

# Stat310

## Moments

Hadley Wickham

# Engineer Your Career

Monday, February 15

7:00 PM - 8:30 PM

McMurtry Auditorium

Find out what you can do with a degree in engineering from a panel of successful Rice engineering graduates who have gone into a variety of professions. (Plus get dessert!)

[http://engineering.rice.edu/EventsList.aspx?  
EventRecord=13137](http://engineering.rice.edu/EventsList.aspx?EventRecord=13137)

# Homework

Due today.

From now on, if late, put in Xin Zhao's mail box in the DH mailroom.

Another one due next Thursday

Buy a stapler

Use official name

1. Finish off proof
2. More about expectation
3. Variance and other moments
4. The moment generating function
5. The Poisson distribution
6. Feedback

**Proof, continued**

# Expectation of a function

# Expectation

Expectation is a **linear operator**:

Expectation of a sum =  
sum of expectations (additive)

Expectation of a constant \* a function =  
constant \* expectation of function (homogenous)

Expectation of a constant is a constant.

# Your turn

Write (or recall) the mathematical description of these properties.

Work in pairs for two minutes.

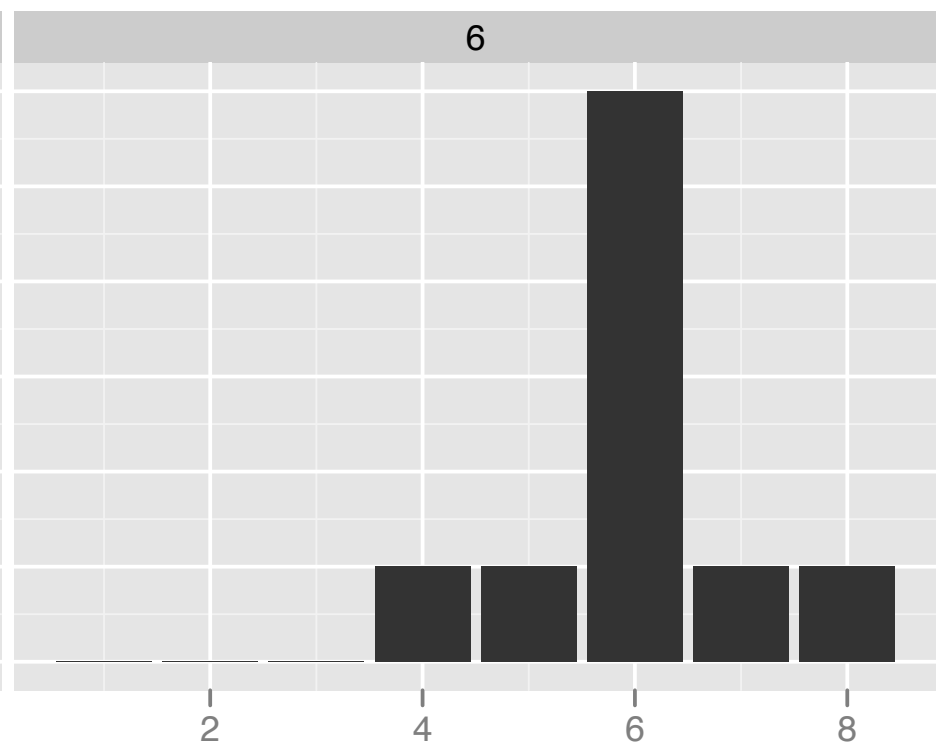
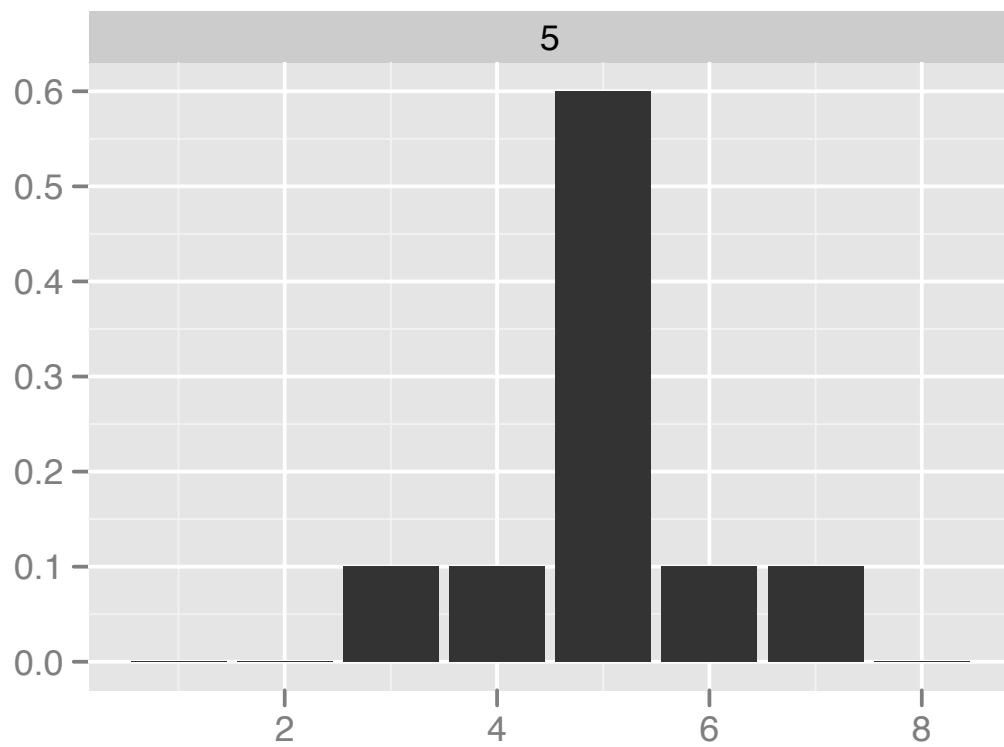
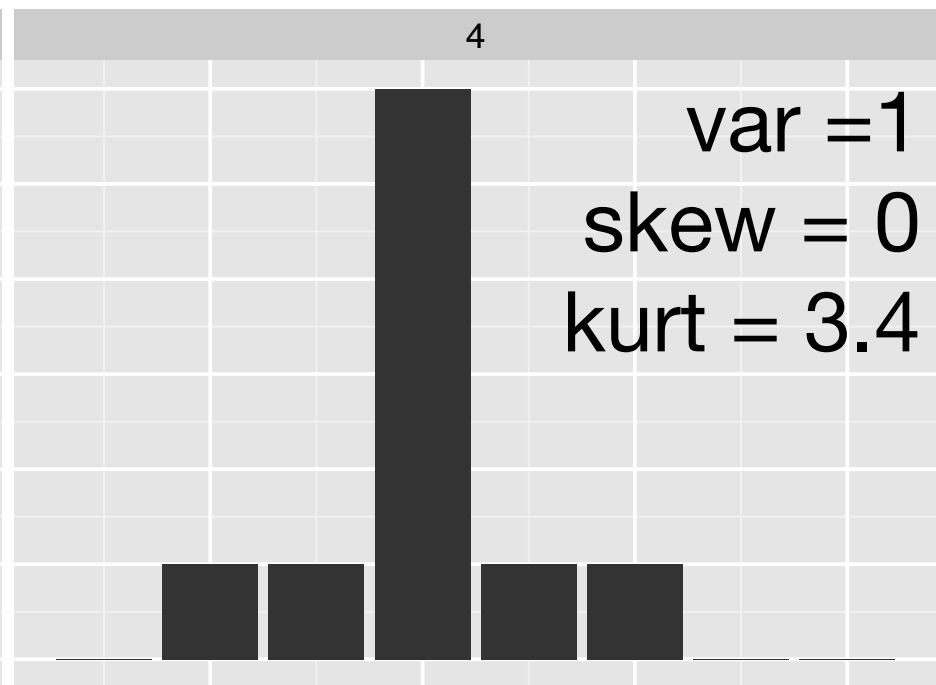
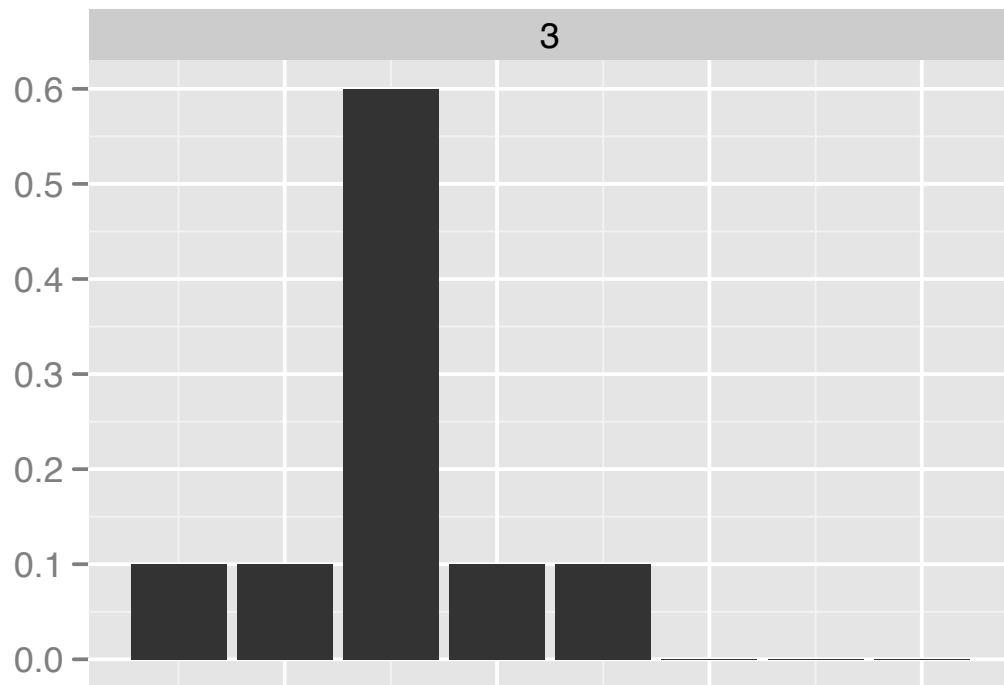
(Extra credit this week is to prove these properties)

