

# Normal Probability Distribution

## CJ 702: Advanced Criminal Justice Statistics

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## 1 Load the USArrest data

First, let's load in the built-in USArrest data, and take a look at the first 10 observations (US states).

```
data(USArrests)
head(USArrests, 10)
```

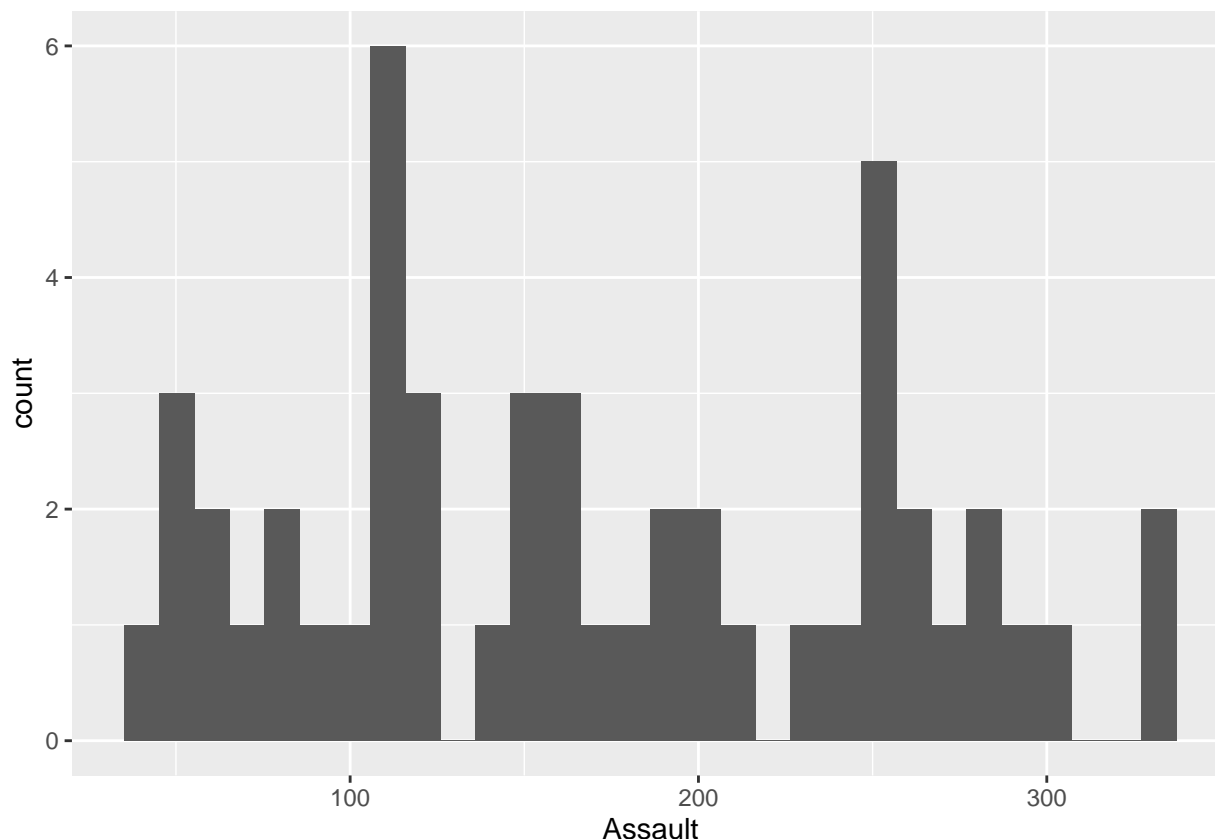
##		Murder	Assault	UrbanPop	Rape
##	Alabama	13.2	236	58	21.2
##	Alaska	10.0	263	48	44.5
##	Arizona	8.1	294	80	31.0
##	Arkansas	8.8	190	50	19.5
##	California	9.0	276	91	40.6
##	Colorado	7.9	204	78	38.7
##	Connecticut	3.3	110	77	11.1
##	Delaware	5.9	238	72	15.8
##	Florida	15.4	335	80	31.9
##	Georgia	17.4	211	60	25.8

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## 2 Viewing the *frequency distribution* for the Assault variable

Now, let's visualize the *frequency distribution* for the Assault variable. You know how to do this, we're just going to present it as a histogram.



## 3 The normal *probability distribution*

However, the *observed frequency distribution* is not the same as the *probability distribution*.

