

4-5-2023

Tilted densities

Even $X \sim \text{pdf } f(x)$

$$M(t) = E(e^{tx}) = \int e^{tx} f(x) dx \quad t \in \mathbb{R}$$

tilted density $f_t(x)$

$$f_t(x) = \frac{e^{tx} f(x)}{M(t)}$$

① $X \sim \text{Exp}(\lambda) \quad f(x) = \lambda e^{-\lambda x}, x \geq 0.$

$$f_t(x) = C e^{tx} e^{-\lambda x} = C e^{-(\lambda-t)x} \quad \begin{matrix} \uparrow \\ t < \lambda \\ \downarrow \\ (\lambda-t) e^{-(\lambda-t)x} \end{matrix}$$
$$\sim \text{Exp}(\lambda-t)$$

$$E_{f_t}(x) = \frac{1}{\lambda-t} \quad \uparrow_t$$

$$F_t(x) = 1 - e^{-\lambda(1-t)x} \quad \downarrow_t \quad 1 - F_t(x) \uparrow_t \forall x$$

$$P_{f_x}(x > z) \uparrow_t$$

② $X \sim \text{Bernoulli}(p)$

$$f(x) = \begin{cases} p, & x=1 \\ 1-p, & x=0 \end{cases} = p^x (1-p)^{1-x}, \quad x=0,1$$

$$M(t) = E(e^{tx}) = pe^t + (1-p)$$

$$f_t(x) = \frac{e^{tx} p^x (1-p)^{1-x}}{pe^t + 1-p} =$$

$$= \frac{(pe^t)^x \cdot (1-p)^{1-x}}{pe^t + 1-p} = q^x (1-q)^{1-x} \quad x=0,1$$

$$\boxed{q = \frac{pe^t}{pe^t + 1-p}}$$

$$X_{f_t} \sim \text{Ber}\left(\frac{pe^t}{pe^t + 1-p}\right)$$

③ $X \sim \mathcal{N}(\mu, \sigma^2)$

$$X|_{f_t} \sim \mathcal{N}(\mu + \sigma^2 t, \sigma^2)$$

Парасимплекс

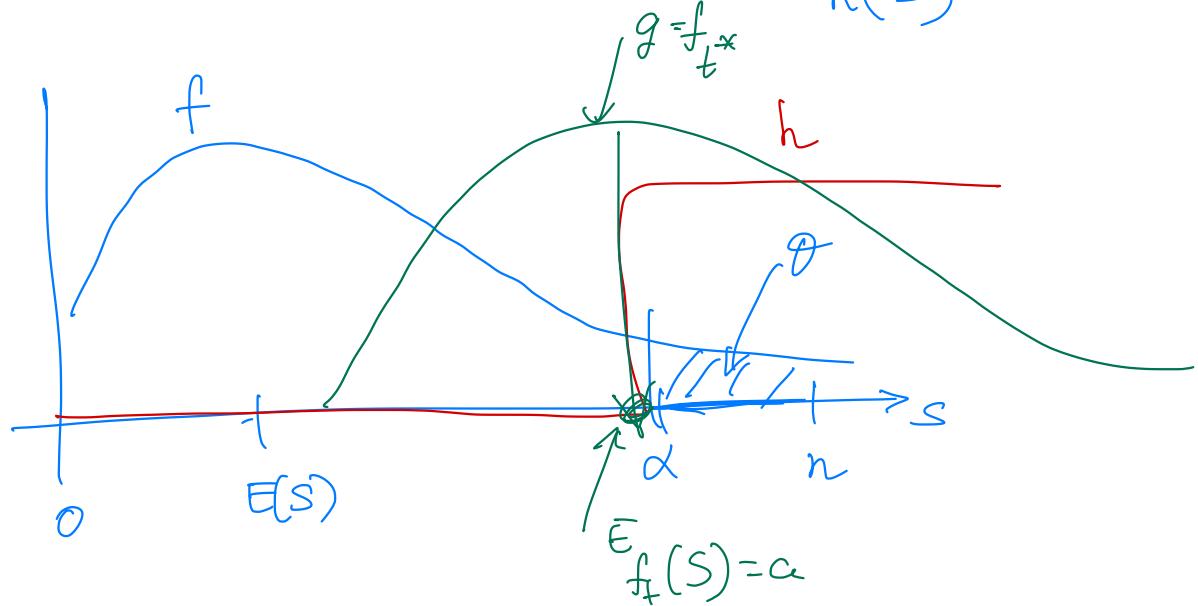
① X_1, \dots, X_n ares,

$$X_j \sim \text{Ber}(p_j)$$

$$S = X_1 + \dots + X_n$$

$$\theta = P(S \geq \alpha) \quad , \quad \text{или} \quad \alpha > E(S) = \sum p_j$$

