

$$3.1 (a) \quad P_n(x) = \binom{n}{x} p^x (1-p)^{n-x}$$

$$P_n(x) = \frac{n!}{(n-x)! x!} p^x (1-p)^{n-x}$$

$$\ln [P_n(x)] = \ln(n!) - \ln[(n-x)!] - \ln x! + \ln p^x + \ln [1-p]^{n-x}$$

For large  $n$ .

$$\Rightarrow \ln(n!) = n \ln n - n \approx n \ln(n)$$

$$\begin{aligned} \therefore \ln [P_n(x)] &= n \ln(n) - (n-x) \ln(n-x) - \ln x! \\ &\quad + x \ln p + (n-x) \ln(1-p) \end{aligned}$$

$$n \ln(n) \approx n \ln(n-x)$$

$$\therefore \ln [P_n(x)] = x \ln(n-x) - \ln x! + x \ln(p) + (n-x) \ln(1-p)$$

For large  $n$  small  $p$   $(n-x) \ln(1-p) = -np$

$$\therefore \ln [P_n(x)] = x \ln(n) - \ln x! + x \ln(p) - np$$

$$P_n(x) = \frac{n^x p^x e^{-np}}{x!} = \frac{e^{-np} (np)^x}{x!} = \frac{e^{-N} N^x}{x!}$$

where  $\underline{N = np}$

3.1 (b) We have  $\langle x(x-1) \dots (x-m+1) \rangle$

$$= \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{x!} x(x-1) \dots (x-m+1)$$

For  $x \leq m$  ~~the~~  $x(x-1) \dots (x-m+1) = 0$ .

$$\Rightarrow \sum_{x=m}^{\infty} \frac{e^{-N} N^x}{x!} x(x-1) \dots (x-m+1)$$

$$\Rightarrow \sum_{x=m}^{\infty} \frac{e^{-N} N^x}{(x-m)(x-m-1) \dots 1}$$

Let  ~~$x \leq m$~~   $y = x - m$ ,  $x = y + m$

$$\Rightarrow \sum_{y=0}^{\infty} \frac{e^{-N} N^{y+m}}{y(y-1) \dots 1}$$

$$\Rightarrow \sum_{y=0}^{\infty} \frac{e^{-N} N^y \cdot N^m}{y!} = N^m \sum_{y=0}^{\infty} \frac{e^{-N} N^y}{y!} = N^m e^{-N} e^N = N^m$$

$$\boxed{= N^m}$$

$$3.1 \langle c \rangle \quad \langle x \rangle = \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{x!} x = \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{(x-1)!} = \sum_{x=1}^{\infty} \frac{e^{-N} N^{x-1}}{(x-1)!}$$

$$= N \sum_{x-1=0}^{\infty} \frac{e^{-N} N^{x-1}}{(x-1)!} = N e^{-N} e^N = N$$

$$\langle x^2 \rangle = \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{x!} x^2 = \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{x!} (x-1+1)$$

$$= \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{x!} x(x-1) + \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{x!} x$$

$$= \sum_{x=0}^{\infty} \frac{e^{-N} N^x}{(x-2)!} + N$$

$$= N^2 + N$$

$$\sigma^2 = \langle x^2 \rangle - \langle x \rangle^2 = N^2 + N - N^2 = N$$

$$\sigma = \sqrt{N}$$

$$\frac{\sigma}{\langle x \rangle} = \frac{\sqrt{N}}{N} = \frac{1}{\sqrt{N}}$$

3.2) Fractional error in estimation of average value is given by Relative Standard deviation

$$\text{For Poisson distribution } \frac{\sigma}{E[x]} = \frac{1}{\sqrt{N}}$$

$$\text{Given error} = 1\% = 0.01 = \frac{1}{100}$$

$$\sqrt{N} = 100 \Rightarrow N = 10^4$$

$$\text{Also for } \sqrt{N} = 10^6 \Rightarrow N = 10^{12}$$

$$\text{For visible light energy is } E = \frac{hc}{\lambda}$$

$$\text{where } \lambda = 700 \text{ nm}$$

$$\text{For } n \text{ photons } \frac{E}{s} = n \frac{hc}{\lambda}$$

$$\text{For } N = 10^4 \quad \frac{E}{s} = \frac{10^4}{s} \times 6.6 \times 10^{-34} \text{ J}\cdot\text{s} \times \frac{3 \times 10^8 \text{ m/s}}{7 \times 10^{-7} \text{ m}}$$

$$= 2.82 \times 10^{-30} \times 10^{15} \text{ J/s}$$

$$P = 2.82 \times 10^{-15} \text{ W}$$

$$\text{For } N = 10^{12} \quad P = 2.82 \times 10^{-7} \text{ W}$$

3.3 (a) Given  $R = 10\text{k}\Omega$ ,  $\text{SNR} = 20\text{dB}$ ,  $\Delta F = 20\text{kHz}$