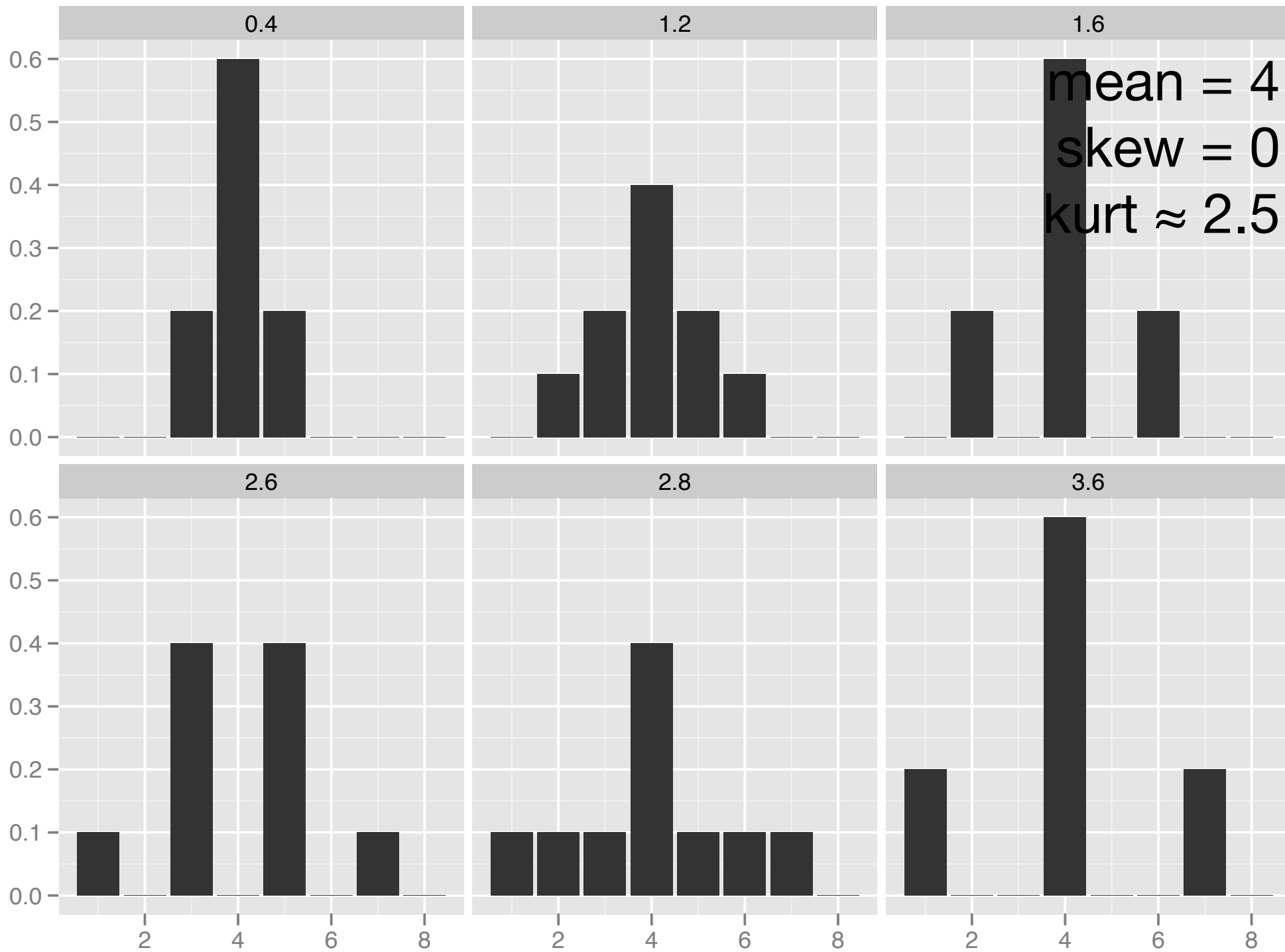
# Moments

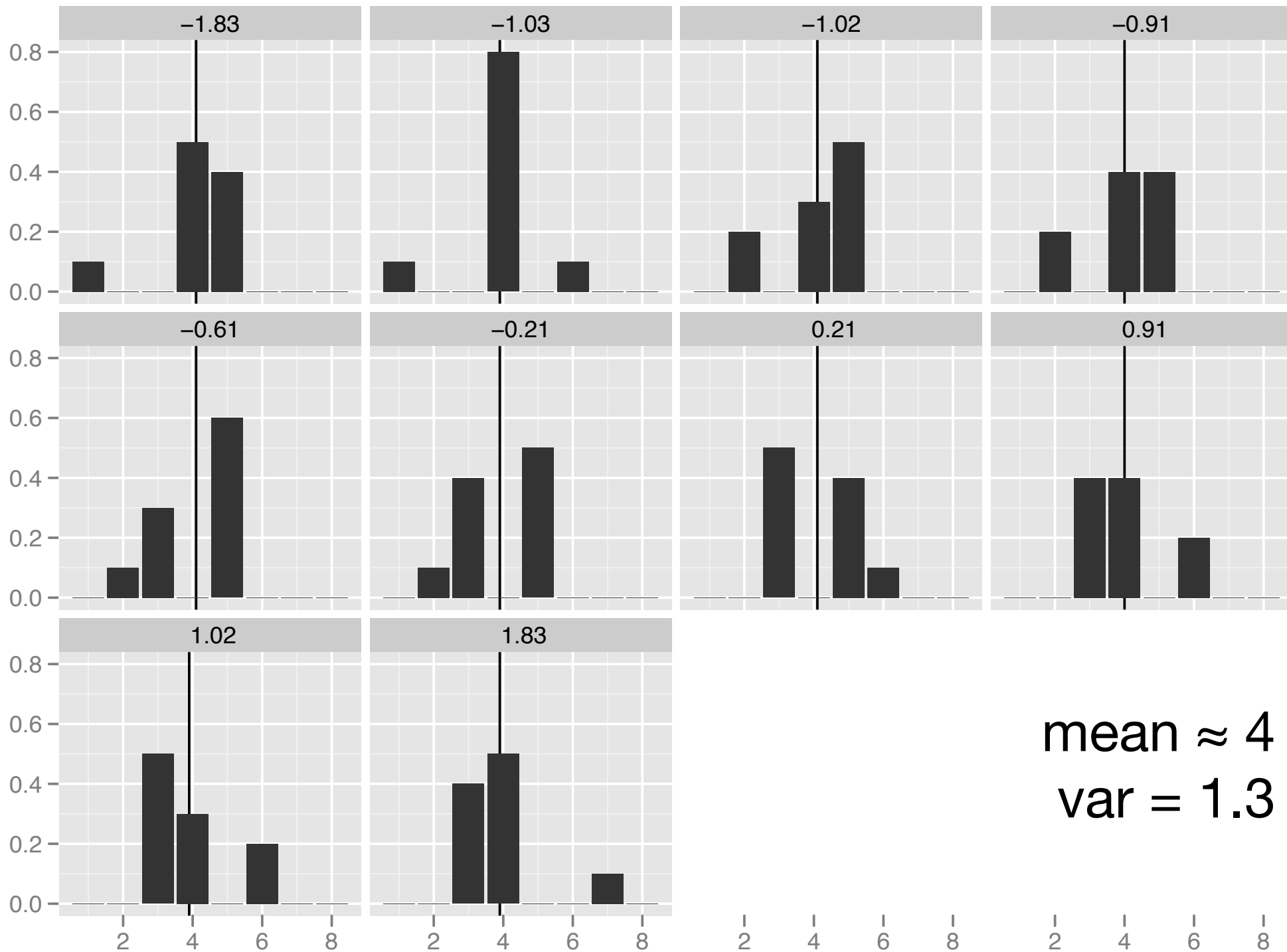
The  $i$ th **moment** of a random variable is defined as  $E(X^i) = \mu'_i$ . The  $i$ th **central moment** is defined as  $E[(X - E(X))^i] = \mu_i$

