

Exercises- 23

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1. Cauchy distribution = $T(x|0,1,1)$.

$$T(x|\mu, \sigma^2, \nu) = \frac{1}{Z} \left(1 + \frac{1}{\nu} \left(\frac{x-\mu}{\sigma} \right)^2 \right)^{-\frac{(\nu+1)}{2}}.$$

Thus, $T(x|0,1,1) = \frac{1}{Z} \left(1 + x^2 \right)^{-1} = \frac{1}{\pi} \frac{1}{1+x^2}$.

and cdf of the distribution,

$$F(x) = \int_{-\infty}^x \frac{1}{\pi} \frac{1}{1+t^2} dt = \frac{\arctan x + \pi/2}{\pi}.$$

and $F'(p) = \tan(\pi(p - \frac{1}{2}))$.

Thus, we can sample from $Unif[-\frac{1}{2}, \frac{1}{2}]$ and transform it by $\tan \pi$.

2. The Cauchy distribution,

$$T(x|\mu, \sigma^2, 1) = \frac{1}{Z} \cdot \frac{1}{1 + \left(\frac{x-\mu}{\sigma} \right)^2}.$$

where $Z = \pi \sigma$.

The Gamma distribution,

$$Ga(x|a,b) = \frac{b^a}{\Gamma(a)} x^{a-1} e^{-bx}.$$

The optimal coefficients,

$$M = \max_x \left\{ \frac{Ga(x|a,b)}{T(x|\mu, \sigma^2, 1)} \right\}.$$

$$\frac{b^a}{\Gamma(a)} x^{a-1} e^{-bx} \geq \frac{1}{Z} \left(1 + \left(\frac{x-\mu}{\sigma} \right)^2 \right)^{-1}.$$

Similarly, the optimal value for μ and σ^2 should be,

$$\operatorname{argmin}_{\mu, \sigma^2} \{ M(\mu, \sigma^2|a,b) \}.$$

3. $z_t = f_t(z_{t-1}) + \mathcal{N}(0, Q_{t-1})$.

$$y_t = H_t z_t + \mathcal{N}(0, R_t).$$

Derive - $p(z_t|z_{t-1}, y_t)$ and $p(y_t|z_{t-1})$.

$$p(z_t|z_{t-1}) = \mathcal{N}(z_t | f_t(z_{t-1}), Q_{t-1}).$$

$$p(y_t|z_t) = \mathcal{N}(y_t | H_t z_t, R_t).$$

Now, $p(z_t|z_{t-1}, y_t) = \frac{p(z_t, z_{t-1}, y_t)}{p(z_{t-1}, y_t)}$.

$$\propto p(z_t|z_{t-1}) p(y_t|z_t)$$

$$\propto \exp \left\{ -\frac{1}{2} (z_t - f_t(z_{t-1}))^T Q_{t-1}^{-1} (z_t - f_t(z_{t-1})) \right\} \exp \left\{ -\frac{1}{2} (y_t - H_t z_t)^T R_t^{-1} (y_t - H_t z_t) \right\}.$$

$$\propto \exp \left\{ -\frac{1}{2} z_t^T Q_{t-1}^{-1} z_t - \frac{1}{2} f_t(z_{t-1})^T Q_{t-1}^{-1} f_t(z_{t-1}) + z_t^T Q_{t-1}^{-1} f_t(z_{t-1}) - \frac{1}{2} y_t^T R_t^{-1} y_t - \frac{1}{2} z_t^T H_t^T R_t^{-1} H_t z_t + y_t^T R_t^{-1} z_t^T H_t^T \right\}.$$

$$\Sigma = Q_{t-1}^{-1} + (H_t^T R_t^{-1} H_t)^{-1}$$

$$\mu = \Sigma (Q_{t-1}^{-1} f_t(z_{t-1}) + H_t^T R_t^{-1} y_t).$$

On the other hand,

$$p(y_t|z_{t-1}) = \int p(y_t, z_t|z_{t-1}) dz_t.$$

$$= \int p(y_t|z_t) p(z_t|z_{t-1}) dz_t,$$

$$\propto \exp \left\{ -\frac{1}{2} y_t^T R_t^{-1} y_t \right\} \exp \left\{ -\frac{1}{2} (f_t^T Q_{t-1}^{-1} + y_t^T R_t^{-1} H_t) \Sigma (Q_{t-1}^{-1} f_t + H_t^T R_t^{-1} y_t) \right\}.$$

Thus, we get,

$$\Sigma = (R_t^{-1} - R_t^{-1} H_t \Sigma H_t^T R_t^{-1})^{-1}$$

$$\mu = \Sigma R_t^{-1} H_t \Sigma Q_{t-1}^{-1} f_t(z_{t-1}).$$

x ————— x ————— x