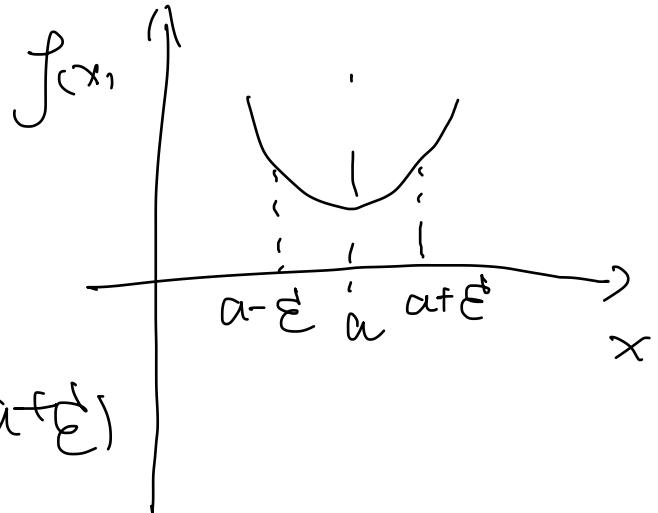


**2.26** Let  $f(x)$  be a pdf and let  $a$  be a number such that, for all  $\epsilon > 0$ ,  $f(a + \epsilon) = f(a - \epsilon)$ .

Such a pdf is said to be *symmetric* about the point  $a$ .

- (b) Show that if  $X \sim f(x)$ , symmetric, then the median of  $X$  (see Exercise 2.17) is the number  $a$ .

$$\text{Let } x = a + \epsilon, x - a = \epsilon$$



$$\int_a^\infty f_{x_1} dx = \int_0^\infty f(a + \epsilon) d(a + \epsilon)$$

$$= \int_0^\infty f(a + \epsilon) d\epsilon \quad \dots \quad \textcircled{1}$$

$$\text{Let } x = a - \epsilon, a - x = \epsilon$$

$$\int_{-\infty}^a f_{x_1} dx = \int_{-\infty}^0 f(a - \epsilon) d(a - \epsilon)$$

$$= \int_0^\infty f(a - \epsilon) d\epsilon \quad \dots \quad \textcircled{2}$$

$$\therefore f(a+\epsilon) = f(a-\epsilon)$$

$$\therefore \textcircled{1} = \textcircled{2}$$

$$\textcircled{1} + \textcircled{2} = \int_{-\infty}^a f(x_1) dx_1 + \int_a^{\infty} f(x_1) dx_1 = 1$$

Therefore,  $\int_{-\infty}^a f(x_1) dx_1 = \int_a^{\infty} f(x_1) dx_1 = \frac{1}{2}$

$a$  is the median of  $x$ .

**3.28** Show that each of the following families is an exponential family.

(b) gamma family with either parameter  $\alpha$  or  $\beta$  known or both unknown

$$f(x|\alpha, \beta) = \frac{1}{P(\alpha)\beta^\alpha} x^{\alpha-1} e^{-x/\beta}, \quad x > 0, \alpha > 0, \beta > 0$$

①: If  $\alpha$  known,

$$h(x) = \frac{x^{\alpha-1}}{P(\alpha)} \quad C(\beta) = \frac{1}{\beta^\alpha} \quad R = 1$$

$$W_1(\beta) = -\frac{1}{\beta} \quad t_1(x) = x.$$

Therefore, in this case, gamma family is an exponential family.

② :  $\beta$  known.

$$f(x|\alpha) = e^{-x/\beta} \cdot \frac{1}{P(\alpha, \beta)} e^{(\alpha-1) \cdot \log x}$$

$$= e^{-x/\beta} \cdot \frac{1}{P(\alpha, \beta)} e^{(\alpha-1) \cdot \log x}$$

$$h(x) = e^{-x/\beta} \quad C(\alpha) = \frac{1}{P(\alpha, \beta)} \quad k = 1$$

$$W_i(\alpha) = \alpha - 1 \quad t_i(x) = \log x$$

Therefore, in this case, gamma family is an exponential family, too,

③: both  $\alpha$  and  $\beta$  unknown

$$f(x|\alpha, \beta) = \frac{1}{P(\alpha) P(\beta)} x^{\alpha-1} e^{-\frac{x}{\beta}}$$

$$= \frac{1}{P(\alpha) P(\beta)} e^{(\alpha-1)\log x} \cdot e^{-\frac{x}{\beta}}$$

$$= \frac{1}{P(\alpha) P(\beta)} e^{(\alpha-1)\log x + \left(-\frac{x}{\beta}\right)}$$

