Poisson Distribution: Expected Value and Variance:

Assume we have a random variable Y, such that $Y \sim Poisson(\lambda)$, then we can find its expected value and variance the following way. Recall that the expected value for any discrete random variable is a sum of all possible values multiplied by their likelihood of occurring P(y). Thus,

$$E(Y) = \sum_{y=0}^{\infty} y \frac{e^{-\lambda} \lambda^{y}}{y!}$$

Now, when y=0, the entire product is 0, so we can start the sum from y=1 instead. Additionally, we can divide the numerator and denominator by "y", since "y" will be non-zero in every case.

$$= \sum_{y=1}^{\infty} y \frac{e^{-\lambda} \lambda^y}{y!} = \sum_{y=1}^{\infty} \frac{e^{-\lambda} \lambda^y}{(y-1)!}$$

Since lambda is a constant number, we can take out $\lambda e^{-\lambda}$, in front of the sum.

$$= \lambda e^{-\lambda} \sum_{y=1}^{\infty} \frac{\lambda^{y-1}}{(y-1)!}$$

Since we have "y-1" in both the numerator and denominator and the sum starts from 1, this is equivalent to starting the sum from 0 and using y instead.

$$= \lambda e^{-\lambda} \sum_{y=0}^{\infty} \frac{\lambda^{y}}{y!}$$

Calculus dictates that for any constant "c", $\sum_{x=0}^{\infty} \frac{c^x}{x!} = e^c$. We use this to simply the expression to:

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

Lastly, since any value to the negative power is the same as 1 divided by that same value, then $e^{-\lambda}e^{\lambda}=1$.

$$=\lambda$$

Now, let's move on to the variance. We are first going to express it in terms of expected values and then we are going to apply a similar approach to the one we used for the expected value.

We start off with the well-known relationship between the expected value and the variance, the variance is equal to the expected value of the squared variable, minus the expected value of the variable, squared.

$$Var(Y) = E(Y^2) - E(Y)^2$$

This next step might seem rather unintuitive, but we are simply expressing the squared variable in a way which makes it easier to manipulate. Knowing how to do proper operations with expected values, allows us to simplify the expression and plug in values we already know:

$$= E((Y)(Y-1) + Y) - E(Y)^{2}$$

$$= E((Y)(Y-1)) + E(Y) - E(Y)^{2}$$

$$= E((Y)(Y-1)) + (\lambda - \lambda^{2})$$

We turn to the definition of the expected value once again.

$$= \sum_{y=0}^{\infty} (y)(y-1) \frac{e^{-\lambda} \lambda^{y}}{y!} + (\lambda - \lambda^{2})$$

From here on out, the steps are pretty much the same once we took for the expected value:

- 1) we change the starting value of the sum, since the first 2 are zeroes
- 2) we cross our repeating values in the numerator and denominator
- 3) we take out the constant factors in front of the sum
- 4) adjust the starting value of the sum once again
- 5) substitute the sum with e^{λ}

$$= \sum_{y=2}^{\infty} (y)(y-1) \frac{e^{-\lambda} \lambda^{y}}{y!} + (\lambda - \lambda^{2})$$

$$= \sum_{y=2}^{\infty} \frac{e^{-\lambda} \lambda^{y}}{(y-2)!} + (\lambda - \lambda^{2})$$

$$= \lambda^{2} e^{-\lambda} \sum_{y=2}^{\infty} \frac{\lambda^{y-2}}{(y-2)!} + (\lambda - \lambda^{2})$$

$$= \lambda^{2} e^{-\lambda} \sum_{y=0}^{\infty} \frac{\lambda^{y}}{y!} + (\lambda - \lambda^{2})$$

$$= \lambda^{2} e^{-\lambda} e^{\lambda} + (\lambda - \lambda^{2})$$

$$= \lambda^{2} + \lambda - \lambda^{2}$$

$$= \lambda$$

Therefore, both the mean and variance for a Poisson Distribution are equal to lambda (λ) .