

C6: Discrete Distributions

	Poisson	Binomial	N-Binomial	Geometric
PMF	$p_k = e^{-\lambda} \frac{\lambda^k}{k!}$	$p_k = \binom{m}{k} q^k (1-q)^{m-k}$	$p_k = \binom{k+r-1}{k} \left(\frac{\beta}{1+\beta}\right)^k \left(\frac{1}{1+\beta}\right)^r$ $k = 0, 1, 2, \dots, r > 0, \beta > 0$	$p_k = \left(\frac{\beta}{1+\beta}\right)^k \left(\frac{1}{1+\beta}\right)$ $k = 0, 1, 2, \dots, \beta > 0$
Moments	$E[X] = \lambda$ $Var(X) = \lambda$	$E[X] = mq$ $Var(X) = q(1-q)$	$E[X] = r\beta$ $Var(X) = r\beta(1+\beta)$	$E[X] = \beta$ $Var(X) = \beta(1+\beta)$
PMF/MGF	$G_Z = e^{\lambda(z-1)}$ $M_X = e^{\lambda(e^t-1)}$	$G_Z = (1-q+qz)^m$ $M_X = (1-q+qe^t)^m$	$G_Z = (1-\beta(z-1))^{-r}$ $M_X = (1-\beta(e^t-1))^{-r}$	$G_Z = (1-\beta(z-1))^{-1}$ $M_X = (1-\beta(e^t-1))^{-1}$
Likelihood Functions:				
$\lg \ell = \sum_{i=1}^{\infty} n_i \lg p(x_i \theta)$	x_i	Observed datum		
	$n_p, i = 1, 2, \dots, k$	Number of times x_i is observed. $n_i = 1$ for point data.		
	k	Number of groups		
Grouped Data $\lg l'(\theta) = 0$	$\frac{\partial \lg \ell}{\partial \lambda} = -n + \frac{n\bar{x}}{\lambda}$	$\frac{\partial \lg \ell}{\partial q} = \frac{n\bar{x}}{q} - \frac{mn - n\bar{x}}{1-q}$ $n = \sum_{i=0}^k n_i$	$\frac{\partial \lg \ell}{\partial \beta} = \sum_{k=0}^{\infty} n_k \left(\frac{k}{\beta} + \frac{r-k}{1+\beta} \right)$ $\frac{\partial \lg \ell}{\partial r} = -n_k \ln(1+\beta) + \sum_{k=i}^{\infty} n_k \sum_{m=0}^{k-1} \frac{1}{r+m}$	
MLEs	$\hat{\lambda} = \bar{X}$	$\hat{q} = \frac{1}{m} \frac{\sum_{k=0}^{\infty} kn_k}{\sum_{k=0}^{\infty} n_k}$	$\hat{\beta} = \frac{\bar{X}}{\hat{r}}$ r must be estimated numerically: (Newton – Raphson) $r_k = r_{k-1} - \frac{H(r_{k-1})}{H'(r_{k-1})}$ $H(r_k) = n \lg \left(1 + \frac{\bar{x}}{r} \right) - \sum_{k=1}^{\infty} n_k \left(\sum_{m=0}^{k-1} \frac{1}{r+m} \right) = 0$	$\hat{\beta} = \bar{X}$

	p_0	a	b	Parameter Space
Po (a, b, 0)	$e^{-\lambda}$			
ZT Po	0	0	λ	$\lambda > 0$
ZM Po	Arbitrary			
Bin	$(1-q)^m$			
ZT Bin	0	$-\frac{q}{1-q}$	$(m+1)\frac{q}{1-q}$	$0 < q < 1$ $m = 1, 2, \dots$
ZM Bin	Arbitrary			
NB	$(1+\beta)^{-r}$			$r > 0, \beta > 0$
ZT NB	0	$\frac{\beta}{1+\beta}$	$(r-1)\frac{\beta}{1+\beta}$	$r > -1, r \neq 0, \beta > 0$
ZM NB	Arbitrary			$r > -1, r \neq 0, \beta > 0$
Geo	$(1+\beta)^{-1}$			
ZT Geo	0	$\frac{\beta}{1+\beta}$	0	$\beta > 0$
ZM Geo	Arbitrary			
Log	0	$\frac{\beta}{1+\beta}$	$-\frac{\beta}{1+\beta}$	$\beta > 0$
ZM Log	Arbitrary			

(a, b, 0) Class	(a, b, 1) Class
$p_k = \left(a + \frac{b}{k}\right) p_{k-1}, k = 1, 2, 3$ Estimation : $k \frac{\hat{p}_k}{\hat{p}_{k-1}} = k \frac{n_k}{n_{k-1}}$	$\frac{p_k}{p_{k-1}} = a + \frac{b}{k}, k = 2, 3, 4, \dots$ <ul style="list-style-type: none"> Zero-Truncated: $p_0^T = 0$ (no observed data at zero) Zero-Modified: $p_0^M > 0$ (Unusually large, or poor fit, number of observations at zero)
ZM Estimation	ZT Estimation
For some arbitrary constant c , $p_k^M = c p_k$ $G_Z^M = p_0^M + \sum_k p_k^M z^k$ $G_Z^M(1) = G_Z(1) = 1$ $\therefore c = \frac{1 - p_0^M}{1 - p_0}$	Set : $p_0^M = 0$ $p_z^T = \frac{p_z - p_0}{1 - p_0}$ $p_z^T = \frac{p_k}{1 - p_0}$ $p_z^M = (1 - p_z^M) p_z^T$

Other on Poisson
Poisson as a limiting case of NB $\beta = \frac{\lambda}{r} \Rightarrow \lim_{r \rightarrow \infty} \left[1 - \frac{\lambda(z-1)}{r} \right]^{-r} = \dots = e^{\lambda(z-1)}$ $\Rightarrow \lim_{r \rightarrow \infty} NB\left(r, \frac{r}{\lambda+r}\right) = Po(\lambda)$