The normal probability distribution (*assumed* by Ordinary Least Squares Regression) is typically indicated with the following expression:

$$N(\mu, \sigma^2)$$

This expression is simply saying that you are working with a normal distribution with a given mean,  $\mu$ , and standard deviation,  $\sigma$ .

We could visualize this *theoretical* (reads: *assumed*) probability distribution using the following probability density function:

$$f(x) = \frac{1}{\sigma\sqrt{2\pi}} e^{-\frac{1}{2}\left(\frac{x-\mu}{\sigma}\right)^2}$$

As ever,  $\mu$  is the mean of your variable,  $\sigma$  is the standard deviation of your variable. As for the letters you may not recognize:  $\pi$  is 3.14159..., and  $e$  is Euler's constant, or 2.71828... Finally,  $x$  is any possible

value that can be assumed by your independent variable. Below, I am going to treat this as a sequence of numbers that starts at the minimum assault rate, 45, ends at the maximum assault rate,  $-\infty$ , and increases in intervals of 0.01.

As seen below, this equation generates a normal distribution that represents all theoretically possible values of a given *normally distributed continuous variable*. Unlike your *frequency distribution* (i.e., the histogram above), which is beholden to the observations that *actually exist*, this *probability distribution* can *theoretically* take on any value for variable  $x$  (here, the assault rate per 100,000).

```
# Find the mean:
mu <- USArrests$Assault %>% mean(na.rm = TRUE)

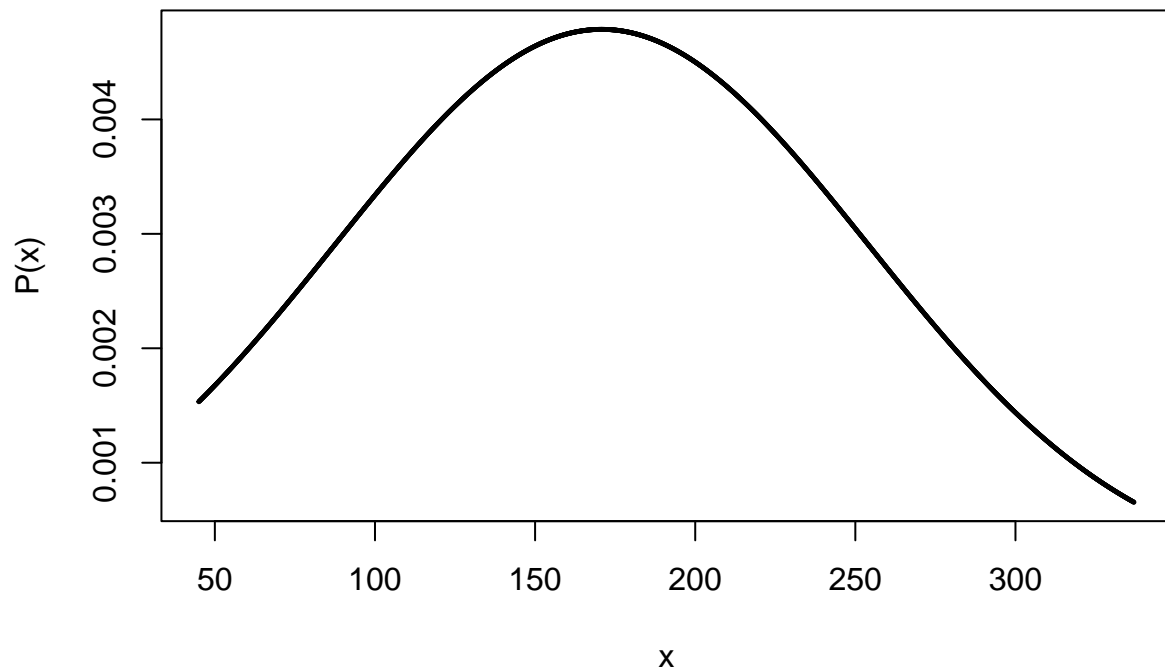
