

Stat 2911 Lecture Notes

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Poisson approximation to the
binomial distribution, faulty
monitor, sum of independent RVs
(convolution): binomial, Poisson
RVs

The Poisson Dist. $P_{\bar{X}}(k) = e^{-\lambda} \frac{\lambda^k}{k!} \quad k \in \mathbb{N} \quad \lambda > 0$

Models the # of events in a given time interval, assuming:

- 1) The dist. of the number of events at any time interval depends only on the interval's length/duration.
- 2) The # of events recorded in two disjoint time intervals are ind. of one another.
- 3) No two events occur at the same time point.

Let $\underline{X} = \underline{X}_t = \# \text{ of events in } (0, t]$.

We showed $P(\underline{X}=0) = e^{-\lambda} = e^{-\lambda} \frac{\lambda^0}{0!}$ by showing

$$P(\underline{X}_t=0) = e^{-\lambda t} \text{ for some } \lambda > 0.$$

To study $P(\underline{X}=j)$ for $j \geq 1$ we look at

$Y_n = \# \text{ of intervals } (\frac{k-1}{n}, \frac{k}{n}] \text{ in } (0, 1] \text{ in which an event occurred}$

$Y_n \sim \text{Binomial}(n, p_n = 1 - e^{-\lambda/n})$ with $\lambda > 0$

$Y_n \leq \underline{X}_t$ but $\lim_{n \rightarrow \infty} Y_n(\omega) = \underline{X}_t(\omega)$

intuitively

$$\Rightarrow P(Y_n=k) \xrightarrow{n} P(\underline{X}=k)$$

Note that $Y_n(\omega) \rightarrow \underline{X}_t(\omega)$ even if $\underline{X}_t(\omega) = +\infty$ and that this analysis shows $P(\underline{X}=+\infty)=0$.

The expected value of a $\text{Binomial}(n, p)$ RV is np .

For Y_n this is $np_n = n(1 - e^{-\lambda_n})$. (expected # of "intervals" with ≥ 1 events)

What is $\lim np_n$?

First order Taylor series expansion of e^x :

$$e^x = 1 + x + R_1(x) \text{ where } R_1(x)/x \xrightarrow{x \rightarrow 0} 0$$

$$\begin{aligned} np_n &= n(1 - e^{-\lambda_n}) \\ &= n \left[1 - \left(1 - \lambda_n + R_1(-\lambda_n) \right) \right] \\ &= \lambda + \frac{R_1(-\lambda_n)}{-\lambda_n} \cdot \lambda \xrightarrow{n \rightarrow \infty} \end{aligned} \quad (x = -\lambda_n)$$

Claim. If $Y_n \sim \text{Binom}(n, p_n)$ s.t. $np_n \xrightarrow{n \rightarrow \infty} \lambda > 0$, then for any fixed $k \in \mathbb{N}$,

$$P(Y_n = k) \xrightarrow{n \rightarrow \infty} e^{-\lambda} \frac{\lambda^k}{k!}.$$

Cor. $\bar{X} = \sum_i n \text{Poisson}(\lambda)$: $P_{\bar{X}}(k) = e^{-\lambda} \frac{\lambda^k}{k!} \quad k \in \mathbb{N}$.

Proof of Cor.

$$\text{If } k \in \mathbb{N}, \quad P(\bar{X}_i = k) = \lim_n P(Y_n = k) = e^{-\lambda} \frac{\lambda^k}{k!}.$$

Proof of Claim. $P(Y_n = k) = \overbrace{\binom{n}{k} p_n^k}^I \overbrace{(1-p_n)^{n-k}}^{II}$ $k=0, 1, \dots, n$

$$(I) \quad \binom{n}{k} p_n^k = \frac{n(n-1)\dots(n-k+1)}{k!} p_n^k \\ = \frac{(np_n)^k}{k!} \frac{n}{n} \frac{n-1}{n} \dots \frac{n-k+1}{n} \xrightarrow[n \rightarrow \infty]{} 1$$

$$(II) \quad (1-p_n)^{n-k} = (1-p_n)^n (1-p_n)^{-k}$$

Because $np_n \xrightarrow{n \rightarrow \infty} \lambda \in \mathbb{R}$ it follows that $p_n \rightarrow 0$.

