

MAT02023 - Inferência A

LISTA 2 - CONTINUAÇÃO DE CONCEITOS DE PROBABILIDADE

Amostra Aleatória

Exercício 1 (a) Descreva a σ álgebra e suas probabilidades.

(b) Uniforme discreta

(c) Calcule os momentos de uma uniforme discreta, exemplo $\mathbb{E}X = \sum_{k=1}^n kP(X = k)$.

(d)

(e)

(f) Binomial

Exercício 2 (a) $P(\mathbf{X} = \mathbf{x}) = \left(\frac{2}{3}\right)^{\sum_{i=1}^9 x_i} \left(\frac{1}{3}\right)^{9 - \sum_{i=1}^9 x_i}$.

(b) Defina $Y = \sum_{i=1}^9 X_i$, então $Y \sim \text{Binomial}(9, 2/3)$.

(c) $E(\bar{X}) = E\left(\frac{Y}{9}\right) = 9 \times \frac{2}{3} \times \frac{1}{9} = \frac{2}{3}$.

(d) $E(S^2) = \frac{n}{n-1} [E(X_i^2) - E(\bar{X}^2)] = \dots = \text{Var}(X_i)$. Então $E(S^2) = \frac{2}{9}$.

Exercício 3 (a) $P(\mathbf{X} = \mathbf{x}; \lambda) = \lambda^n e^{-\lambda \sum_{i=1}^n x_i}$.

(b) $e^{-2\lambda n}$

Exercício 4 (a)

(b)

(c)

(d)

Exercício 5 Dica: $X_i \sim \text{Bernoulli}(p) \implies \sum_{i=1}^n X_i \sim \text{Binomial}(n, p)$.

$$P(\bar{X} = x) = \binom{n}{nx} p^{nx} (1-p)^{n-x}.$$

(a) Dica: Soma de variáveis aleatórias com distribuição Poisson é ainda Poisson.

$$P(\bar{X} = k/n) = \frac{e^{-n\lambda} (n\lambda)^k}{k!}.$$

(b) Dica: $\text{Exp}(\lambda) = \Gamma(1, 1/\lambda)$ e soma de n variáveis $\Gamma(1, 1/\lambda)$ é $\Gamma(n, 1/\lambda)$.

Exercício 6 (a) $E(\bar{X}) = \bar{X}$ e $\text{Var}(\bar{X}) = \sigma^2/n$.

(b) $E(S^2) = \sigma^2$.

Função Geradora de Momentos

Exercício 7 $M_X(t) = e^{\lambda(e^t-1)}$.

Exercício 8 $M_X(t) = \left(\frac{\lambda}{\lambda-t}\right)$.

Exercício 9 $M_X(t) = \left(\frac{\beta}{\beta-t}\right)^\alpha$.

Exercício 10 usar $M_X(t)$.

Exercício 11 $M_X(t) = e^{\frac{t_1^2+t_2^2}{2}}$.

Teoremas Limite

Exercício 12 $l = 0,328$.

Exercício 13 $n \approx 62$.

Exercício 14 $P(0.45 < \bar{X} < 0.55) \approx 0,866$.

Exercício 15 (a) $Y(n, p)$.

(b) $P(47,5 < Y < 52,5) \approx 0,3829$.

Exercício 16 $P(Y/n > 0.25) \approx 0,0062$.

Exercício 17 $0,3085$.

Exercício 18

$$a) \lim_{m \rightarrow \infty} P(|\bar{x}_m - \mu| < \varepsilon) = 1$$

Pela desigualdade de Chebyshev: $P(g(n) \geq n) \leq \frac{E[g(n)]}{n}$

$$P(|\bar{x}_m - \mu| < \varepsilon) = 1, \text{ então } P(|\bar{x}_m - \mu| \geq \varepsilon) = 0$$

$$\lim_{m \rightarrow \infty} P(|\bar{x}_m - \mu| \geq \varepsilon^2) \leq \frac{E(\bar{x}_m - \mu)^2}{\varepsilon^2} = \lim_{m \rightarrow \infty} P(|\bar{x}_m - \mu| \geq \varepsilon^2) \leq \frac{\sigma^2}{m\varepsilon^2}$$

$$\frac{\sigma^2}{m\varepsilon^2} = \frac{\sigma^2}{\infty} = 0 \text{ na medida em que } m \text{ vai para infinito}$$

