

# Etude loi Binomiale

•  $B(m, p)$

• Support  $D_x = \{0, 1, \dots, m\}$

$$P_x = \begin{cases} C_m^x p^x (1-p)^{m-x} & \text{si } x \in D_x \\ 0 & \text{ailleurs} \end{cases}$$

$$\begin{aligned} E[X] &= \sum_{x \in \mathbb{N}} x P_x(x) = \sum_{x \in D_x} C_m^x p^x (1-p)^{m-x} \cdot x \\ &= \sum_{x \in D_x} \frac{m!}{x!(m-x)!} \cdot p^x \cdot (1-p)^{m-x} \cdot x \\ &= \sum_{x=1}^m \frac{(m \cdot p) (m-1)!}{(x-1)!(m-x)!} \cdot p^{x-1} \cdot (1-p)^{m-x} \\ &= \sum_{x=0}^{m-1} \frac{(m \cdot p) (m-1)!}{x!(m-1-x)!} \cdot p^x \cdot (1-p)^{m-1-x} \\ &= m \cdot p \cdot (p + 1 - p)^{m-1} \\ &= mp \end{aligned}$$

$$\begin{aligned} E[X^2] &= \sum_{x=0}^m x^2 C_m^x p^x (1-p)^{m-x} \\ &= \sum_{x=0}^m x^2 \frac{m!}{x!(m-x)!} \cdot p^x \cdot (1-p)^{m-x} \\ &= \sum_{x=1}^m x \cdot \frac{(m-1)!}{(x-1)!(m-x)!} \cdot p^{x-1} (1-p)^{m-x} \cdot mp \\ &= mp \left( \sum_{x=0}^{m-1} (x+1) \frac{(m-1)!}{x!(m-1-x)!} \cdot p^x \cdot (1-p)^{m-1-x} \right) \\ &= mp \left( \sum_{x=0}^{m-1} x \frac{(m-1)!}{x!(m-1-x)!} \cdot p^x \cdot (1-p)^{m-1-x} + \sum_{x=0}^{m-1} \frac{(m-1)!}{x!(m-1-x)!} \cdot p^x \cdot (1-p)^{m-1-x} \right) \\ &= mp \left( (m-1)p + 1 \right) \end{aligned}$$

$$\Rightarrow \text{Var}[X] = E[X^2] - E[X]^2$$

$$= mp((n-1)p + 1) - (mp)^2$$

$$= mp(mp - p + 1) - m^2 p^2$$

$$= (mp)^2 + mp(1-p) - (mp)^2$$

$$= mp(1-p)$$

# Etude loi de Poisson

$$P[\lambda]$$

$$\text{Support } D_x : \mathbb{N}$$

$$P_x = \begin{cases} \frac{e^{-\lambda} \lambda^x}{x!} & \text{si } x \in D_x \\ 0 & \text{ailleurs} \end{cases}$$

$$\begin{aligned} E[X] &= \sum x P_x(x) = \sum_0^{+\infty} x \cdot \frac{e^{-\lambda} \lambda^x}{x!} \\ &= \sum_1^{+\infty} \frac{e^{-\lambda} \lambda^x}{(x-1)!} \\ &= \lambda \sum_0^{+\infty} \frac{e^{-\lambda} \lambda^x}{x!} \\ &= \lambda \end{aligned}$$

$$\begin{aligned} E[X^2] &= \sum x^2 P_x(x) = \sum_0^{+\infty} x^2 \frac{e^{-\lambda} \lambda^x}{x!} \\ &= \sum_1^{+\infty} x \cdot \frac{e^{-\lambda} \lambda^x}{(x-1)!} \\ &= \sum_0^{+\infty} (x+1) \cdot \frac{e^{-\lambda} \lambda^{x+1}}{x!} \\ &= \lambda \left( \sum_0^{+\infty} \frac{e^{-\lambda} \lambda^x}{x!} + \sum_0^{+\infty} \frac{x e^{-\lambda} \lambda^x}{x!} \right) \\ &= \lambda (1 + \lambda) \\ &= \lambda^2 + \lambda \end{aligned}$$

$$\Rightarrow \text{Var}[X] = E[X^2] - E[X]^2 = \lambda$$

## Approximation de la loi binomiale par poisson.

Soient  $\{p_1, p_2, \dots, p_n\}$  une suite de réels  
tq  $\forall i \quad ip_i = \lambda$ .

