

STATS 2107  
Statistical Modelling and Inference II

Workshop 4: Sampling distributions part 1

Matt Ryan

School of Mathematical Sciences, University of Adelaide

Semester 2 2022

The sampling distribution of the sample mean

## What is a sampling distribution?

Suppose  $Y_1, Y_2, \dots, Y_n$  is a random sample, and  $T$  is a statistic on the  $Y_i$ . Then the distribution of  $T$  is called the *sampling distribution*.

## The sample mean

For example, suppose each  $Y_i \sim N(\mu, \sigma^2)$  and  $T = \bar{Y}$ . Then the sampling distribution is

$$\bar{Y} \sim N\left(\mu, \frac{\sigma^2}{n}\right).$$

## What is meant by sampling distribution?

$$\begin{array}{cccccc} Y_{11}, & Y_{12}, & \dots, & Y_{1n} & \rightarrow & T_1 \\ Y_{21}, & Y_{22}, & \dots, & Y_{2n} & \rightarrow & T_2 \\ Y_{31}, & Y_{32}, & \dots, & Y_{3n} & \rightarrow & T_3 \\ Y_{41}, & Y_{42}, & \dots, & Y_{4n} & \rightarrow & T_4 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \end{array}$$

## Does the practice match the theory?

In theory, if our data is normal, the sample mean is normal. Let's test this.

1. Consider samples of size 3,  $Y_1, Y_2, Y_3 \sim N(5, 2^2)$ .
2. Every time we take a sample, calculate the mean

$$\bar{Y} = \frac{1}{3} (Y_1 + Y_2 + Y_3) .$$

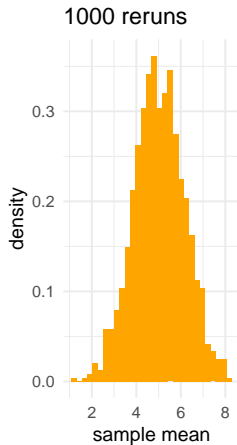
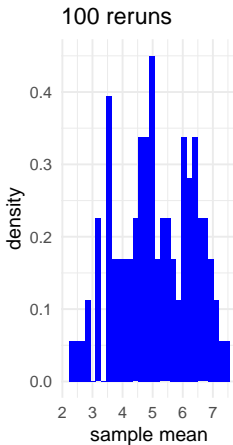
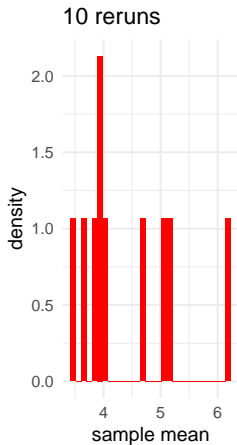
3. Generate 10, 100, and 1000 samples to look at the distribution.
4. Is it normal?

# Some R code to do this

```
# Set up some parameters
N <- 10
mu <- 5
sig <- 2
n <- 3

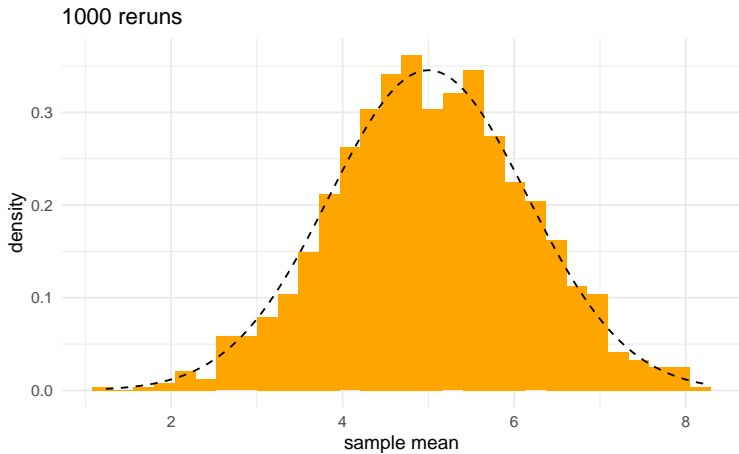
# Get the samples and calculate the mean
norm_sample_3_10 <- N %>%
  rerun(rnorm(n, mu, sig)) %>%
  map_dbl(mean) #Hey look, a new function!
```