$$\text{Hence } (1-p_n)^{-k} \xrightarrow{n \rightarrow \infty} 1$$

$$(1-p_n)^n = e^{n \log(1-p_n)}$$

Taylor expansion: $\log(1-x) = -x + R_1(x)$ where $\frac{R_1(x)}{x} \xrightarrow{x \rightarrow 0} 0$

$$\Rightarrow n \log(1-p_n) = n[-p_n + R_1(p_n)] \\ = -np_n + \frac{R_1(p_n)}{p_n} \cdot (np_n) \xrightarrow{n \rightarrow \infty} 0$$

$$\Rightarrow \lim_n (1-p_n)^n = \lim_n e^{n \log(1-p_n)} \\ = e^{\lim_n n \log(1-p_n)}$$

$$= e^{-\lambda}$$

$$\Rightarrow \lim_n P(Y_n = k) = e^{-\lambda} \frac{\lambda^k}{k!} \quad \square$$

Example. Poisson approx. to the binomial dist.!

$$n=100, p=0.01 \quad \lambda = np = 1$$

$$Y_n \sim \text{Binom}(n, p) \quad P(Y_n=3) \approx 0.0610$$

$$\bar{X}_n \sim \text{Poisson}(\lambda) \quad P(\bar{X}_n=3) \approx 0.0613$$

Going back to Bartkiewicz:

of horse kick fatalities in 1 corps year

20 corps \times 10 years = 200 Corp years

# of deaths	count	frequency	Poisson approx.
0	109	0.545	0.543
1	65	0.325	0.331
2	22	0.110	0.101
3	3	0.015	0.021
4	1	0.005	0.003

What is missing?

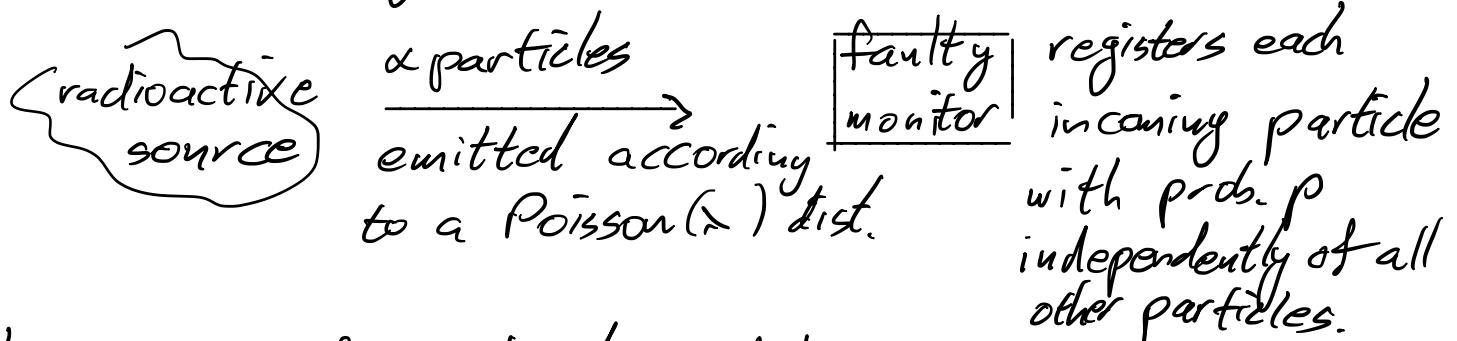
How was λ determined?

$$\begin{aligned} \text{Empirical mean} &= \frac{109}{200} \times 0 + \frac{65}{200} \times 1 + \frac{22}{200} \times 2 + \frac{3}{200} \times 3 + \frac{1}{200} \times 4 \\ &= 0.61 \end{aligned}$$

For a Poisson RV \bar{X} , $E(\bar{X}) = \lambda \Rightarrow \hat{\lambda} = 0.61$.

More on that later!

Example . Faulty monitor (Rice p-88)



Let $\bar{X} = \#$ of registered particles in an hour.

What is the dist. of \bar{X} ?

Let $N = \#$ of emitted particles in the same hour (unobserved).