$$\Rightarrow \lim_{m \rightarrow +\infty} \binom{m}{x} p_m^x q^{m-x}$$

$$\lim_{m \rightarrow +\infty} \frac{m!}{x!(m-x)!} p_m^x \cdot \frac{m^x}{m^x} q^{m-x}$$

$$\lim_{m \rightarrow +\infty} \frac{1}{x!} \cdot \frac{m \cdot (m-1) \cdot \dots \cdot (m-x+1)}{m \cdot m \cdot \dots \cdot m} \binom{m}{x} p_m^x q^{m-x}$$

$$\lim_{m \rightarrow +\infty} \frac{1}{x!} \cdot \underbrace{\left[ \frac{m}{m} \cdot \frac{m-1}{m} \cdot \dots \cdot \frac{(m-x+1)}{m} \right]}_{x \text{ éléments}} \lambda^x \left(1 - \frac{m p_m}{m}\right)^{m-x}$$

$$\lim_{m \rightarrow +\infty} \frac{1}{x!} \cdot \lambda^x \left[ \frac{m}{m} \cdot \frac{(m-1)}{m} \cdot \dots \cdot \frac{(m-x+1)}{m} \right] \cdot \left(1 - \frac{\lambda}{m}\right)^{m-x}$$

$$* \lim_{m \rightarrow +\infty} \left[ \frac{m}{m} \cdot \frac{(m-1)}{m} \cdot \dots \cdot \frac{(m-x+1)}{m} \right] = 1$$

$$* \lim_{m \rightarrow +\infty} \left(1 - \frac{\lambda}{m}\right)^{m-x} = e^{-\lambda}$$

$$\Rightarrow \lim_{m \rightarrow +\infty} \binom{m}{x} p_m^x q_m^{m-x} = \frac{e^{-\lambda} \lambda^x}{x!}$$

# Etude loi Uniforme

$$\cdot U([a; b])$$

$$\cdot C_x = [a; b]$$

$$\cdot f_x = \begin{cases} \frac{1}{b-a} & \text{si } x \in C_x \\ 0 & \text{sinon} \end{cases}$$

$$\cdot F_x = \int_{-\infty}^x f_x(t) dt = \begin{cases} 0 & \text{si } x < a \\ \frac{x-a}{b-a} & \text{si } a \leq x \leq b \\ 1 & \text{sinon} \end{cases}$$

$$\cdot M_x(t) = E[e^{xt}] = \int_{-\infty}^{+\infty} e^{xt} \cdot f_x(t) dx$$

$$= \int_a^b e^{xt} \cdot \frac{1}{b-a} dx$$

$$= \frac{1}{b-a} \cdot \left[ \frac{e^{xt}}{t} \right]_a^b \quad t \neq 0$$

$$= \frac{e^{tb} - e^{ta}}{t(b-a)}$$

$$\cdot \text{si } t=0 \Rightarrow 1$$

$$M_x(t) = \begin{cases} \frac{e^{tb} - e^{ta}}{t(b-a)} & t \neq 0 \\ 1 & \text{sinon} \end{cases}$$

$$E[X] = \frac{dM_X(t)}{dt} \Big|_{t=0} = \frac{1}{(b-a)} \cdot \frac{(be^{tb} - ae^{ta})t - (e^{tb} - e^{ta})}{t^2}$$

$$= \frac{1}{b-a} \cdot \left( \frac{bte^{tb} - e^{tb}}{t^2} + \frac{(e^{ta} - ae^{ta})}{t^2} \right)$$

$$* e^{tb} \underset{v(t)}{\sim} 1 + tb + \frac{(tb)^2}{2}$$

$$\Rightarrow bte^{tb} - e^{tb} \underset{v(t)}{\sim} bt(1+tb) - 1 - tb - \frac{(tb)^2}{2}$$

$$\underset{v(t)}{\sim} \frac{t^2 b^2}{2} - 1$$

$$\Rightarrow e^{at} - ae^{at} \underset{v(t)}{\sim} 1 - \frac{t^2 a^2}{2}$$

$$\Rightarrow E[X] = \frac{1}{(b-a)} \cdot \frac{b^2 - a^2}{2} = \frac{a+b}{2}$$

$$E[X^2] = \frac{d^2 M_X(t)}{dt^2} \Big|_{t=0} = \frac{1}{b-a} \cdot \frac{b^2 x [b^2 t e^{tb} + b e^{tb} - b e^{tb} - a^2 t e^{at}] - 2 [b t e^{tb} - e^{tb} + e^{ta} - a t e^{ta}]}{t^3}$$