$$\text{Portanto, } \lim_{m \rightarrow \infty} P(|\bar{x}_m - \mu| < \varepsilon) = 1$$

$$b) Z_m = \frac{\bar{x}_m - E(\bar{x}_m)}{\sqrt{\text{Var}(\bar{x}_m)}} \sim N(0, 1) \text{ na medida em que } m \rightarrow \infty$$

$$m_{Z_m}(t) = \frac{\bar{x} - E(\bar{x})}{\sqrt{\text{Var}(\bar{x})}} = \frac{\bar{x} - \mu}{\sigma/\sqrt{m}} \xrightarrow{m \rightarrow \infty} N(0, 1)$$

$$m_{Z_m}(t) = E(e^{tZ_m}) = E\left(e^{t \left(\frac{\bar{x} - \mu}{\sigma/\sqrt{m}}\right)}\right) = E\left(e^{t \left(\frac{\sum_{i=1}^m x_i - m\mu}{\sigma\sqrt{m}}\right)}\right) = E\left(e^{\frac{t}{\sqrt{m}} \left[\sum_{i=1}^m (x_i - \mu)\right]}\right) = E\left(e^{\frac{t}{\sqrt{m}} \sum_{i=1}^m (x_i - \mu)}\right) \stackrel{\text{i.i.d.}}{=} \prod_{i=1}^m E\left(e^{t \left(\frac{x_i - \mu}{\sigma/\sqrt{m}}\right)}\right) \quad * \text{chamamos } y_i = \frac{x_i - \mu}{\sigma/\sqrt{m}}$$

Utilizando Série de Taylor.

$$m_{y_i}\left(\frac{t}{\sqrt{m}}\right)^m = \left[m_{y_i}(0) \cdot \frac{(t/\sqrt{m} - 0)^0}{0!} + m'_{y_i}(0) \frac{(t/\sqrt{m} - 0)^1}{1!} + m''_{y_i}(0) \frac{(t/\sqrt{m} - 0)^2}{2!} + R_y(t/\sqrt{m})\right]^m$$

$$= \left[1 + 0 + m \frac{t^2/m^2}{2!} + R_y(t/\sqrt{m})\right]^m = \left[1 + \frac{t^2}{2m} + R_y(t/\sqrt{m})\right]^m = \left[1 + \frac{1}{m} \left(\frac{t^2}{2} + m R_y(t/\sqrt{m})\right)\right]^m$$

Aplicamos no limite de $m \rightarrow \infty$

$$\lim_{m \rightarrow \infty} \left(1 + \frac{1}{m} \left(\frac{t^2}{2} + m R_y(t/\sqrt{m})\right)\right)^m = \exp\left\{\frac{t^2}{2} + m R_y(t/\sqrt{m})\right\} = \exp\left\{\frac{t^2}{2}\right\}$$

fgm da $N(0, 1)$

$$c) E(\bar{x}) = \mu \quad \text{e} \quad \text{Var}(\bar{x}) = \frac{1}{m} \sigma^2$$

$$E(\bar{x}) = E\left(\frac{\sum_{i=1}^m x_i}{m}\right) = E\left[\frac{1}{m} \sum_{i=1}^m x_i\right] = \frac{1}{m} E\left(\sum_{i=1}^m x_i\right) = \frac{1}{m} \sum_{i=1}^m E(x_i) = \frac{1}{m} \cdot m \mu = \mu //$$

$$\text{Var}(\bar{x}) = \text{Var}\left(\frac{1}{m} \sum_{i=1}^m x_i\right) = \frac{1}{m^2} \text{Var}\left(\sum_{i=1}^m x_i\right) = \frac{1}{m^2} \sum_{i=1}^m \text{Var}(x_i) = \frac{1}{m^2} \sum_{i=1}^m \sigma^2 = \frac{m \sigma^2}{m^2} = \frac{\sigma^2}{m} //$$

$$d) U = \sum_{i=1}^n \left(\frac{x_i - \mu_i}{\sigma_i}\right)^2 \sim \chi_n^2 \quad \text{sabemos que } Z_i = \frac{x_i - \mu_i}{\sigma_i} \sim N(0, 1)$$