For $k \in \mathbb{N}$,

$$\begin{aligned}
 P(\bar{X}=k) &= \sum_{n=0}^{\infty} P(\bar{X}=k | N=n) P(N=n) \\
 &= \sum_{n=k}^{\infty} \binom{n}{k} p^k (1-p)^{n-k} e^{-\lambda} \frac{\lambda^n}{n!} \\
 &= \frac{e^{-\lambda} (\lambda p)^k}{k!} \sum_{n=k}^{\infty} \frac{\lambda^{n-k} (1-p)^{n-k}}{(n-k)!} \\
 &\stackrel{m=n-k}{=} \frac{e^{-\lambda} (\lambda p)^k}{k!} \sum_{m=0}^{\infty} \frac{[\lambda (1-p)]^m}{m!} \\
 &= \frac{e^{-\lambda} (\lambda p)^k}{k!} e^{\lambda (1-p)} = e^{-\lambda p} \frac{(\lambda p)^k}{k!}
 \end{aligned}$$

$\Rightarrow \bar{X} \sim \text{Poisson}(\lambda p)$

It makes sense: our three assumptions still hold!
Moreover, recalling the expected # of counts is the Poisson parameter, it makes sense that $\lambda \rightarrow \lambda p$.

Dist. of a sum of RVs (arithmetic of RVs)

If X and Y are jointly-distributed (defined on the same sample space) N -valued RVs then $Z = X + Y$ is also an N -valued RV, and for $n \in N$,

$$\begin{aligned} p_Z(n) &= P(Z=n) \\ &= P(X+Y=n) \\ &= \sum_{k=0}^n P(X=k, Y=n-k) \\ &= \sum_{k=0}^n p_{XY}(k, n-k) \end{aligned}$$

If X and Y are ind., then

$$p_Z(n) = \sum_{k=0}^n p_X(k) p_Y(n-k) \quad \forall n \in N$$

p_Z is the convolution of p_X and p_Y denoted as

$$p_Z \equiv p_X * p_Y .$$

Examples

1) $X \sim \text{binomial}(m, p)$ is indep. of $Y \sim \text{binomial}(n-m, p)$.

Let $Z = X+Y$.

It is clear that $Z \sim$.

Formally, for $\ell \in \{0, 1, \dots, n\}$

$$\begin{aligned} P(Z=\ell) &= \sum_{k=0}^{\ell} \binom{m}{k} p^k (1-p)^{m-k} \binom{n-m}{\ell-k} p^{\ell-k} (1-p)^{n-m-(\ell-k)} \\ &= p^\ell (1-p)^{n-\ell} \sum_{k=0}^{\ell} \binom{m}{k} \binom{n-m}{\ell-k} \end{aligned}$$

Claim. $\sum_{k=0}^{\ell} \binom{m}{k} \binom{n-m}{\ell-k} = \binom{n}{\ell}$

Cor. $P(Z=\ell) = \binom{n}{\ell} p^\ell (1-p)^{n-\ell}$ or, $Z \sim \text{binom}(n, p)$

Proof (of Claim).

(I) Combinatorics : count the same object twice

$\binom{n}{\ell}$ = # of committees of ℓ students
from a class of n

Suppose the class has m boys and $n-m$ girls, then
the same number can be counted by

$$\sum_{k=0}^{\ell} \# \text{ of the committees with } k \text{ boys} = \sum_{k=0}^{\ell} \binom{m}{k} \binom{n-m}{\ell-k}$$

$$(II) \text{ Algebra: } (1+x)^n = (1+x)^m (1+x)^{n-m}$$

Consider the coefficient of x^l on both sides

$$\text{LHS: } \binom{n}{l} \quad (\text{binomial theorem})$$

$$\begin{aligned} \text{RHS: } & \sum_{k=0}^l \left[\text{coeff of } x^k \text{ in } (1+x)^m \right] \cdot \left[\text{coeff of } x^{l-k} \text{ in } (1+x)^{n-m} \right] \\ &= \sum_{k=0}^l \binom{m}{k} \binom{n-m}{l-k} \end{aligned}$$

2) $\bar{X} \sim \text{Poisson}(\lambda)$ ind. of $\bar{Y} \sim \text{Poisson}(\delta)$.

What is the dist. of $Z = \bar{X} + \bar{Y}$? For $n \in \mathbb{N}$,

$$\begin{aligned} P(Z=n) &= \sum_{k=0}^n P_{\bar{X}}(k) P_{\bar{Y}}(n-k) \\ &= \sum_{k=0}^n e^{-\lambda} \frac{\lambda^k}{k!} e^{-\delta} \frac{\delta^{n-k}}{(n-k)!} \\ &= e^{-(\lambda+\delta)} \frac{1}{n!} \sum_{k=0}^n \binom{n}{k} \lambda^k \delta^{n-k} \\ &= e^{-(\lambda+\delta)} (\lambda+\delta)^n / n! \end{aligned}$$

$$\Rightarrow Z \sim \text{Pois}(\lambda+\delta)$$

Imagine two ind. radioactive sources,