$$= \frac{1}{b-a} \cdot \frac{e^{tb} (b^2 t^2 - 2bt + 2) - e^{ta} (a^2 t^2 - 2at + 2)}{t^3}$$

$$* e^{tb} \underset{v(t)}{\sim} 1 + tb + \frac{(tb)^2}{2} + \frac{(tb)^3}{3!} \Rightarrow e^{tb} (b^2 t^2 - 2bt + 2) \underset{v(t)}{\sim} \frac{b^2 t^2 + b^3 t^3 - 2bt + 2}{v(t)} \sim \frac{b^2 t^2 + b^3 t^3 - 2bt + 2}{3} \sim \frac{t^3 b^3}{3}$$

$$* e^{ta} \underset{v(t)}{\sim} \frac{t^3 a^3}{3}$$

$$\Rightarrow E[X^2] = \frac{a^2 + ab + b^2}{3}$$

$$Var[X] = \frac{(b-a)^2}{12}$$

# Loi exponentielle

• Exp( $\theta, v$ )

•  $G_x = ]v; +\infty[$

$$f_x = \begin{cases} \theta e^{-\theta(t-v)} & \text{si } x \in G_x \\ 0 & \text{ailleurs} \end{cases}$$

$$F_x(x) = \begin{cases} 1 - e^{-\theta(t-v)} & \text{si } x \in G_x \\ 0 & \text{sinon} \end{cases}$$

$$M_x(t) = E[e^{xt}] = \int_{-\infty}^{+\infty} e^{xt} \cdot f_x(x) dx$$

$$= \int_v^{+\infty} e^{xt} \cdot \theta e^{-\theta(x-v)} dx$$

$$= \theta e^{\theta v} \int_v^{+\infty} e^{-x(\theta-t)} dx$$

$$= \theta e^{\theta v} \left[ \frac{e^{-x(\theta-t)}}{\theta-t} \right]_v^{+\infty}$$

$$= \theta e^{\theta v} \cdot \frac{e^{-v(\theta-t)}}{\theta-t}$$

$$= \frac{\theta \cdot e^{\theta v t}}{\theta-t}$$

$$E[X] = \left. \frac{dM_X(t)}{dt} \right|_{t=0} = \frac{\theta v e^{tv} (\theta - t) + \theta e^{tv}}{(\theta - t)^2}$$

$$\stackrel{(0)}{=} \frac{\theta^2 v + \theta}{\theta^2} \stackrel{(0)}{=} v + \frac{1}{\theta}$$

$$E[X^2] = \left. \frac{d^2 M_X(t)}{dt^2} \right|_{t=0} = \frac{e^{tv} (\theta^3 v^2 - \theta v^2 t + \theta v - \theta^2 v) (\theta - t) + 2e^{tv} (\theta^2 v - \theta v t + \theta)}{(\theta - t)^3}$$

$$\stackrel{(0)}{=} \frac{\theta^3 v^2 + 2\theta^2 v + \theta}{\theta^3}$$

$$\Rightarrow \text{Var}[X] = E[X^2] - E[X]^2$$

$$= \left( v^2 + \frac{1}{\theta^2} + \frac{v}{\theta} \right) - \left( v + \frac{1}{\theta} \right)^2$$

$$= v^2 + \frac{2}{\theta^2} + \frac{2v}{\theta} - v^2 - \frac{1}{\theta^2} - 2\frac{v}{\theta}$$

$$= \frac{1}{\theta^2}$$



# Loi Normale

$$\mathcal{N}(\mu, \sigma^2)$$

$$C_X = \mathbb{R}$$

$$f_X(x) = \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

$$F_X(x) = \int_{-\infty}^x f_X(t) dt$$

$$M_X(t) = E[e^{xt}] = \int_{-\infty}^{+\infty} e^{xt} \cdot \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}} dx$$
$$= \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{xt} \cdot e^{-\frac{(x-\mu)^2}{2\sigma^2}} \cdot \frac{1}{2} dx$$

$$\text{On pose } X = \frac{x-\mu}{\sigma} \Rightarrow dX = \frac{dx}{\sigma}$$

$$\Rightarrow M_X(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{(\sigma X + \mu)t - \frac{X^2}{2}} dX$$