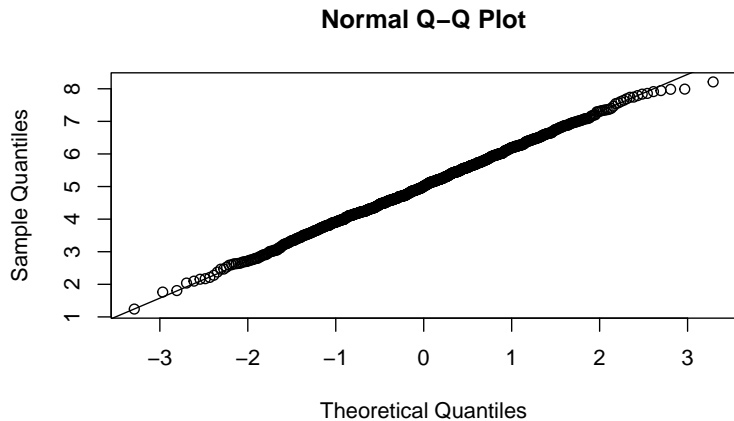
# Histograms





# Is this normal?





Your turn

# What to do

1. Adapt the given code to produce the histograms for  $N = 10, 100, 1000$ .
2. Explore the distribution as you increase  $n$ .
3. Explore the distribution as you change  $\mu$  and  $\sigma^2$ .

Non-normal data

## The problem

Our distributional result relies on the fact that  $Y_i \sim N(\mu, \sigma^2)$ , although we know

$$E[\bar{Y}] = \mu$$

and

$$\text{Var}(\bar{Y}) = \frac{\sigma^2}{n}.$$

## CLT to the rescue?

Let  $Y_1, Y_2, \dots, Y_n$  be independent and identically distributed random variables with  $E[Y_i] = \mu$  and  $\text{Var}(Y_i) = \sigma^2 < \infty$ . Define

$$U_n = \frac{\bar{Y} - \mu}{\sigma/\sqrt{n}}.$$

Then the distribution of  $U_n$  converges to the standard normal distribution function as  $n \rightarrow \infty$ .

# The problem

The CLT only kicks in for large  $n$ , the worse the distribution, the larger the  $n$  needed.



$$\chi_5^2$$

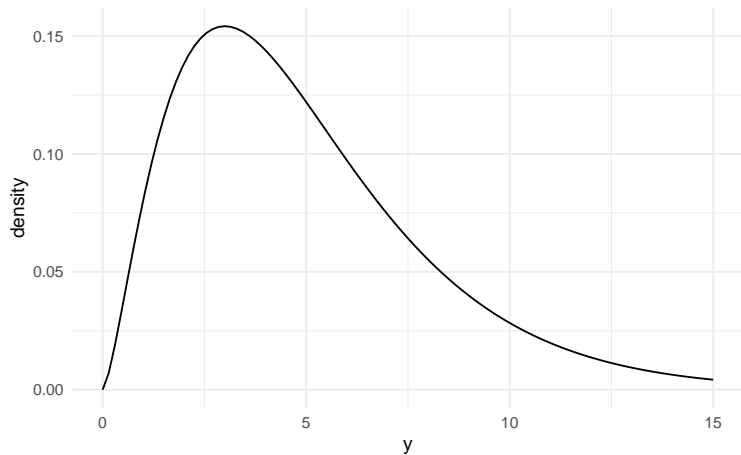
Let's explore the sampling distribution of the sample mean for  $Y_1, Y_2, \dots, Y_n \sim \chi_5^2$ . We will

1. Consider samples of size 3,  $Y_1, Y_2, Y_3 \sim \chi_5^2$ .
2. Every time we take a sample, calculate the mean

$$\bar{Y} = \frac{1}{3} (Y_1 + Y_2 + Y_3) .$$

3. Generate 10, 100, and 1000 samples to look at the distribution.
4. Is it normal? Expect to see  $N(5, 10/3)$ .

Is the  $\chi^2_5$  normal?