The mean is the \_\_\_\_\_ moment. The variance is the \_\_\_\_\_ moment.

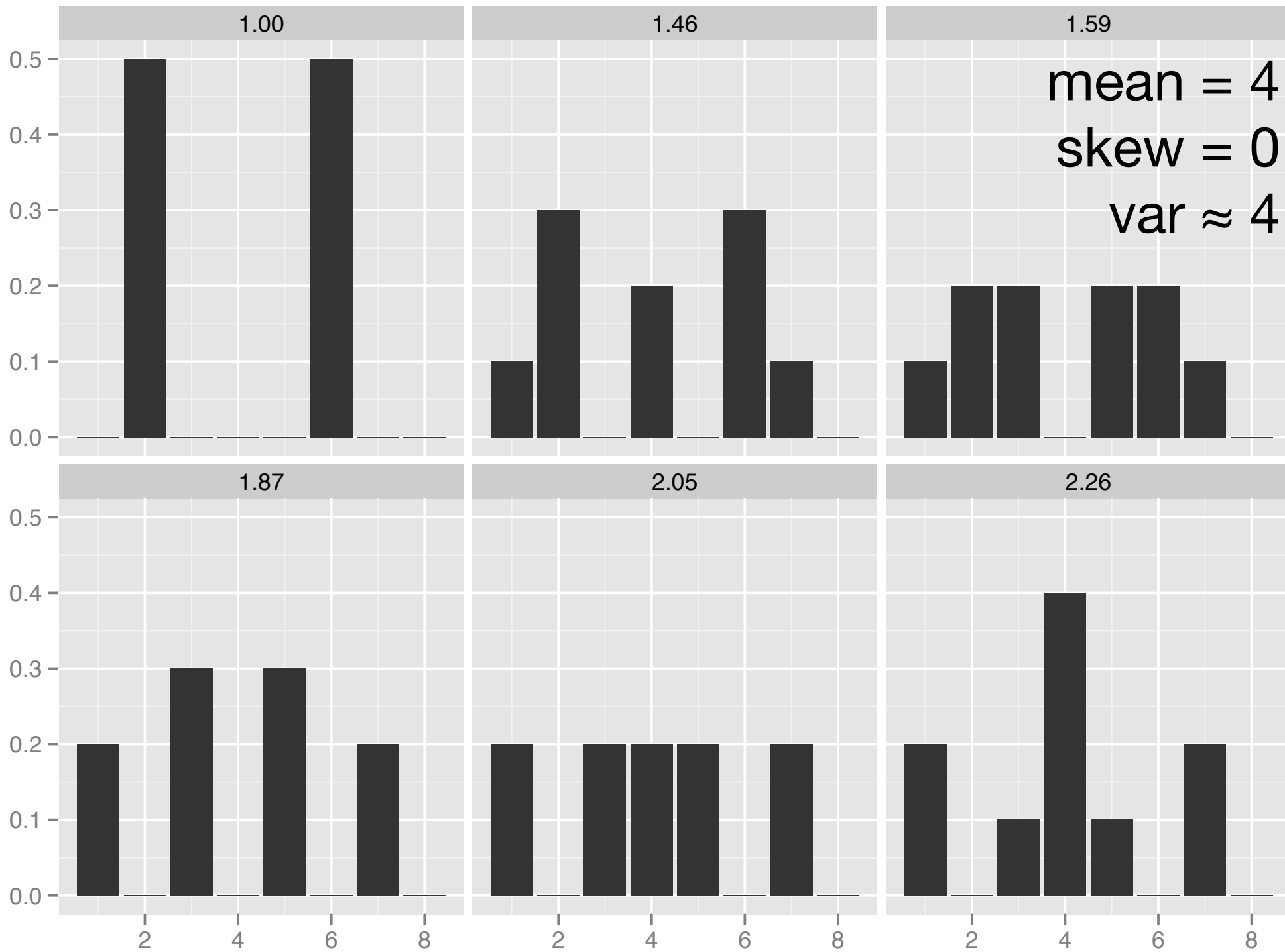
	Name	Symbol	Formula
1	mean	$\mu$	$\mu'_1$
2	variance	$\sigma^2$	$\mu_2 = \mu'_2 - \mu^2$
3	skewness	$\alpha_3$	$\mu_3 / \sigma^3$
4	kurtosis	$\alpha_4$	$\mu_4 / \sigma^4$







mean  $\approx 4$   
var = 1.3



# mgf

The **moment generating function (mgf)**  
is  $M_X(t) = E(e^{Xt})$

(Provided it is finite in a neighbourhood of 0)

Why is it called the mgf? (What happens if you differentiate it multiple times).

Useful property: If  $M_X(t) = M_Y(t)$  then  $X$  and  $Y$  have the same pmf.

**Plus**, once we've got it, it can make it much easier to find the mean and variance

# Expectation of binomial (take 2)

Figure out mgf.

(Random mathematical fact: binomial theorem)

Differentiate & set to zero.

Then work out variance.

# Your turn

Compute mean and variance of the binomial. Remember the variance is the 2<sup>nd</sup> central moment, not the 2<sup>nd</sup> moment.

# Poisson

# Poisson distribution

$X$  = Number of times some event happens

(1) If number of events occurring in non-overlapping times is **independent**, and

(2) probability of exactly one event occurring in short interval of length  $h$  is  $\propto \lambda h$ , and

(3) probability of two or more events in a sufficiently short interval is basically 0

# Poisson

$X \sim \text{Poisson}(\lambda)$

Sample space: positive integers

$\lambda \in [0, \infty)$

# Examples

Number of calls to a switchboard

Number of eruptions of a volcano

Number of alpha particles emitted from a radioactive source

Number of defects in a roll of paper

# Example

On average, a small amount of radioactive material emits ten alpha particles every ten seconds. If we assume it is a Poisson process, then:

What is the probability that no particles are emitted in 10 seconds?

Make sure to set up mathematically.

# mgf, mean & variance

Random mathematical fact.

Compute mgf.

Compute mean & 2<sup>nd</sup> moment.

Compute variance.

# Next week

Repeat for continuous variables.

Make **absolutely sure** you have read 2.5 and 2.6. (hint hint)

**Feedback**