$$\begin{aligned} m_U(t) &= E(e^{t \sum_{i=1}^n Z_i^2}) = E[\prod_{i=1}^n e^{t Z_i^2}] = \prod_{i=1}^n E(e^{t Z_i^2}) = \prod_{i=1}^n \int_{-\infty}^{\infty} e^{t z_i^2} (2\pi)^{-1/2} e^{-\frac{1}{2} z_i^2} dz_i = \\ &= \prod_{i=1}^n \int_{-\infty}^{\infty} (2\pi)^{-1/2} e^{-\frac{1}{2} z_i^2 (1-2t)} dz_i = \prod_{i=1}^n \int_{-\infty}^{\infty} (2\pi)^{-1/2} \left(\frac{1}{1-2t}\right)^{1/2} \left(\frac{1}{1-2t}\right)^{1/2} e^{-\frac{1}{2} \left(\frac{z_i}{\sqrt{1-2t}}\right)^2} dz_i = \\ &= \prod_{i=1}^n \frac{1}{\sqrt{1-2t}} \underbrace{\int_{-\infty}^{\infty} (2\pi)^{-1/2} \left(\frac{1}{1-2t}\right)^{1/2} e^{-\frac{1}{2} \left(\frac{z_i}{\sqrt{1-2t}}\right)^2} dz_i}_1 \end{aligned}$$

$$\text{então, } m_U(t) = \prod_{i=1}^n \left(\frac{1}{1-2t}\right) = \left[\frac{1}{1-2t}\right]^n \rightarrow \text{FMG de } \chi_n^2.$$

$$e) \text{ Se } U = \sum_{i=1}^n \left(\frac{x_i - \mu_i}{\sigma}\right)^2 \sim \chi_n^2$$

$$\text{então } \sum_{i=1}^m \frac{(x_i - \mu)^2}{\sigma^2} \sim \chi_m^2 \quad \text{pois } \sum_{i=1}^n \left(\frac{x_i - \mu}{\sigma}\right)^2 = \sum_{i=1}^m \frac{(x_i - \mu)^2}{\sigma^2} \quad \text{se } n=m$$

$$f) i. \bar{Z} \sim N(0, 1/m)$$

$$\bar{Z} = \frac{1}{m} \sum_{i=1}^m Z_i$$

$$\sum_{i=1}^m Z_i \sim N(\mu, \sigma^2) \quad \text{onde } \mu = \sum_{i=1}^m \mu_i = 0 + 0 + \dots + 0 = 0$$

$$\text{e } \sigma^2 = \text{Var}(\bar{Z}) = \text{Var}\left(\frac{\sum_{i=1}^m Z_i}{m}\right) = \frac{1}{m^2} \cdot m \sigma_i^2 = \frac{1}{m}$$

$$\therefore \bar{Z} \sim N(0, 1/m)$$

$$ii) \bar{Z} \text{ e } \sum_{i=1}^m (Z_i - \bar{Z})^2 \text{ são independentes. } (m=2)$$

$$\bar{Z} = \frac{Z_1 + Z_2}{2}$$

$$\sum_{i=1}^2 (Z_i - \bar{Z})^2 = \left[Z_1 - \frac{(Z_1 + Z_2)}{2}\right]^2 + \left[Z_2 - \frac{(Z_1 + Z_2)}{2}\right]^2 = \frac{(Z_1 - Z_2)^2}{4} + \frac{(Z_2 - Z_1)^2}{4}$$

$$\bar{z} = \sum_{i=1}^m \frac{z_i}{m} = \frac{z_1 + z_2}{m} \quad \text{e} \quad \sum_{i=1}^2 (z_i - \bar{z})^2 = \frac{(z_2 - z_1)^2}{2}$$

Termos que mostram que $z_1 + z_2$ e $z_2 - z_1$ são independentes

$$m_{z_1+z_2}(t_1) = E(e^{t_1(z_1+z_2)}) = E(e^{t_1 z_1 + t_1 z_2}) = \exp\left\{\frac{t_1^2}{2}\right\} \exp\left\{\frac{t_1^2}{2}\right\} = \exp\{t_1^2\}$$

$$\cdot m_{z_2-z_1}(t_2) = \exp\{t_2^2\} \quad (\text{Usando resultado anterior})$$

