# Lecture 3 Exercises

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## Normal distribution exercises

1. Generate 10 draws from a normal random variable with mean 5 and variance 4. What is the sample mean of these variables? What is the true mean?

```
set.seed(12)
xsamp <- rnorm(10,5,2)
mean(xsamp)</pre>
```

## [1] 4.065572

2. How would you expect the previous answer to change if we increased the number of draws to 10,000? Check your intuition.

```
set.seed(12)
xsamp <- rnorm(10000,5,2)
mean(xsamp)</pre>
```

## [1] 4.997451

3. The probability density function (pdf) for a normal random variable X with mean  $\mu$  and variance  $\sigma^2$  is as follows,

$$f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} exp(\frac{1}{2\sigma^2}(x-\mu)^2)$$

What is the value of  $f_X(0)$  when  $\mu = 0$  and  $\sigma^2 = 2$ ? Calculate using the above expression and built-in R function.

```
#P(X<=2)
pnorm(2,5,2)
```

## [1] 0.0668072

```
#P(X>2)
1-pnorm(2,5,2)
```

## [1] 0.9331928

```
pnorm(2,5,2,lower.tail = FALSE)
## [1] 0.9331928
```

## Poisson distribution exercises

1. Generate a random sample of size 100,000 from a Poisson distribution with rate parameter 5

```
set.seed(80417)
xsamp <- rpois(100000,5)</pre>
```

2. Compute the sample mean and variance. What do you notice?

```
mean(xsamp)
```

```
## [1] 4.99062
```

```
var(xsamp)
```

## [1] 5.014122

3. Compute the empirical estimates of P(X=0), P(X=1), and P(X=2) for  $X \sim Pois(5)$  from your sample.

```
mean(xsamp == 0)
```

```
## [1] 0.00644
```

```
mean(xsamp == 1)
```

## [1] 0.03455

```
mean(xsamp == 2)
```

## [1] 0.08522

```
est <- sapply(0:2, function(x) mean(xsamp == x))</pre>
```

4. Compare the estimates to the true value of the probability mass function (pmf) at x = 0, 1, 2. Use the pmf and the built-in R function. Note the pmf is given by:

$$P(X = x) = \frac{\lambda^x e^{-\lambda}}{x!}$$
 where  $x \in \{0, 1, 2, ... \text{ and } \lambda > 0\}$ 

```
dpois(0:2,5)
```

```
## [1] 0.006737947 0.033689735 0.084224337
```

est

## [1] 0.00644 0.03455 0.08522

```
sapply(0:2,dpois,lambda=5)
```

```
## [1] 0.006737947 0.033689735 0.084224337
```

5. Compute an estimate of P(0 < X < 3) from the sample and compare to true value.

```
mean((xsamp < 3) & (xsamp > 0))
```

## [1] 0.11977

```
dpois(2,5) + dpois(1,5)
## [1] 0.1179141
ppois(2,5) - ppois(0,5)
## [1] 0.1179141
6. Find the smallest k such that 0.30 < P(X < k) for X ~ Pois(5) using a while loop. Then, do this using a built-in R function.

cdf_k = 0
k = 0
while(cdf_k < 0.3){
k=k+1
cdf_k = ppois(k,lambda=5)
}
k
## [1] 4
qpois(.3,5)</pre>
```

## [1] 4

## Group exercises

To display your results, create a table in Rmarkdown using the kable() function. Try to make it as clean as possible (i.e. column headers, title, digits, etc.).

Fun note: you will prove these results formally in the probability course!

#### Group 1

Generate 100,000 samples from a geometric distribution with p=0.3. Estimate  $P(X \ge s + t | X \ge t)$  and  $P(X \ge s)$  for s=4 and t=1,2,3,4,5,6. Compare to the true values. What do you notice? Google 'memoryless property distribution' and take a look at the wiki page on memorylessness. What does this suggest about the geometric distribution?

## Group 2

Generate 10,000 samples from a Bin(3,0.5) and another 10,000 samples from Bin(5,0.5). Compute the empirical cdf of the sum of the two samples and compare to the distribution function of a  $Z \sim Bin(8,0.5)$  random variable. What does this suggest about the distribution of X + Y where  $X \sim Bin(n_1, p)$  and  $Y \sim Bin(n_2, p)$ ?