

Variable	sucesos	Modelo	Función de probabilidad o densidad	Función acumulativa o de distribución	Esperanza	Varianza	Parámetros
Disc.	Única obs.	Bernoulli	$f(x) = \begin{cases} p, \forall x = 1 \\ 1 - p, \forall x = 0 \end{cases}$		$E(x) = p$	$Var(x) = p.q$	p $q = 1 - p$
Disc.	Indep.	Binomial	$f(X = x) = \frac{n!}{x!(n-x)!} p^x (1-p)^{n-x}$	$F(x) = \sum_{\forall x_i \leq x} f(x_i)$	$E(x) = n.p$	$Var(x) = n.p.q$	n p
Disc.	Indep.	Geométrico	$P(N = n) = f(n) = (1-p)^{n-1} p$	$F(x) = 1 - (1-p)^n$	$E(x) = \frac{1}{p}$	$Var(x) = \frac{1-p}{p^2}$	n p
Disc.	Indep.	Poisson	$P(X = x) = f(x) = \frac{e^{-\lambda} \lambda^x}{x!}$		$E(x) = \lambda = n.p$	$Var(x) = \lambda = n.p$	$\lambda = n.p$
Disc.	Dep.	Hipergeométrico	$P(X = x) = f(x) = \frac{\binom{k}{x} \binom{N-k}{n-x}}{\binom{N}{n}}$		$E(x) = n \left(\frac{k}{N} \right)$	$Var(x) = \frac{nk(N-k)}{N^2} \left(\frac{N-n}{N-1} \right)$	k N n
Cont.	Indep.	Exponencial	$f(t) = \lambda e^{-\lambda t}$	$F(t) = 1 - e^{-\lambda t}$	$E(T) = \frac{1}{\lambda}$	$Var(T) = \frac{1}{\lambda^2}$	$\lambda = n.p$ t
Cont.		Gamma	$f(x) = \frac{\lambda(\lambda x)^{k-1} e^{-\lambda x}}{(k-1)!}$	$F(x) = \int_0^x f(t) dt = \frac{\Gamma(k, \lambda x)}{\Gamma(k)} = \frac{\text{función gamma incompleta}}{\text{función gamma}}$	$E(x) = \frac{k}{\lambda}$	$Var(x) = \frac{k}{\lambda^2}$	$\lambda = n.p$ k
Cont.		Normal	$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$	$F(x) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^x e^{-\frac{1}{2}\left(\frac{t-\mu}{\sigma}\right)^2} dt$	$E(x) = \mu$	$Var(x) = \sigma^2$	μ σ
Cont.		Normal estándar	$f(z) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}z^2}$	$F(x) = \frac{1}{\sigma\sqrt{2\pi}} \int_{-\infty}^z e^{-\frac{1}{2}t^2} dt$	$E(z) = 0$	$Var(z) = 1$	$z = \frac{x-\mu}{\sigma}$ $\mu = 0$ $\sigma = 1$
Cont.		Chi-cuadrado	$f(\chi^2) = \begin{cases} \frac{1}{2^{\frac{v}{2}} \Gamma\left(\frac{v}{2}\right)} \chi^{2\left(\frac{v-2}{2}\right)} e^{-\frac{\chi^2}{2}}, & \text{para } \chi^2 > 0 \\ 0, & \text{en otro caso} \end{cases}$	$V = \sum_{i=1}^v \left(\frac{X_i - \mu_i}{\sigma_i} \right)^2 \square \chi_v^2$	$E(\chi^2) = v$	$Var(\chi^2) = 2v$	v χ^2

Cont.		T de Student	$f(t) = \frac{1}{\sqrt{\pi v}} \frac{\Gamma\left(\frac{v+1}{2}\right)}{\Gamma\left(\frac{v}{2}\right)} \left(1 + \frac{t^2}{v}\right)^{-\frac{1}{2}v-1}$ <p>Con $t_v = \frac{x}{\sqrt{\frac{V}{v}}} \left\{ \begin{array}{l} x \sim N(0,1) \\ V \sim \chi^2_v \end{array} \right.$</p>		$E(t) = 0$	$Var(t) = \frac{v}{v-2}, \quad v > 2$	$\begin{matrix} v \\ x \\ V \end{matrix}$
Cont.		F de Snedecor	$f(F) = \frac{\Gamma\left(\frac{v_1+v_2}{2}\right) v_1^{\frac{v_1}{2}} v_2^{\frac{v_2}{2}} F^{\frac{v_1-1}{2}}}{\Gamma\left(\frac{v_1}{2}\right) \Gamma\left(\frac{v_2}{2}\right) (v_2 + v_1 F)^{\frac{v_1+v_2}{2}}}, \quad \text{para } F > 0$	$F_{v_1, v_2} = \frac{\frac{\chi^2_{v_1}}{v_1}}{\frac{\chi^2_{v_2}}{v_2}}$			
Cont.		Log-Normal	$f(x) = \frac{1}{x\sigma_y\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{\ln x - \mu_y}{\sigma_y}\right)^2}, \quad x \geq 0$		$E(x) = e^{\left(\mu_y + \frac{\sigma_y^2}{2}\right)}$	$Var(x) = \mu_x^2 \left(e^{\sigma_y^2} - 1\right)$	$\begin{matrix} \mu_x \\ \mu_y \\ \sigma_y \end{matrix}$
Cont.	Indep.	Gumbel	$f(x) = \alpha e^{-\alpha(x-\mu_0)} e^{-e^{-\alpha(x-\mu_0)}}$	$F(x) = e^{-e^{-\alpha(x-\mu_0)}}, \quad -\infty \leq x \leq +\infty$	$E(x) = \mu = \mu_0 + \frac{0.577}{\alpha}$	$Var(x) = \sigma^2 = \frac{1.645}{\alpha^2}$	$\begin{matrix} \alpha \\ \mu_0 = x \end{matrix}$
Cont.	Indep.	Weibull	$f(z) = \frac{k}{\mu_0 - \varepsilon} \left(\frac{z - \varepsilon}{\mu_0 - \varepsilon}\right)^{k-1} e^{-\left(\frac{z - \varepsilon}{\mu_0 - \varepsilon}\right)^k}$	$F(z) = 1 - e^{-\left(\frac{z - \varepsilon}{\mu_0 - \varepsilon}\right)^k}$	$E(x) = \mu = \varepsilon + (\mu_0 - \varepsilon) \Gamma\left(1 + \frac{1}{k}\right)$	$Var(x) = \sigma^2 = (\mu_0 - \varepsilon)^2 \left[\Gamma\left(1 + \frac{2}{k}\right) - \Gamma^2\left(1 + \frac{1}{k}\right) \right]$	