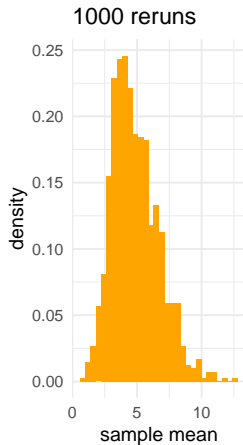
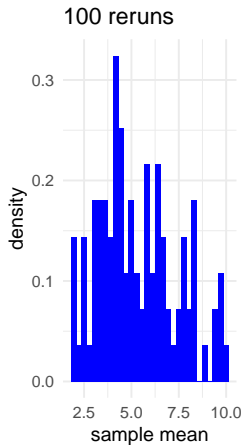
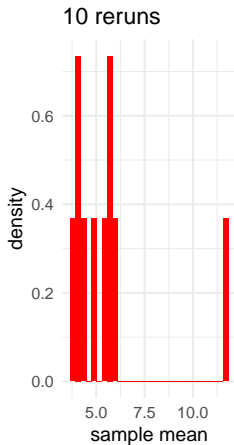


# Some R code to do this

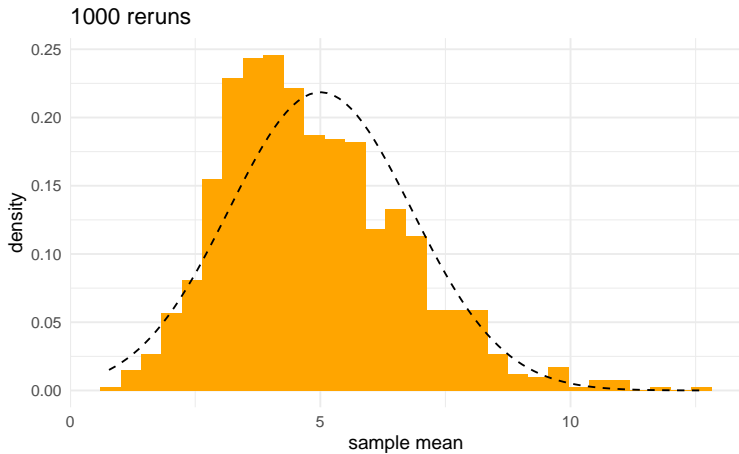
```
# Set up some parameters
N <- 10
df <- 5
n <- 3

# Get the samples and calculate the mean
chi_sample_3_10 <- N %>%
  rerun(rchisq(n, df)) %>%
  map_dbl(mean)
```

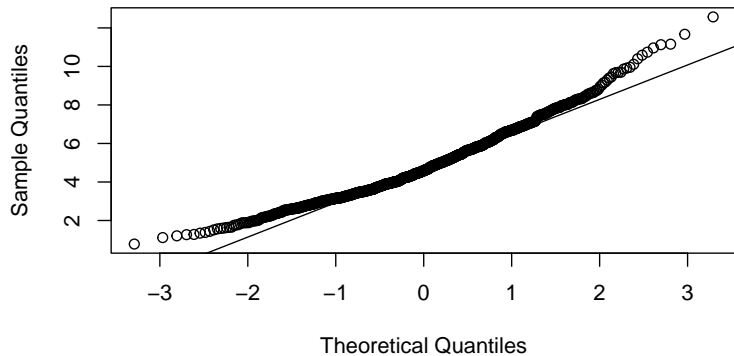
# Histograms



# Is this normal?



**Normal Q-Q Plot**



Your turn

## What to do

1. Explore the distribution of the sample mean as you increase the sample size  $n$  from the  $\chi^2_5$ . When does it start to become normal?
2. Look at the  $t_5$  distribution. Explore the sampling distribution of the sample mean. When does it start to become normal?
3. If you had a dataset with no knowledge of its distribution, how might you explore the sampling distribution of the sample mean?