$$h(x_1) : ]_{(x>0)}(x), C(\alpha, \beta) = \frac{1}{P(\alpha) P(\beta)}, R=2$$

$$W_1(\alpha) = \alpha - 1 \quad f_1(x) = \log x$$

$$W_2(\beta) = -\frac{1}{\beta} \quad f_2(x) = x$$

Therefore, in this case, gamma family is an exponential family, too,

(d) Poisson family

$$f(x|\lambda) = \frac{e^{-\lambda}\lambda^x}{x!} \quad x=0,1,2,\dots$$

$$= \frac{1}{x!} \cdot e^{-\lambda} \cdot e^{\log \lambda^x}$$

$$= \frac{1}{x!} \cdot e^{-\lambda} \cdot e^{x \log \lambda}$$

$$h(x) = \frac{1}{x!} \quad c(\lambda) = e^{-\lambda} \quad k=1$$

$$W_1(\lambda) = \log \lambda \quad t_1(x) = x.$$

Therefore, Poisson family is

exponential family -

3.37 Show that if  $f(x)$  is a pdf, symmetric about 0, then  $\mu$  is the median of the location-scale pdf  $(1/\sigma)f((x-\mu)/\sigma)$ ,  $-\infty < x < \infty$ .

$$\text{Let } x = \mu + \varepsilon$$

$$\frac{1}{\lambda} f\left(\frac{x-\mu}{\lambda}\right) = \frac{1}{\lambda} f\left(\frac{(\mu+\varepsilon)-\mu}{\lambda}\right)$$

$$= \frac{1}{\lambda} f\left(\frac{\varepsilon}{\lambda}\right) \cdots (*)$$

$\because f(x)$  is a pdf, symmetric about 0

$$\therefore f(x) = f(-x), \quad x \in \mathbb{R}$$

$$(*) = \frac{1}{\lambda} f\left(-\frac{\varepsilon}{\lambda}\right) = \frac{1}{\lambda} f\left(\frac{(\mu-\varepsilon)-\mu}{\lambda}\right)$$

Therefore,  $\mu$  is the median.

**3.39** Consider the Cauchy family defined in Section 3.3. This family can be extended to a location-scale family yielding pdfs of the form

$$f(x|\mu, \sigma) = \frac{1}{\sigma \pi \left( 1 + \left( \frac{x-\mu}{\sigma} \right)^2 \right)}, \quad -\infty < x < \infty.$$

The mean and variance do not exist for the Cauchy distribution. So the parameters  $\mu$  and  $\sigma^2$  are not the mean and variance. But they do have important meaning. Show that if  $X$  is a random variable with a Cauchy distribution with parameters  $\mu$  and  $\sigma$ , then:

(a)  $\mu$  is the median of the distribution of  $X$ , that is,  $P(X \geq \mu) = P(X \leq \mu) = \frac{1}{2}$ .

$$Z = \frac{x-\mu}{\sigma}$$

The standard Cauchy distribution

is that  $\frac{1}{\pi(1+Z^2)}$ ,

$$P(Z \geq 0) = \int_0^\infty \frac{1}{\pi(1+Z^2)} dZ$$

$$\int \frac{1}{1+Z^2} dZ = \tan^{-1}(Z)$$

$$\begin{aligned} \text{Therefore, } P(Z \geq 0) &= \left. \frac{1}{\pi} \tan^{-1}(Z) \right|_0^\infty \\ &= \frac{1}{\pi} \left( \frac{\pi}{2} - 0 \right) = \frac{1}{2} \end{aligned}$$

and  $Z=0$  is indeed the median for a standard Cauchy distribution, so  $\mu$  is the median of the distribution of  $X$ ,  $P(X \geq \mu) = P(X \leq \mu) = \frac{1}{2}$

3.42 Refer to Exercise 3.41 for the definition of a stochastically increasing family.

(a) Show that a location family is stochastically increasing in its location parameter.

$$\left. \begin{array}{l} X_1 \sim f(x - \theta_1) \\ X_2 \sim f(x - \theta_2) \\ \theta_1 > \theta_2 \\ Z \sim f(z) \end{array} \right\} \text{Cdf: } \left\{ \begin{array}{l} F(x| \theta_1) \\ F(x| \theta_2) \end{array} \right.$$

$$\begin{aligned} F(x|\theta_1) &= P(X_1 \leq x) = P(Z + \theta_1 \leq x) \\ &= P(Z \leq x - \theta_1) = F(x - \theta_1) \quad \dots \quad \textcircled{1} \end{aligned}$$