$$= \frac{e^{\mu t}}{\sqrt{2\pi}} \int_{-\infty}^{+\infty} e^{\sigma t X - \frac{X^2}{2} + \frac{(\sigma t)^2}{2} - \frac{(\sigma t)^2}{2}} dX$$

$$= \frac{e^{\mu t}}{\sqrt{2\pi}} \cdot e^{\frac{(\sigma t)^2}{2}} \int_{-\infty}^{+\infty} e^{-\frac{(X - \sigma t)^2}{2}} dX$$

$$= e^{\frac{t\mu + (\sigma t)^2}{2}}$$

$$\bullet E[X] = \left. \frac{dM_X(t)}{dt} \right|_{t=0} = \mu$$

$$\begin{aligned} \bullet E[X^2] &= \frac{dM_X(t)}{dt} = \left( e^{\mu t + \frac{\sigma^2 t^2}{2}} \right)'' \\ &= \left( (\mu + t\sigma^2) e^{\mu t + \frac{\sigma^2 t^2}{2}} \right)' \\ &= e^{\frac{\mu t + \sigma^2 t^2}{2}} (\sigma^2 [\mu^2 + t\sigma^2\mu + t\sigma^2 + t^2\sigma^3]) \\ &= \sigma^2 + \mu^2 \end{aligned}$$

$$\Rightarrow \text{Var}[X] = \sigma^2$$

# Loi Lognormale

Rappel: •  $X$  v.a.c. de support  $C_X$ .  
•  $\phi: \mathbb{R} \rightarrow \mathbb{R}$  monotone telle que

$$Y = \phi(X)$$

• On note  $\psi = \phi^{-1}$

Alors  $Y$  est v.a.c de support  $C_Y = \phi(C_X)$   
et de fct de densité:

$$f_Y(y) = \begin{cases} f_X(\psi(y)) \cdot |\psi'(y)| & \text{si } y \in C_Y \\ 0 & \text{sinon} \end{cases}$$

On pose  $\begin{cases} \phi(x) = e^x; & \psi'(x) = e^x > 0 \text{ monotone} \\ \psi(y) = \ln y; & \psi'(y) = \frac{1}{y} \end{cases}$

$C_Y = ]0; +\infty[$

$$f_Y(y) = \begin{cases} \frac{1}{y} \cdot \frac{1}{\sigma\sqrt{2\pi}} \cdot e^{-\frac{(\ln y - \mu)^2}{\sigma^2 \cdot 2}} & \text{si } y > 0 \\ 0 & \text{sinon} \end{cases}$$

$$\bullet F_Y(y) = \int_{-\infty}^y f_Y(x) dx = \int_0^y \frac{1}{x\sigma\sqrt{2\pi}} \cdot e^{-\frac{(\ln x - \mu)^2}{2\sigma^2}} dx$$

$$X = \frac{\ln x - \mu}{\sigma} \Rightarrow dx = \sigma x dX$$

$$\Rightarrow F_Y(y) = \int_{-\infty}^{\frac{\ln y - \mu}{\sigma}} \frac{1}{\sqrt{2\pi}} \cdot e^{-\frac{X^2}{2}} dX$$

$$= F_{\mathcal{N}(0,1)}\left(\frac{\ln y - \mu}{\sigma}\right)$$

avec  $F_{\mathcal{N}(0,1)}$  la fct<sup>e</sup> de répartition de la loi normale centrée réduite.

$$\bullet \text{ On a } X : \mathcal{N}(\mu_x, \sigma_x^2) \Rightarrow M_X(t) = e^{\mu_x t + \frac{\sigma_x^2 t^2}{2}}$$

$$\rightsquigarrow E[Y] = E[e^{\ln Y}] = E[e^X] = M_X(1)$$

$$\Rightarrow E[Y] = e^{\mu_x + \frac{\sigma_x^2}{2}}$$

$$\rightsquigarrow E[Y^2] = E[e^{2 \ln Y}] = E[e^{2X}] = M_X(2)$$

$$\Rightarrow E[Y^2] = e^{2\mu_x + 2\sigma_x^2}$$

$$\rightsquigarrow \text{Var}[Y] = E[Y^2] - E[Y]^2$$

$$= e^{2\mu_x + 2\sigma_x^2} - e^{2\mu_x} e^{\sigma_x^2}$$

$$= e^{2\mu_x + \sigma_x^2} (e^{\sigma_x^2} - 1)$$