0.1.1 Frequency Distributions

Frequency Distributions
Poisson
Binomial
Geometric
Negative Binomial
Exercises

## Frequency and Severity

In describing losses, we often have 3 types of random variables.

- Frequency or number of losses. This is always a discrete random variable
- Severity or amount of loss / payment. This tends to be continuous, but can be discrete, continuous, or mixed.
- Aggregate or total loss comes from summing a (Frequency) number of Severity variables. This is discrete if severity is discrete, and continuous (except at 0 ) if severity is continuous.

We will focus on the most common frequency distributions in this lesson.
$N$ is a discrete random variable if it has a probability mass function $p_{k}$,

$$
\begin{aligned}
p_{k} & =\mathrm{P}[N=k] \\
0 & \leq p_{k} \leq 1 \\
\sum_{k} p_{k} & =1
\end{aligned}
$$

$p_{k}$ determines everything we care about. E.g.

$$
\begin{aligned}
\mathrm{E}[N] & =\sum_{k} k \cdot p_{k} \\
\mathrm{E}[g(N)] & =\sum_{k} g(k) \cdot p_{k}
\end{aligned}
$$

For us, discrete random variables will usually be counting / frequency variables, meaning the possible values are $\{0,1,2, \ldots\}$
Some key discrete distributions are included in tables at the exam.

## Poisson

$N \sim \operatorname{Poisson}(\lambda)$ if

$$
\mathrm{P}[N=n]=e^{-\lambda} \cdot \frac{\lambda^{n}}{n!} \quad n=0,1,2, \ldots
$$

Recall that the Taylor series for $e^{\lambda}$ is

$$
e^{\lambda}=1+\lambda+\frac{\lambda^{2}}{2!}+\cdots+\frac{\lambda^{n}}{n!}+\cdots
$$

The $e^{-\lambda}$ term makes the probabilities to sum to 1 .

$$
\mathrm{E}[N]=\lambda=\operatorname{Var}[N]
$$

If $N_{1}$ and $N_{2}$ are independent Poisson random variables then $N_{1}+N_{2} \sim \operatorname{Poisson}\left(\lambda_{1}+\lambda_{2}\right)$

## Binomial

$N$ is binomial if we have

- $m$ independent trials
- Each trial has probability $q$ of a loss
- $N=$ total number of losses

$$
\begin{aligned}
\mathrm{P}[N=n] & =\binom{m}{n} q^{n}(1-q)^{m-n} \\
\mathrm{E}[N] & =m q \\
\operatorname{Var}[N] & =m q(1-q)<\mathrm{E}[N]
\end{aligned}
$$

If $N_{1}$ and $N_{2}$ are independent binomial random variables with the same $q$ then $N_{1}+N_{2} \sim \operatorname{Bin}\left(m_{1}+m_{2}, q\right)$.

If they have different $q$ 's then the sum is gross.
If $m=1$ we have a Bernoulli random variable

## Geometric

One way to get a geometric $N$ with mean/parameter $\beta$ is if:

- Each year independently has a loss with probability $1 /(1+\beta)$
- $N=\#$ of years before the first loss

$$
\begin{aligned}
\mathrm{P}[N=0] & =\frac{1}{1+\beta} \quad \mathrm{P}[N=k]=\left(\frac{\beta}{1+\beta}\right)^{k} \cdot \frac{1}{1+\beta} \\
\mathrm{P}[N \geq k] & =\left(\frac{\beta}{1+\beta}\right)^{k} \\
\mathrm{E}[N] & =\beta \\
\operatorname{Var}[N] & =\beta(1+\beta)>\mathrm{E}[N]
\end{aligned}
$$

The geometric (like the exponential) is memoryless:

- $\mathrm{P}[N=d+n \mid N \geq d]=\mathrm{P}[N=n]$
- $\mathrm{E}[N-d \mid N \geq d]=\beta$

If $N_{1}, N_{2}, \ldots, N_{r}$ are iid geometric $(\beta)$, then $N=\sum N_{i}$ is a negative binomial $(r, \beta)$.
$N=\#$ years with no loss before $r$-th year with a loss.

$$
\begin{aligned}
\mathrm{P}[N=0] & =\frac{1}{(1+\beta)^{r}} \\
\mathrm{P}[N=k] & =\binom{r+k-1}{r-1} \cdot\left(\frac{1}{1+\beta}\right)^{r} \cdot\left(\frac{\beta}{1+\beta}\right)^{k} \\
& =\frac{r(r+1) \ldots(r+k-1)}{k!} \cdot \frac{\beta^{k}}{(1+\beta)^{r+k}} \\
\mathrm{E}[N] & =r \beta \quad \operatorname{Var}[N]=r \beta(1+\beta)
\end{aligned}
$$

These last 2 lines hold even if $r$ is not an integer.
The coefficient in $\mathrm{P}[N=k]$ has $k$ terms in numerator and denominator.
A geometric is a negative binomial with $r=1$.

## Comparisons

The mean and variances of these distributions come up often enough that you should memorize them.

| Distribution | Mean |  | Variance |
| :---: | :---: | :---: | :---: |
| Binomial | $m q$ | $>$ | $m q(1-q)$ |
| Poisson | $\lambda$ | $=$ | $\lambda$ |
| Geometric | $\beta$ | $<$ | $\beta(1+\beta)$ |
| Negative | $r \beta$ | $<$ | $r \beta(1+\beta)$ |
| Binomial |  |  |  |

The number of losses $N$ has mean 6 and variance 24 .
If given a choice between a binomial, Poisson, geometric, or negative binomial distribution, which would you choose?
What values would you use for the parameters?

## Exercise 1

The number of losses $N$ has mean 6 and variance 24 .
If given a choice between a binomial, Poisson, geometric, or negative binomial distribution, which would you choose?
What values would you use for the parameters?
$\operatorname{Var}[N]>\mathrm{E}[N]$, so it is not binomial or Poisson.
For a negative binomial,

$$
\begin{aligned}
\mathrm{E}[N] & =r \beta=6 \\
\operatorname{Var}[N] & =r \beta(1+\beta)=24 \\
\frac{\operatorname{Var}[N]}{\mathrm{E}[N]} & =1+\beta=4 \\
\beta & =3 \quad \Rightarrow r=2
\end{aligned}
$$

Since $r=2$ and not 1 , it is a negative binomial, not a geometric. Alternatively, a geometric with mean 6 has variance $6(6+1)=42$, not 24 , so we knew it wasn't geometric from that as well.

If $N$ is Poisson with mean 2.4, find the mode of $N$.

## Exercise 2

If $N$ is Poisson with mean 2.4, find the mode of $N$.

We want to find what $k$ maximizes $p_{k}$.
The data or table functions on the calculator show it is $k=2$.
The algebraic approach is to look at $p_{k} / p_{k-1}$.
$p_{k}>p_{k-1}$ if and only if $p_{k} / p_{k-1}>1$. The mode is the largest $k$ such that $p_{k}>p_{k-1}$. But

$$
\frac{p_{k}}{p_{k-1}}=\frac{e^{-2.4} \cdot \frac{2.4^{k}}{k!}}{e^{-2.4} \cdot \frac{2.4^{k-1}}{(k-1)!}}=\frac{2.4}{k}
$$

$p_{k} / p_{k-1}>1$ if $k<2.4$, i.e., $k \leq 2$, and $p_{k} / p_{k-1}<1$ for $k \geq 3$, making 2 the mode.
The calculator approach will be more useful on the exam.