Johnson Noise is given by

$$\langle V_{\text{noise}}^2 \rangle = 4kTR\Delta F$$

$$\langle V_{\text{noise}}^2 \rangle = 4 \times 1.38 \times 10^{-23} \text{ J/K} \times 300\text{K} \times \frac{10000}{\Omega} \times 20 \times 10^3 \text{ Hz}$$

$$\langle V_{\text{noise}}^2 \rangle = 3.3 \times 10^{-12} \text{ V}^2$$

$$V_{\text{noise}} = 1.8 \times 10^{-6} \text{ V}$$

$$10 \log \left( \frac{\langle V_{\text{signal}}^2 \rangle}{3.3 \times 10^{-12}} \right) = 20$$

$$\log \left( \frac{\langle V_{\text{signal}}^2 \rangle}{3.3 \times 10^{-12}} \right) = 2$$

$$\langle V_{\text{signal}} \rangle = \underline{\underline{1.8 \times 10^5 \text{ V}}}$$

$$(b) \quad \langle \frac{1}{2} CV^2 \rangle = \frac{1}{2} kT$$

$$C = \frac{kT}{\langle V^2 \rangle} = \frac{1.38 \times 10^{-3} \times 300}{3.3 \times 10^{-12}} \text{ F}$$

$$= 1.2 \times 10^{-9} \text{ F} = 1.2 \text{ nF}$$

$$(c) \quad \langle I_{\text{noise}}^2 \rangle = 2q \langle I \rangle \Delta F, \text{ given } I_{\text{noise, RMS}} = 0.01 \langle I \rangle$$

$$\therefore 10^{-4} \langle I^2 \rangle = 2q \langle I \rangle \Delta F, \quad \langle I \rangle = \frac{2 \times 1.6 \times 10^{-19} \times 20 \times 10^3}{10^{-4}} \text{ A}$$

$$\langle I \rangle = \frac{10^{-4}}{2q \times 20\text{kHz}} \neq 64 \times 10^{-12} \text{ A}$$

3.4(a) Let probability of being in state 0 at time  $t$  be  $P_0(t)$  and similarly for state 1 we have  $P_1(t)$

$$\text{So we get } \frac{dP_0}{dt} = -\alpha P_0 + \beta P_1$$

$$\frac{dP_1}{dt} = \alpha P_0 - \beta P_1$$

$$\frac{d}{dt} \begin{bmatrix} P_0 \\ P_1 \end{bmatrix} = \begin{bmatrix} -\alpha & \beta \\ \alpha & -\beta \end{bmatrix} \begin{bmatrix} P_0 \\ P_1 \end{bmatrix}$$

$$\frac{d}{dt} \vec{P} = A \cdot \vec{P}, \text{ where } A = \begin{bmatrix} -\alpha & \beta \\ \alpha & -\beta \end{bmatrix}$$

(b) Diagonalization

characteristic equation for  $A$  is  $|A - \lambda I| = 0$

$$\begin{vmatrix} -\alpha - \lambda & \beta \\ \alpha & -\beta - \lambda \end{vmatrix} = 0$$

$$(-\alpha - \lambda)(-\beta - \lambda) - \alpha\beta = 0 \Rightarrow \alpha\beta + \alpha\lambda + \beta\lambda + \lambda^2 - \alpha\beta = 0$$

$$\Rightarrow \lambda^2 + \lambda(\alpha + \beta) = 0 \Rightarrow \lambda(\lambda + \alpha + \beta) = 0$$

Eigen values are  $\lambda = 0$  and  $\lambda = -(\alpha + \beta)$

For  $\lambda = 0$ , Eigen vector is given by should satisfy

$$(A - \lambda I)x = 0$$

$$\begin{bmatrix} -\alpha & \beta \\ \alpha & -\beta \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$$-\alpha x_1 + \beta x_2 = 0$$

~~$\alpha x_1 + \beta x_2 = 0$~~   $(\alpha x_1, -\beta x_2) = 0$

~~$\alpha x_1 + \beta x_2 = 0$~~  on  $\boxed{x_1 = \frac{\beta}{\alpha} x_2}$  — (1)

For  $\lambda = -(\alpha + \beta)$ , Eigen vector is

$$(A - \lambda I)x = 0$$

$$\begin{bmatrix} -\alpha + \alpha + \beta & \beta \\ \alpha & -\beta + \alpha + \beta \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

$\beta x_1 + \beta x_2$   ~~$\alpha x_1 + \beta x_2 = 0$~~   $= 0$

$\alpha$   ~~$(\alpha x_1 + \beta x_2) = 0$~~   $(x_1 + x_2) = 0 \Rightarrow x_1 = -x_2$ . — (2)

From (1) & (2)

Eigen vectors are  $\begin{bmatrix} 1 \\ \alpha/\beta \end{bmatrix}$  and  $\begin{bmatrix} 1 \\ -1 \end{bmatrix}$

Some have  $C = \begin{bmatrix} 1 & 1 \\ \alpha/\beta & -1 \end{bmatrix}$

$$\det C = -1 - \alpha/\beta$$

$$C^{-1} = \frac{1}{-(1+\alpha/\beta)} \begin{bmatrix} -1 & -1 \\ -\alpha/\beta & 1 \end{bmatrix}$$

$$C^{-1} = \begin{bmatrix} \frac{1}{1+\alpha/\beta} & \frac{1}{1+\alpha/\beta} \\ \frac{\alpha/\beta}{1+\alpha/\beta} & \frac{1}{1+\alpha/\beta} \end{bmatrix}$$