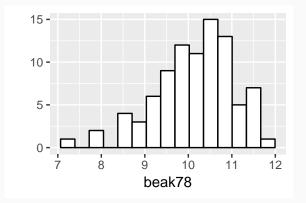
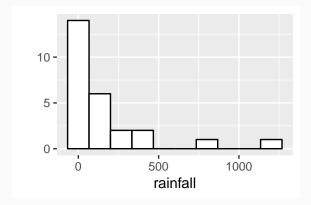
Checking For Normality

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Checking for normality

Some distributions look approximately normal





It's sometimes important to check if normality is a valid approximation.

- 1. Idea: Is the 68-95-99.7 rule approximately correct for a given dataset?
- 2. More generally, do the percentiles (quantiles) of the data match with the percentiles (quantiles) of the theoretical normal distribution?
- 3. Compare the *p*th percentile (quantile) of the data and the *p*th percentile (quantile) of a $N(\bar{x}, s^2)$ distribution. If they are pretty close, then normality is a good approximation.

```
mu <- mean(beak78)
sigma <- sd(beak78)</pre>
qnorm(p = 0.2, mean = mu, sd = sigma)
## [1] 9.375
quantile(x = beak78, probs = 0.2)
## 20%
## 9.46
```

That matches almost exactly, what about other percentiles (quantiles)?

```
qnorm(p = 0.7, mean = mu, sd = sigma)
## [1] 10.61
quantile(x = beak78, probs = 0.7)
## 70%
## 10.6
```

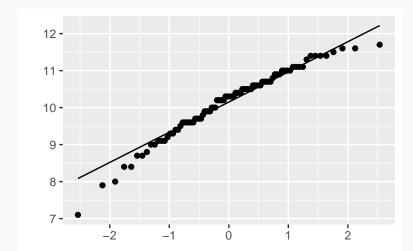
```
qnorm(p = 0.9, mean = mu, sd = sigma)
## [1] 11.3
quantile(x = beak78, probs = 0.9)
## 90%
## 11.14
```

These are all pretty close!

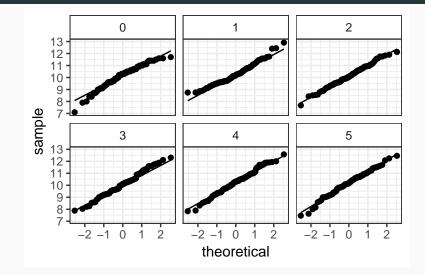
- 1. Plots the observed quantiles against the quantiles of a $N(\bar{x}, s^2)$ density.
- 2. If the points lie close to a line, then the normal approximation is approximately correct.
- 3. Can just plot the observed quantiles against N(0, 1) and look for a straight line (more on why later).

QQplot

qplot(sample = beak78, geom = "qq") + geom_qq_line()

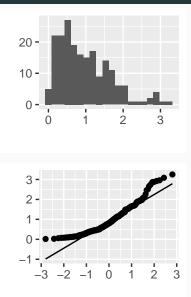


But what does a "good" qqplot look like?

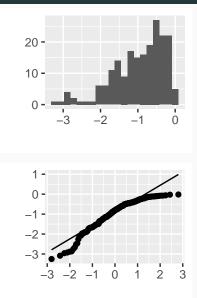


Top left is real data, rest are simulated from $N(\bar{x}, s^2)$ — maybe a little non-normal?

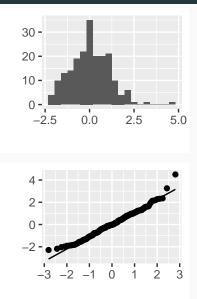
Problem: Skewed right



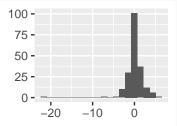
Problem: Skewed left

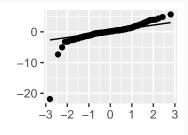


Problem: Outliers

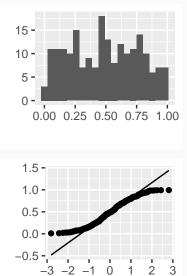


Problem: Heavy tails





Problem: Light tails



The rainfall data

qplot(sample = rainun, geom = "qq") + geom_qq_line()

