

From plots, it can be observed that Chi-square is not sub gaussian except when  $n \rightarrow \infty$ , it will tend to gaussian.

$$3) f(t) = \sup_{\lambda} (\lambda t - \log(E(e^{\lambda X}))$$

$X$  - centered.

$$X = \begin{cases} 1, & p \\ 0, & 1-p \end{cases}$$

$$\Rightarrow Y = X - p = \begin{cases} 1-p, & p \\ -p, & 1-p \end{cases}$$

$$E(Y) = 0$$

$$\Rightarrow E(e^{\lambda Y}) = e^{-\lambda p} (1-p + p e^{\lambda})$$

$$\Rightarrow f(t) = \sup_{\lambda} (\lambda t + \lambda p - \log(1-p + p e^{\lambda}))$$

$\underbrace{\lambda p - \log(1-p + p e^{\lambda})}_{g(\lambda)}$

$$g'(\lambda) = 0$$

$$\Rightarrow p+t - \frac{pe^\lambda}{1-p+pe^\lambda} = 0$$

$$\Rightarrow pe^\lambda = (p+t)(1-p) + (p+t)pe^\lambda$$

$$\Rightarrow e^\lambda = \frac{(p+t)(1-p)}{p(1-p-t)}$$

$$\Rightarrow \lambda = \log \left( \frac{(p+t)(1-p)}{p(1-p-t)} \right)$$

(wherever defined)

$$\text{say } k = \frac{(p+t)(1-p)}{p(1-p-t)}$$

$$\Rightarrow \lambda = \log k$$

~~$$\Rightarrow f(t) = (t+p) \log k$$~~

$$\Rightarrow f(t) = (p+t) \log k - \log(1-p+kp)$$

$$\Rightarrow f(t) = (p+t) \log k - \log((k-1)p+1)$$

$$\text{where } k = \frac{(p+t)(1-p)}{p(1-(p+t))}$$

$\forall t$  wherever

$f(t)$  is defined

4.) Bennett's Inequality :-

$x_1, x_2, \dots, x_n$  - centred

$$x_i \leq b, \quad b > 0$$

$$v = \sum_{i=1}^n x_i^2, \quad S = \sum_{i=1}^n x_i$$

$$\frac{v}{b^2} \phi(\lambda b)$$

$$E(e^{\lambda S}) \leq e^{\frac{v}{b^2} \phi(\lambda b)}$$

$$P(S \geq t) = P(e^{\lambda S} \geq e^{\lambda t}) \leq \frac{E(e^{\lambda S})}{e^{\lambda t}}$$

$$\Rightarrow P(S \geq t) \leq e^{-\lambda t + \frac{v}{b^2} \phi(\lambda b)}$$

$$f(\lambda) = \frac{\theta}{b^2} \phi(\lambda b) - \lambda t$$

$$f'(\lambda) = 0$$

$$\Rightarrow \frac{\theta}{b^2} \cdot \phi'(\lambda b) \cdot b = t$$

$$\Rightarrow \phi'(\lambda b) = \frac{bt}{\theta}$$

$\Rightarrow$

$$\phi(\lambda b) = e^{\lambda b} - \lambda b - 1$$

$$\phi'(\lambda b) = e^{\lambda b} - 1$$

$$\left( \frac{\phi}{\phi'(\lambda b)} \right)$$

$$\Rightarrow e^{\lambda b} = 1 + \frac{bt}{\theta}$$

$$\Rightarrow \lambda = \left( \frac{1}{b} \right) \log \left( 1 + \frac{bt}{\theta} \right)$$

~~$$\Rightarrow P(S \geq t) \leq e^{-\frac{t}{b} \log \left( 1 + \frac{bt}{\theta} \right)}$$~~



~~P(x > t)~~

$$\begin{aligned}
 \Rightarrow P(S > t) &\leq e^{-\lambda t + \frac{\nu}{b^2} (e^{\lambda b} - \lambda b - 1)} \\
 &= e^{-\frac{t}{b} \log\left(1 + \frac{bt}{\nu}\right) + \frac{\nu}{b^2} \left(\frac{tb}{\nu} - \log\left(1 + \frac{tb}{\nu}\right)\right)} \\
 &= e^{\underbrace{\left(-\frac{t}{b} \log\left(1 + \frac{tb}{\nu}\right) + \frac{t}{b} - \frac{\nu}{b^2} \log\left(1 + \frac{tb}{\nu}\right)\right)}_{h(t)}}
 \end{aligned}$$

$$h(t) = \frac{\nu}{b^2} \left( \frac{tb}{\nu} - \log\left(1 + \frac{tb}{\nu}\right) - \frac{tb}{\nu} \log\left(1 + \frac{tb}{\nu}\right) \right)$$

$$= -\frac{\nu}{b^2} \left( \left(1 + \frac{tb}{\nu}\right) \log\left(1 + \frac{tb}{\nu}\right) - \frac{tb}{\nu} \right)$$

$$\Rightarrow \boxed{P(S > t) \leq e^{-\frac{\nu}{b^2} \left( h\left(\frac{bt}{\nu}\right) \right)}}$$

$$h(u) = (1+u) \log(1+u) - u$$