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$$\begin{cases} E[\pi_i(\alpha)^{2q}] \leq \frac{q!}{2} (4v)^q \\ * \Rightarrow P\left[\left|\sum_{i=1}^k (\pi_i(\alpha)^2 - 1)\right| \geq 4v\sqrt{2kt} + 4vt\right] \leq 2e^{-t} \end{cases} \quad \underline{1}$$

(1)

Gamma Distribution.

Consider a r.v.  $Y$  with gamma dist  $\varepsilon, a, b \geq 0$

$$E[e^{tY}] = \frac{1}{\Gamma(a) b^a} \int_0^{\infty} 2^{a-1} e^{-2\left(\frac{1}{b} - t\right)} \dots$$

$$\begin{aligned} EY &= ab \\ \text{Var } Y &= ab^2 \\ &\neq 2 \end{aligned}$$

ln ↓

$$\psi_Y(t) = -a \ln(1-bt)$$

$$X \equiv Y - EY \quad X \text{ is centered}$$

$$\psi_X(t) = \ln E e^{t(Y-EY)} = -abt - a \ln(1-bt) \rightarrow 0 < t \leq \frac{1}{b}$$

$0 < bt < 1$

$$\text{If } t \in (0,1) \quad -t - \ln(1-t) \leq \frac{t^2}{2(1-t)}$$

$$\psi_X(t) \leq \frac{ab^2 t^2}{2(1-bt)}$$

- Gamma dists have fractals. (fractal distribution)

Defn: A r.v.  $Z$  is said to be sub-gamma on the right

tail with  $\text{Var}(Z) = v$ ,  
Scale  $(Z) = c$

if  $\psi_z(t) \leq \frac{v^2}{2(1-ct)}$   $t \in (0, \frac{1}{c})$

$$Z \in \Gamma_+(v, c)$$

$$Z \in \Gamma_-(v, c)$$

Addition of normal = normal  
 " of (normal)<sup>2</sup> = chi-square

Compute Cramer's Transform.

$$\psi_z^*(t) \triangleq \sup_{\lambda \geq 0} (\lambda t - \psi_z(\lambda))$$

$$\psi_z(\lambda) \text{ is defined } \lambda \in (0, \frac{1}{c}), \quad \lambda = \frac{1}{c} \left( 1 - \frac{1}{\sqrt{1 + \frac{2ct}{v}}} \right)$$

$$\psi_z^*(t) \geq \frac{v}{c^2} \left( 1 + \frac{ct}{v} - \sqrt{1 + \frac{2ct}{v}} \right) = \frac{v}{c^2} h\left(\frac{ct}{v}\right)$$

$$h(u) = 1 + u - \sqrt{1 + 2u}, \quad h^{-1}(u) = u + \sqrt{2u}$$

$$\text{If } Z \in \Gamma_+(v, c) \quad P[Z > t] \leq e^{-\frac{v}{c^2} h\left(\frac{ct}{v}\right)}$$

$$f^{-1} \circ f(x) = x \quad \text{and} \quad f(x) = y$$

$f(x)$  is increasing.

$$\frac{d f^{-1}}{d y} \cdot f'(x) = 1 \Rightarrow \frac{d f^{-1}}{d y} = \frac{1}{f'(x)} > 0, \quad f'(x) > 0, \text{ so } f^{-1} \text{ is also increasing.}$$

$\times \left\{ \begin{array}{l} \text{Decide } h(u) \text{ is increasing / decreasing} \\ h' = 1 - \frac{1}{2} (1+2u)^{-1/2} \cdot z = 1 - \frac{1}{(1+2u)^{1/2}} \end{array} \right. \quad [1+2u]^{1/2} \quad \underline{3}$

$a = \frac{ct}{v} \quad P\left[z > \frac{va}{c}\right] \leq e^{-v/c^2 h(a)}$

$P\left[z > \frac{v}{c} (u + \sqrt{2u})\right] \leq e^{-\frac{v}{c} u}$

$\left. \begin{array}{l} z \in \Gamma_+(v, c) \\ t = \frac{vu}{c^2} \end{array} \right\} P[z > ct + z\sqrt{vt}] \leq e^{-t}$

\* the bounds of sub-gaussians are valid sub-gamma.

