}} (\mathbf{z}_{new}^p + \mathbf{z}_{new}^f) \\ \mathbf{z}_{new}^p &= (\mu^W)^T \tau \mathbf{p}_{new} \\ \mathbf{z}_{new}^f &= (\mu^H)^T \psi \mathbf{f}_{new} \\ \tau &= diag\{ [\mathbb{E}_{Q(\tau_m)}(\tau_m)]_{m=1}^{Ml} \} \\ \psi &= diag\{ [\mathbb{E}_{Q(\psi_m)}(\psi_m)]_{m=1}^{M(s+1)} \} \end{split}$$

${\rm index}\; SPE$

$$SPE = \delta_{new}^T \psi \delta_{new} \ \delta_{new} = \mathbf{f}_{new} - \mu^H \mathbf{z}_{new}$$