# Find the standard deviation:
sigma <- USArrests$Assault %>% sd(na.rm = TRUE)

# Generate a sequence of all possible values for x:
min <- USArrests$Assault %>% min(na.rm = TRUE)
max <- USArrests$Assault %>% max(na.rm = TRUE)

x <- seq(min, max, by = 0.1)

# Insert these values into the normal probability density function:
npd <- (1 / (sigma * sqrt(2 * pi))) * exp(1)^(-(x - mu)^2 / (2 * sigma^2))

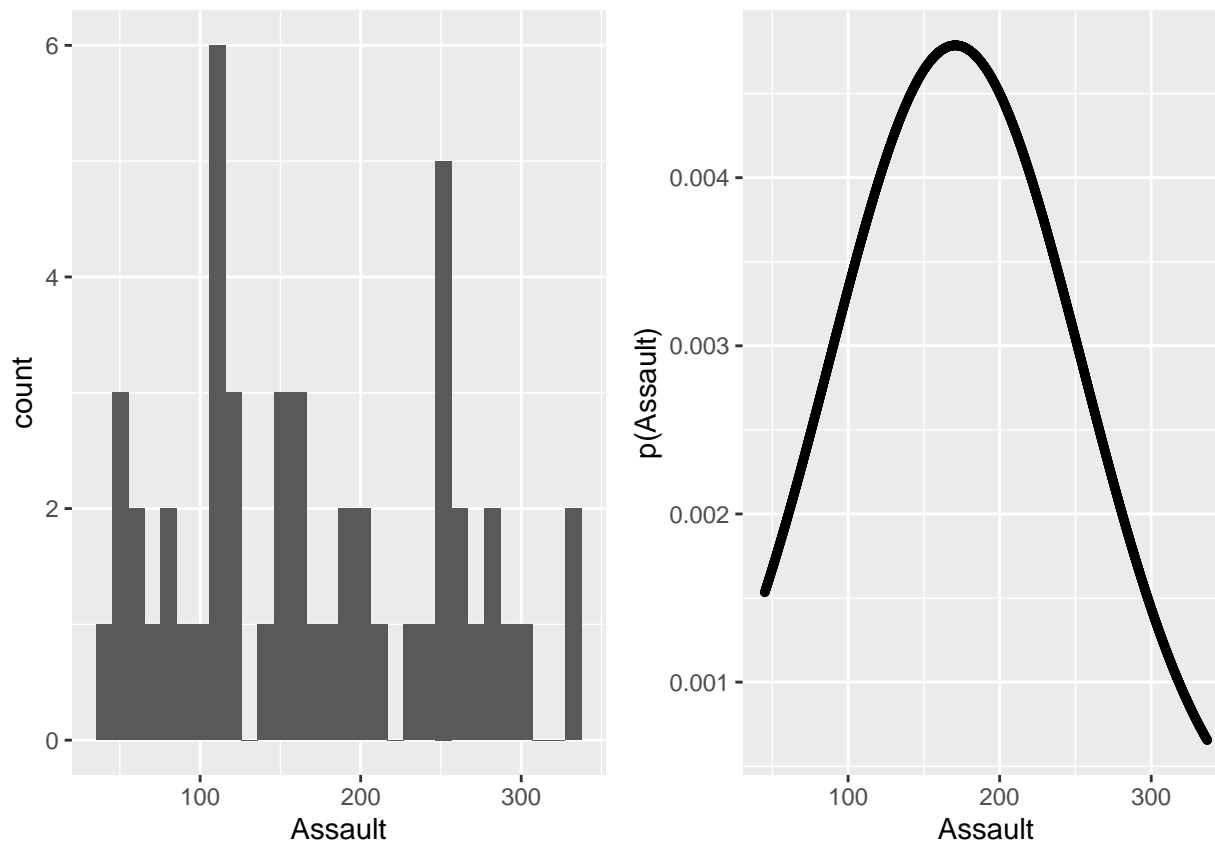
# Plot the probability distribution:
plot(x, npd,
      ylab = "P(x)",
      cex = 0.2)
```



## 4 The problem of non-normal observed distributions

This probably all sounds great, but there is a hitch. Let's look at our frequency distribution and our probability distribution side-by-side:

```
plot1 <- ggplot(USArrests, aes(x = Assault)) +  
  geom_histogram(bins = 30)  
  
plot2 <- ggplot(data.frame(x, npd), aes(x = x, y = npd)) +  
  geom_point(size = 1) +  
  ylab("p(Assault)") +  
  xlab("Assault")  
  
grid.arrange(plot1, plot2, ncol = 2)
```



Ask yourself: does the observed distribution on the left look much like the probability distribution we have generated on the right? No, not particularly.

However, when your ordinary least squares regression *assumes* that your Assault variable is normally distributed, this exact probability distribution is assumed to be ‘true’. So, if your observed frequency distribution looks *nothing* like the probability distribution, you might want to rethink the assumption that your variable is normally distributed!