$$\theta \approx 0 \quad \theta = E(\underbrace{1(S > \alpha)}_{h(s)})$$



$$\theta = E_f(1(S > \alpha))$$

$$f(x_1, \dots, x_n) = f_1(x_1) \dots f_n(x_n)$$

② Monte Carlo
N samples

$$\left\{ \begin{array}{l} \text{Sup. } X_1, \dots, X_n \\ \downarrow \\ \text{Ber}(p_1) \dots \text{Ber}(p_n) \end{array} \right.$$

$$h(S) = 1(S > \alpha)$$

2) Tilted densities new f_1, \dots, f_n for t :

$$f_t(x_1, \dots, x_n) = f_{t,1}(x_1) \cdots f_{t,n}(x_n)$$

$$\theta = E_{f_t} \left[1(S \geq \alpha) \cdot \frac{f(x_1, \dots, x_n)}{f_t(x_1, \dots, x_n)} \right]$$

$$= E_{f_t} \left[1(S \geq \alpha) \cdot \underbrace{\prod_{j=1}^n \frac{f_j(x'_j)}{f_{jt}(x_j)}}_{\hat{\theta}} \right]$$

$$f_{jt}(x_j) = \frac{e^{tx_j} f_j(x'_j)}{M_j(t)} \quad M_j(t) = E_{f_j}(e^{tx_j})$$

$$\frac{f_j(x_j)}{f_{jt}(x_j)} = e^{-tx_j} M_j(t)$$

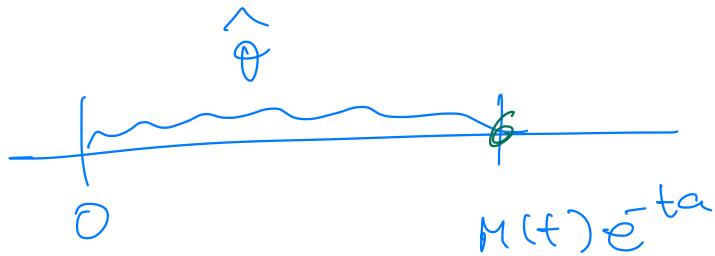
$$\hat{\theta} = 1(S \geq \alpha) \cdot e^{-t \underbrace{(x_1 + \dots + x_n)}_S} \underbrace{\prod_{j=1}^n M_j(t)}_{M(t)}$$

$$\hat{\theta} = 1(S \geq \alpha) \cdot \underbrace{M(t)}_{M(t)} e^{-tS}$$

$$M(t) = \prod_{j=1}^n (p_j e^{t+1-p_j})$$

$$1(s \geq a) e^{-ts} \leq e^{-ta} \quad \forall s.$$

$$\Rightarrow \boxed{\hat{\theta} \leq M(t) e^{-ta}}$$



En tējorofe t^* : $M(t) e^{-ta}$ ← min.

$$\begin{aligned} (M(t) e^{-ta})' &= M'(t) e^{-ta} - a M(t) e^{-ta} = \\ &= [M'(t) - a M(t)] e^{-ta} = 0 \end{aligned}$$

$$\Rightarrow M'(t) - a M(t) = 0'$$

$$\text{Otrais } M(t) = E(e^{ts})$$

$$M'(t) = E_f(S e^{ts})$$

$$\Rightarrow E_f [S e^{ts} - a e^{ts}] = 0 \Rightarrow$$

$$\Rightarrow E_f [(S-a) e^{ts}] = 0 \Rightarrow$$

$$\Rightarrow E_{f_{t^*}}(S-a) = 0 \Rightarrow$$

$$t^* \quad \boxed{E(S) = a} \quad \leftarrow \text{numerically - } f_{t^*}$$

$$f_{j,t} \quad \text{Bernoulli } (p_j e^t + (1-p_j))$$

$$E_{f_{t^*}}(S) = \sum_{i=1}^n \frac{p_i e^t}{p_i e^t + (1-p_i)} = a \quad (\text{solve for } t)$$

$$\sum_{i=1}^n \left[1 - \frac{1-p_i}{p_i e^t + (1-p_i)} \right] = a \Rightarrow$$

$$\Rightarrow \sum \frac{1-p_i}{p_i e^t + (1-p_i)}$$

Monte Carlo $\hat{\theta} = 1(S > a)$

$$\text{IS} \quad \hat{\theta} : 1(S > a) \cdot m(t) e^{-tS}$$

$$m(t) = \prod_{i=1}^n (p_i e^t + (1-p_i))$$