Normal quantile plots

- see that normal distributions of data (or being normal enough) important
- only tools we have to assess this are histograms and maybe boxplots
- > a better tool is normal quantile plot:

plot data against what you expect if data actually normal
 look for points to follow a straight line, at least approx
 ggplot code: aes sample; geoms stat_qq and stat_qq_line

Packages

The usual:

library(tidyverse)

Kids learning to read



Get the groups separately

kids %>% filter(group == "t") -> treatment
kids %>% filter(group == "c") -> control

to check

treatment %>% count(group)

A tibble: 1 x 2

group n <chr> <int>

1 t 21

control %>% count(group)

A tibble: 1 x 2
group n
<chr> <int>
1 c 23

The treatment group

ggplot(treatment, aes(sample = score)) +
stat_qq() + stat_qq_line()



only problem here is lowest value a little too low (mild outlier).

Control group

ggplot(control, aes(sample = score)) +
stat_qq() + stat_qq_line()



This time, highest value a little too high, but again, no real problem with normality.

Facetting more than one sample

Use the whole data set and facet by groups

ggplot(kids, aes(sample = score)) +
stat_qq() + stat_qq_line() + facet_wrap(~group)



Blue Jays attendances, skewed to right ggplot(jays, aes(x = attendance)) + geom_histogram(bins = 6



On a normal quantile plot

ggplot(jays, aes(sample = attendance)) +
stat_qq() + stat_qq_line()



Attendances at low end too bunched up: skewed to right.
 Right-skewness can also show up as highest values being too high, or as a curved pattern in the points.

More normal quantile plots

- How straight does a normal quantile plot have to be?
- There is randomness in real data, so even a normal quantile plot from normal data won't look perfectly straight.
- With a small sample, can look not very straight even from normal data.
- Looking for systematic departure from a straight line; random wiggles ought not to concern us.
- Look at some examples where we know the answer, so that we can see what to expect.

Normal data, large sample

d <- tibble(x=rnorm(200))
ggplot(d, aes(x=x)) + geom_histogram(bins=10)</pre>



ggplot(d,aes(sample=x))+stat_qq()+stat_qq_line()



Normal data, small sample

Not so convincingly normal, but not obviously skewed: d <- tibble(x=rnorm(20))</p>

ggplot(d, aes(x=x)) + geom_histogram(bins=5)



Good, apart from the highest and lowest points being slightly off. I'd call this good:

ggplot(d, aes(sample=x)) + stat_qq() + stat_qq_line()



Chi-squared data, df = 10

Somewhat skewed to right:

```
d <- tibble(x=rchisq(100, 10))
ggplot(d,aes(x=x)) + geom_histogram(bins=10)</pre>
```



Somewhat opening-up curve:

ggplot(d,aes(sample=x))+stat_qq()+stat_qq_line()



Chi-squared data, df = 3

Definitely skewed to right:

```
d <- tibble(x=rchisq(100, 3))
ggplot(d, aes(x=x)) + geom_histogram(bins=10)</pre>
```



Clear upward-opening curve:

ggplot(d,aes(sample=x))+stat_qq()+stat_qq_line()



t-distributed data, df = 3

Long tails (or a very sharp peak):

```
d <- tibble(x=rt(300, 3))
ggplot(d, aes(x=x)) + geom_histogram(bins=15)</pre>
```



Low values too low and high values too high for normal.

ggplot(d,aes(sample=x))+stat_qq()+stat_qq_line()



Summary

On a normal quantile plot:

- points following line (with some small wiggles): normal.
- kind of deviation from a straight line indicates kind of nonnormality:
 - > a few highest point(s) too high and/or lowest too low: outliers
 - else see how points at each end off the line:

	High points	
Low points	Too low	Too high
Too low	Skewed left	Long tails
Too high	Short tails	Skewed right

short-tailed distribution OK for t (mean still good), but others problematic (depending on sample size).