$$F(x|\theta_2) = \dots = F(x - \theta_2) \quad \dots \quad \textcircled{2}$$

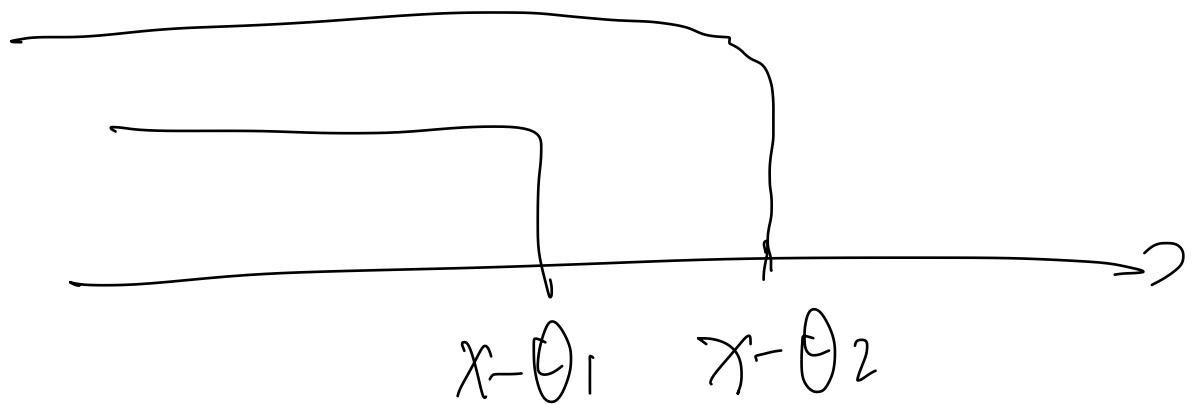
According to \textcircled{1}, \textcircled{2}, since  $\theta_1 > \theta_2$ .

$\Rightarrow x - \theta_2 > x - \theta_1$ , besides,  $F$  is

nondecreasing function.

$$\textcircled{1} - \textcircled{2} = F(x|\theta_1) - F(x|\theta_2)$$

$$= F(x-\theta_1) - F(x-\theta_2) < 0$$



$$F(x-\theta_1) < F(x-\theta_2)$$



$$F(x|\theta_1) < F(x|\theta_2)$$

3.46 Calculate  $P(|X - \mu_X| \geq k\sigma_X)$  for  $X \sim \text{uniform}(0, 1)$  and  $X \sim \text{exponential}(\lambda)$ , and compare your answers to the bound from Chebychev's Inequality.

$$\textcircled{1}: X \sim \text{uniform}(0, 1)$$

$$\mu = \frac{a+b}{2} = \frac{0+1}{2} = \frac{1}{2}$$

$$\sigma = \sqrt{\frac{(b-a)^2}{12}} = \frac{1}{\sqrt{12}} = \frac{1}{2\sqrt{3}} \quad \sigma^2 = \frac{1}{12}$$

$$P(|X - \mu| > k\sigma) = P(X - \mu > k\sigma) + P(X - \mu < -k\sigma)$$

$$= P(X > \mu + k\sigma) + P(X < \mu - k\sigma)$$

$$= 1 - P(\mu - k\sigma \leq X \leq \mu + k\sigma)$$

$$= 1 - P\left(\frac{1}{2} - \frac{k}{2\sqrt{3}} \leq X \leq \frac{1}{2} + \frac{k}{2\sqrt{3}}\right)$$

$$= (\cancel{*})$$

$$\begin{cases} \frac{1}{2} - \frac{k}{2\sqrt{2}} > 0 \\ \frac{1}{2} + \frac{k}{2\sqrt{2}} \leq 1 \end{cases} \Rightarrow k < \sqrt{3}$$

$$(*) = \begin{cases} 1 - \frac{2k}{\sqrt{12}}, & k < \sqrt{3} \\ 0, & \text{otherwise.} \end{cases}$$

$$k = 0.5 \quad (*) = 1 - \frac{1}{\sqrt{12}} = 0.711 \quad \varphi$$

$$k = 1 \quad (*) = 1 - \frac{2}{\sqrt{12}} = 0.423 \quad |$$

$$k = 1.5 \quad (*) = 1 - \frac{3}{\sqrt{12}} = 0.134 \quad 0.44$$

$$k = 2 \quad (*) = 0, \quad k = 2 > \sqrt{3} \quad 0.25$$

$$k = 3 \quad (*) = 0, \quad k = 3 > \sqrt{3}. \quad 0.11$$

②  $X \sim \text{exponential}(\lambda)$ ,  $\mu = \lambda$   
 $\sigma^2 = \lambda^2$

$$\begin{aligned} P(|X-\mu| > k) &= 1 - P(\lambda - \lambda k \leq X \leq \lambda + \lambda k) \\ &= (*) \end{aligned}$$

$$\therefore \lambda - \lambda k \geq 0$$

$$k \leq 1$$

$$\therefore (*) = \begin{cases} 1 + e^{-(k+1)} - e^{-k}, & k \leq 1 \\ e^{-(k+1)}, & k > 1 \end{cases}$$

$$\overbrace{\mathbb{R}^2}$$

$$k = 0.5 \leq |0.617| \quad \varphi$$

$$k = 1 \leq |0.135| \quad |$$

$$k = 1.5 > |0.082| \quad 0.88$$

$$k = 2 > |0.0898| \quad 0.25$$

$$k = 3 > |0.0183| \quad 0.11$$