

**MDP Exercise 1.5\***. Let  $S = \frac{X}{\sqrt{1-Y^2}}$ ,  $T = \frac{Y}{\sqrt{1-X^2}}$ . To show that  $S \not\perp T$ , show that

$$(*) \quad P[S \leq 1/2, T \leq 1/2] \neq P[S \leq 1/2]P[T \leq 1/2].$$

Because  $S, T \sim \text{Uniform}(-1, 1)$ , the right side =  $(3/4)^2 = 9/16$ .

Now draw the region on the left side as the intersection of two semi-circular-type regions in the unit disk. The boundaries of the two semicircles intersect at the point  $(1/\sqrt{5}, 1/\sqrt{5})$ . This intersection includes the entire portion of the disk in the third quadrant, which has probability  $1/4$ . The intersection also includes parts of the disk in the second and fourth quadrants. By the uniform distributions of  $S$  and  $T$  and by Cavalieri's Principle, each of these has area equal to half of the disc quadrant, hence each has probability  $1/8$ , hence so far we have probability  $1/4 + 1/8 + 1/8 = 1/2$ .

Now examine the portion of the intersection that lies in the first quadrant of the disk. By the shape of the semi-circular boundaries, this portion contains the square with lower left vertex  $(0,0)$  and upper right vertex  $(1/\sqrt{5}, 1/\sqrt{5})$ , which square has probability  $1/5\pi$ . Thus the intersection on the left side of  $(*)$  has probability at least  $1/2 + 1/5\pi$ . But this already exceeds  $9/16$ , the probability of the right side of  $(*)$ .

### 512 Practice Exam Question.

Let  $Z_n$  be a sequence of random variables such that

$$\sqrt{n}(Z_n - \mu) \xrightarrow{d} N_1(0, \sigma^2)$$

for some constants  $\mu$  and  $\sigma^2 > 0$ . Use results in MDP §10 to show that

$$Y_n \xrightarrow{p} \mu.$$

**Solution.** Apply MDP Theorem 10.6 with  $X_n = \sqrt{n}(Z_n - \mu)$ ,  $Y_n = 1/\sqrt{n}$ , and  $c = 0$ .

**Exercise 10.6.** (i) Assume  $X_\lambda \sim \text{Poisson}(\lambda)$ . Find a variance-stabilizing transformation for  $X_\lambda$  as  $\lambda \rightarrow \infty$ . That is, find a function  $h$  and constant  $c > 0$  s.t.

$$(10.52) \quad [h(X_\lambda) - h(\lambda)] \xrightarrow{d} N_1(0, c).$$

Use this to obtain an approximate  $(1 - \alpha)$ -confidence interval for  $\lambda$ .

*Hint:* First, to find the form of the function  $h$  consider  $\lambda = n\theta$  where  $0 < \theta < \infty$  is fixed and  $n \rightarrow \infty$ . To obtain (10.52) for all values of  $\lambda$ , one approach is to use Slutsky's perturbation result in Theorem 10.4.

(ii) Let  $X_\lambda \sim \text{Poisson}(\lambda)$  and  $X_\mu \sim \text{Poisson}(\mu)$ , where  $X_\lambda \perp\!\!\!\perp X_\mu$  and  $\lambda$  and  $\mu$  are both large. Based on (i), describe an approximate procedure for testing  $\lambda = \mu$  vs.  $\lambda \neq \mu$ .

**Solution.** (i) From the hint,  $X_{n\theta} \sim \text{Poisson}(n\theta)$  so  $X_{n\theta} \sim Y_1 + \cdots + Y_n$  where the  $Y_i$  are i.i.d.  $\text{Poisson}(\theta)$  rvs. Therefore the CLTh implies that

$$\sqrt{n}(\bar{Y}_n - \theta) \xrightarrow{d} N(0, \theta),$$

so a variance-stabilizing transformation is given by  $h(\theta) = \int \frac{d\theta}{\sqrt{\theta}} = \sqrt{\theta}$ . Because  $h'(\theta) = \frac{1}{2\sqrt{\theta}}$  and  $\bar{Y}_n \sim \frac{X_{n\theta}}{n}$ ,

$$\begin{aligned} \sqrt{n}(\sqrt{\bar{Y}_n} - \sqrt{\theta}) &\xrightarrow{d} N(0, \frac{1}{4}), \\ \sqrt{n}\left(\frac{\sqrt{X_{n\theta}}}{\sqrt{n}} - \sqrt{\theta}\right) &\xrightarrow{d} N(0, \frac{1}{4}), \\ (1) \quad \sqrt{X_{n\theta}} - \sqrt{n\theta} &\xrightarrow{d} N(0, \frac{1}{4}). \end{aligned}$$

Now consider general (non-integral) values  $\lambda \rightarrow \infty$ . Let  $[\lambda]$  be the greatest integer  $\leq \lambda$ . Then from (1) with  $\theta = 1$ ,

$$(2) \quad \sqrt{X_{[\lambda]}} - \sqrt{[\lambda]} \xrightarrow{d} N(0, \frac{1}{4}) \quad \text{as } \lambda \rightarrow \infty.$$

By the Intermediate Value Theorem with  $h(\lambda) = \sqrt{\lambda}$ , for some  $\lambda^* \in ([\lambda], \lambda)$ ,

$$\begin{aligned} \sqrt{\lambda} - \sqrt{[\lambda]} &= (\lambda - [\lambda])h'(\lambda^*) = (\lambda - [\lambda])\frac{1}{2\sqrt{\lambda^*}} \\ (3) \quad &< (\lambda - [\lambda])\frac{1}{2\sqrt{[\lambda]}} = \frac{1}{2}\left(\frac{\lambda}{[\lambda]} - 1\right) \rightarrow 0 \quad \text{as } \lambda \rightarrow \infty. \end{aligned}$$

Next,  $X_\lambda = X_{[\lambda]} + V_\lambda$  where  $V_\lambda \sim \text{Poisson}(\lambda - [\lambda])$  and  $X_{[\lambda]} \perp\!\!\!\perp V_\lambda$ . As in (3), if  $X_{[\lambda]} > 0$  then for some intermediate value  $X^* \in (X_{[\lambda]}, X_\lambda)$ ,

$$(4) \quad \sqrt{X_\lambda} - \sqrt{X_{[\lambda]}} = V_\lambda h'(X^*) = V_\lambda \frac{1}{2\sqrt{X^*}} < V_\lambda \frac{1}{2\sqrt{X_{[\lambda]}}} \xrightarrow{p} 0 \quad \text{as } \lambda \rightarrow \infty,$$

because  $V_\lambda \preceq_{\text{stoch}} \text{Poisson}(1)$  (since  $0 \leq \lambda - [\lambda] \leq 1$ ) and  $\sqrt{X_{[\lambda]}} \xrightarrow{p} \infty$ . Because  $\Pr[X_{[\lambda]} > 0] \rightarrow 1$  as  $\lambda \rightarrow \infty$ , it follows from (3) and (4) that

$$(5) \quad (\sqrt{X_\lambda} - \lambda) - (\sqrt{X_{[\lambda]}} - \sqrt{[\lambda]}) \xrightarrow{p} 0 \quad \text{as } \lambda \rightarrow \infty,$$

so by (2) and Slutsky's Theorem,

$$(6) \quad \sqrt{X_\lambda} - \sqrt{\lambda} \xrightarrow{d} N(0, \frac{1}{4}).$$