

Exercise 5.21.  $\circ \frac{3}{4}$ 

$$\begin{aligned} \textcircled{1} P(\max(X_1, \dots, X_n) > n) &= 1 - P(\forall X_i, X_i \leq n) \\ &= 1 - \left(\frac{1}{2}\right)^n \end{aligned}$$

## Exercise 5.29

Let  $X$  denote the weight of booklets

Thus  $X_i$  is the weight of the  $i$ th booklet

Then  $E X_i = 1$ ;  $\text{Var } X_i = 0.05^2$

$$\begin{aligned} \text{Hence } P\left(\sum_{i=1}^{100} X_i > 100.4\right) &= P\left(\frac{1}{100} \sum_{i=1}^{100} X_i > 1.004\right) \\ &= P(\bar{X}_i > 1.004) \end{aligned}$$

According to Central Limit Theorem

$$P(\bar{X}_i > 1.004) = P\left(Z > \frac{0.004}{0.05/\sqrt{100}}\right) = 0.212$$

## Exercise 5.30

$$\bar{X}_1 \sim n(\mu, \sigma^2/n), \bar{X}_2 \sim n(\mu, \sigma^2/n)$$

$$\text{Hence } \bar{X}_1 - \bar{X}_2 \sim n(0, 2\sigma^2/n)$$

$$P(|\bar{X}_1 - \bar{X}_2| < \sigma/5) \approx 0.99$$

$$0.99 = P\left(\frac{-\sigma/5}{\sigma/\sqrt{n/2}} < \frac{\bar{X}_1 - \bar{X}_2}{\sigma/\sqrt{n/2}} < \frac{\sigma/5}{\sigma/\sqrt{n/2}}\right)$$

$$\text{Hence } 0.99 = P\left(-\frac{1}{5}\sqrt{\frac{n}{2}} < Z < \frac{1}{5}\sqrt{\frac{n}{2}}\right)$$

$$\frac{1}{5}\sqrt{\frac{n}{2}} = 2.576 \Rightarrow n = 50 \cdot (2.576)^2 = 332$$

## Exercise 5.31

$$\sigma^2 = 9; \sigma_{\bar{X}}^2 = 9/100$$

① By Chebychev's Inequality, we have

$$P(-3k/100 < \bar{X} - \mu < 3k/100) \geq 1 - 1/k^2 \geq 0.09 \Rightarrow k \geq \sqrt{10}$$

$$\text{Then } 3k/10 = 0.9487$$

$$P(-0.9487 < \bar{X} - \mu < 0.9487) \geq 0.9$$

② By CLT, we have

$$(\bar{X} - \mu) / \sqrt{0.09} \sim n(0, 1)$$

$$0.09 = P(-1.645 < \frac{\bar{X} - \mu}{0.3} < 1.645) = P(-0.4935 < \bar{X} - \mu < 0.4935)$$

Hence, the restriction of Chebychev's inequality is not as restricted as the normal approximation by CLT, which yields that Chebychev's inequality is a weak restriction.



Exercise 550.

$$X_1 = \cos(2\pi U_1) \sqrt{-2 \log U_2}$$

$$X_2 = \sin(2\pi U_1) \sqrt{-2 \log U_2}$$

We need to transform  $X$  and  $X_2$ .

$$\frac{X_1}{X_2} = \tan(2\pi U_1)$$

$$\textcircled{1} U_1 = \arctan\left(\frac{X_1}{X_2}\right) \cdot \frac{1}{2\pi}$$

$$X_1^2 + X_2^2 = -2 \log U_2 \Rightarrow \textcircled{2} U_2 = e^{-(X_1^2 + X_2^2)/2}$$

$\textcircled{3}$  Jacobian of transformation is

$$\begin{aligned} J &= \begin{vmatrix} \frac{\partial U_1}{\partial X_1} & \frac{\partial U_1}{\partial X_2} \\ \frac{\partial U_2}{\partial X_1} & \frac{\partial U_2}{\partial X_2} \end{vmatrix} = \begin{vmatrix} -\frac{1}{2\pi X_1} \frac{X_2}{(1 + \frac{X_2^2}{X_1^2})} & \frac{1}{2\pi X_1 (1 + \frac{X_2^2}{X_1^2})} \\ -X_1 e^{-(X_1^2 + X_2^2)/2} & -X_2 e^{-(X_1^2 + X_2^2)/2} \end{vmatrix} \\ &= \frac{1}{2\pi (X_1^2 + X_2^2)} e^{-(X_1^2 + X_2^2)/2} \cdot (X_1^2 + X_2^2) \\ &= \frac{1}{\sqrt{2\pi}} e^{-X_1^2/2} \cdot \frac{1}{\sqrt{2\pi}} e^{-X_2^2/2} \end{aligned}$$

Which is the product of the p.d.f of two normal variables  $X_1$  and  $X_2$ .

Hence, Q.E.D