$$\begin{aligned} \text{Portanto, } m(t_1, t_2) &= E(e^{t_1(z_1+z_2) + t_2(z_2-z_1)}) = E(e^{z_1(t_1+t_2) + z_2(t_1+t_2-t_1)}) = \\ &= E(e^{z_1(t_1+t_2)}) \cdot E(e^{z_2(t_1+t_2-t_1)}) = [e^{\frac{t_1^2 + 2t_1 t_2 + t_2^2}{2}}] [e^{\frac{t_1^2 + 2t_1 t_2 + t_2^2}{2}}] = e^{t_1^2} \cdot e^{t_2^2} = \\ &= \exp\{t_1^2\} \cdot \exp\{t_2^2\} \end{aligned}$$

$$\star m_{z_1+z_2}(t_1) \cdot m_{z_2-z_1}(t_2) = m_{z_1+z_2, z_2-z_1}(t_1, t_2), \quad \text{então provamos a independência de } \bar{z} \text{ e } \sum_{i=1}^m (z_i - \bar{z})^2$$

$$\text{iii) } \sum_{i=1}^m (z_i - \bar{z})^2 \sim \chi_{m-1}^2$$

$$\sum z_i^2 - \underbrace{2z_1 \bar{z}}_{-2\bar{z}^2} + \bar{z}^2 + \dots + \sum z_m^2 - \underbrace{2z_m \bar{z}}_{-2\bar{z}^2} + \bar{z}^2 = \sum (z_i - \bar{z})^2 + m\bar{z}^2$$

$$m_{\sum z_i^2}(t) = m_{\sum (z_i - \bar{z})^2 + m\bar{z}^2}(t) = m_{\sum (z_i - \bar{z})^2}(t) \cdot m_{m\bar{z}^2}(t) = m_{\sum z_i}(t)$$

$$m_{\sum (z_i - \bar{z})^2}(t) = \frac{m_{\sum z_i^2}(t)}{m_{m\bar{z}^2}(t)} = \frac{m(t)}{\sum (z_i - \bar{z})^2} = \left(\frac{1}{1-2t}\right)^{\frac{m-1}{2}} \sim \chi_{m-1}^2$$

$$\left(\frac{\left(\frac{1}{1-2t}\right)^{m/2}}{\left(\frac{1}{1-2t}\right)^{1/2}} \right)$$

g) $\frac{z}{\sqrt{v/k}} \sim t_k$ com $z \sim N(0, 1)$ e $v \sim \chi_k^2$

$$J = \begin{vmatrix} \frac{\partial z}{\partial x} & \frac{\partial v}{\partial x} \\ \frac{\partial z}{\partial y} & \frac{\partial v}{\partial y} \end{vmatrix} = \begin{vmatrix} \sqrt{y/k} & 0 \\ 0 & 1 \end{vmatrix} = \det J = \sqrt{y/k}$$

$$f_{xy}(x, y) = (2\pi)^{-1/2} \exp\left\{-\frac{1}{2} \frac{v^2 y}{k}\right\} \frac{1}{\Gamma(k/2)} \left(\frac{1}{2}\right)^{k/2} y^{k/2-1} \exp\left\{-\frac{1}{2} y\right\} \left(\frac{y}{k}\right)^{1/2}$$

$$f_x = \int_{-\infty}^{\infty} f_{xy} dy$$

$$\frac{1}{\sqrt{2\pi} \cdot \Gamma(k/2)} \cdot \left(\frac{1}{2}\right)^{k/2} \int_0^{\infty} y^{k/2+1/2-1} \exp\left\{-\frac{1}{2} y \left[1 + \frac{v^2}{k}\right]\right\} dy \quad (A)$$

$$A = \frac{\Gamma\left(\frac{k+1}{2}\right) \cdot \left(\frac{2}{1+v^2/k}\right)^{\frac{k+1}{2}}}{\Gamma\left(\frac{k}{2}\right) \cdot \left(\frac{2}{1+v^2/k}\right)^{\frac{k}{2}}} \sim K. \text{Gamma}\left(\frac{k+1}{2}, \frac{2}{[1+v^2/k]}\right)$$

$$f_x(v) = \frac{\Gamma\left(\frac{k+1}{2}\right) \cdot \left(\frac{2}{1+v^2/k}\right)^{\frac{k+1}{2}}}{\sqrt{2\pi k} \Gamma(k/2)} \cdot \left(\frac{1}{2}\right)^{k/2} \frac{\Gamma(k/2)}{\Gamma(k/2)} \frac{1}{\sqrt{\pi k}} \frac{1}{(1+v^2/k)^{k/2}} \sim T_k$$