Bernstein's Inequality — if you have these conditions, they are sub-gamma.

Let  $X_1, X_2, \dots, X_k$  be an independent real-valued r.v.s.

Assume  $\exists v, c > 0$  s.t.  $\sum_{i=1}^k E[X_i]^2 < v$  and

$\sum_{i=1}^k E[(X_i)^q] \leq \frac{1}{2} q! v c^{q-2}$  for  $\forall$  integers  $q \geq 3$ . Then, (\*) looks similar to this

if I define  $S = \sum_{i=1}^k (X_i - EX_i)$  and for  $x \in (0, \frac{1}{c})$  and  $t > 0$ .

$\psi_S(\lambda) \leq \frac{v \lambda^2}{2(1-c)} \quad \text{and} \quad \psi_S^*(t) \geq \frac{v}{c^2} h\left(\frac{ct}{v}\right) \quad \text{and}$

$P[S \geq \underbrace{\sqrt{2ct}}_? + ct] \leq e^{-t}$

Proof: Part 1. PTP:  $\psi_s(\lambda) \leq \frac{v \lambda^2}{2(1-c)}$

4

Let's define.  $\phi(t) = e^t - t - 1, \lambda > 0$

$$\phi(\lambda X_i) = e^{\lambda X_i} - (\lambda X_i + 1) = \frac{\lambda^2 X_i^2}{2} + \sum_{q \geq 3} \frac{\lambda^q X_i^q}{q!}$$

monotonic  
convergence  
function.

$$\sum_{i=1}^k E[\phi(\lambda X_i)] \leq \sum_{i=1}^k \left( \frac{\lambda^2 E[X_i^2]}{2} + \sum_{q \geq 3} \frac{\lambda^q E[(X_i)^q]}{q!} \right) \leq \frac{v}{2} \sum_{q \geq 0} \lambda^q c^{q-2}$$

$$0 < \lambda \leq \frac{1}{c} \leq \frac{v \lambda^2}{2(1-c\lambda)}$$

$$\begin{aligned} \psi_s(\lambda) &= \ln E \left[ e^{\sum_i \lambda (X_i - E(X_i))} \right] = - \sum_i \phi(\lambda X_i) \\ &\leq \sum_i (E e^{\lambda X_i} - 1 - \lambda X_i) \\ &= \ln E \left[ \prod_i e^{\lambda (X_i - E(X_i))} \right] = \sum_i (\ln E e^{\lambda X_i} - \lambda E X_i) \\ &= \sum_i (\ln E e^{\lambda X_i} - \lambda E X_i) \quad \ln E [E X_i] \end{aligned}$$

$$x > 0, \ln x \leq x - 1.$$

$$\mathbb{P} \left[ \sup_{v \in T} \left| \sum_{i=1}^k (\pi_i(v)^2 - 1) \right| \geq 4v(\sqrt{2kt} + t) \right] \leq 2N^2 e^{-t}$$

$t$  has to be non-negative.

$$t = \ln \left( \frac{N^2}{8} \right)$$

$$P \left[ \sup_{v \in T} \left| \|\pi(v)\|^2 - 1 \right| \geq \frac{8v \ln \left( \frac{N}{\sqrt{\delta}} \right)}{k} + \sqrt{\frac{8v \ln \left( \frac{N}{\sqrt{\delta}} \right)}{k}} \right] \leq \delta$$

$$k \geq 100 \cdot \varepsilon^{-2} N^2 \ln \left( \frac{N}{\sqrt{\delta}} \right)$$

$$\varepsilon \geq \frac{2\varepsilon^2}{25v}$$

$$\leq \frac{4\varepsilon}{5}$$

$$P \left[ \sup_{v \in T} \left| \|\pi(v)\|^2 - 1 \right| \leq \varepsilon \right] \geq 1 - \delta$$

norm of projection = -1, +1

basis vectors are gaussian.  $\rightarrow$  the summation are  $\chi^2$   
 $\downarrow$   
 special case  
 of subgaussian