## Equilibration, Uncertainty, and Bootstrapping

Or, what do my simulations actually mean?

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## Uncertainties in averages

Average in the mean is:

$$\bullet \langle U \rangle = \int p(x)U(x)dx$$

• But if we sample from the distribution p(x), we can replace **integral** with a **sum over observations** 

• 
$$\langle U \rangle = \frac{1}{N} \sum_{i} U(x_i)$$
 with  $x_i$  sampled from  $p(x)$ 

# How do we find the uncertainty of an estimate?

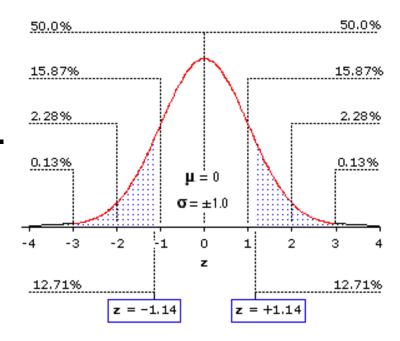
$$\langle U \rangle = \frac{1}{N} \sum_{i} U(x_i)$$

- What do we mean by the uncertainty?
- What we generally mean:
  - If we did the same experiment again and again, how different would the results of each experiment be?

## How do we REPORT the uncertainty?

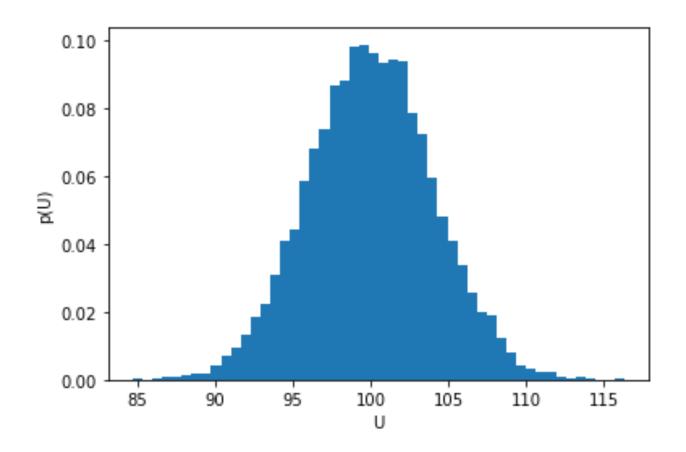
$$\langle U \rangle = \frac{1}{N} \sum_{i} U(x_i)$$

- We will get a <u>distribution</u> of answers.
- We usually report +/- something.
   What is that something?
- We usually report a "standard error of the mean".
- What does that mean?



#### Part 1: Assume we have a distribution for U

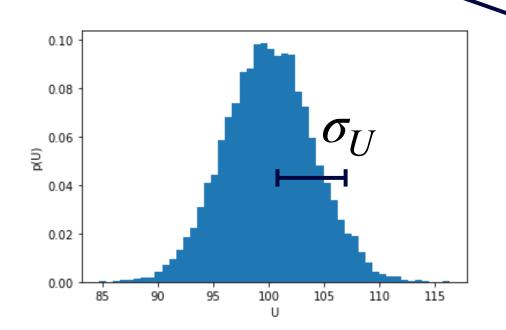
This is the distribution of a single sample from U



# Part 2: What is the standard deviation of this distribution?

- Sample standard deviation
  - Standard deviation, computed from samples

$$\sigma_U = \sqrt{\frac{\left\langle (U - \langle U \rangle)^2 \right\rangle}{N - 1}}$$



N is the number of independent samples you are calculating this from

•  $\sigma_U$  does not change in magnitude as you collect more samples, just gets more precise

# Part 3: What is the standard deviation in of the MEAN of these samples?

- We don't want the error in 1 sample from U
- We want the errors in N samples from U averaged together:  $\langle U \rangle = \frac{1}{N} \left( U_1 + U_2 + \ldots + U_N \right)$
- We know that errors add in the square, WHEN they are independent

$$\sigma_{\langle U \rangle}^2 = \frac{1}{N^2} \sum_{i}^{N} \sigma_U^2$$

$$\sigma_{\langle U \rangle}^2 = \frac{N}{N^2} \sigma_U^2$$

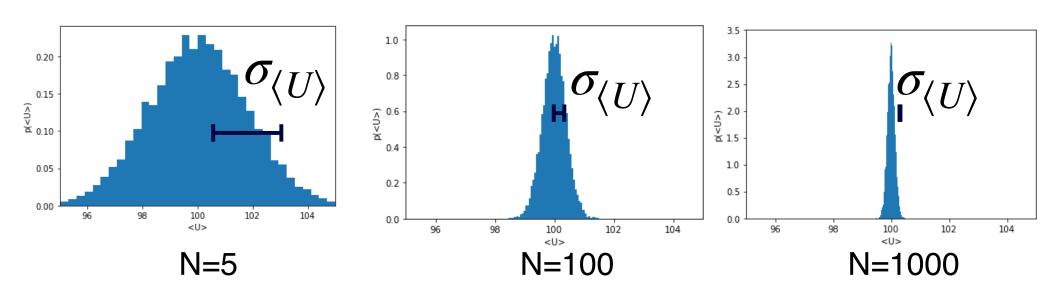
$$\sigma_{\langle U \rangle}^2 = \frac{1}{\sqrt{N}} \sigma_U$$

samples you are averaging together

# Part 3: What is the standard deviation in of the MEAN of these samples?

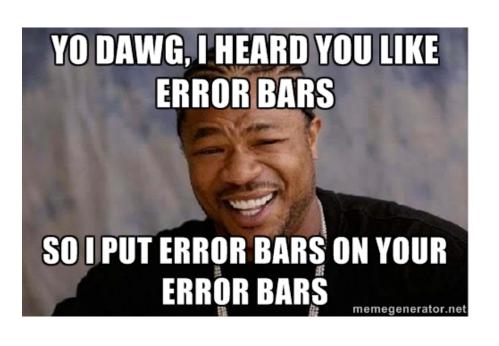
$$\sigma_{\langle U \rangle} = \frac{1}{\sqrt{N}} \sigma_U$$

- $\sigma_{\langle U \rangle}$  does change as the number of samples included in the average increases
- It gets smaller!



## What are the error bars in my error bars?

$$\sigma_{\langle U \rangle} \pm \sigma_{\sigma \langle U \rangle}$$
?



- Rough rule of thumb:
- If you have 40 independent samples, then the error in the standard error is <5%
- If you have 5 independent samples, use student tdistribution.

# What about distributions that are not normal?

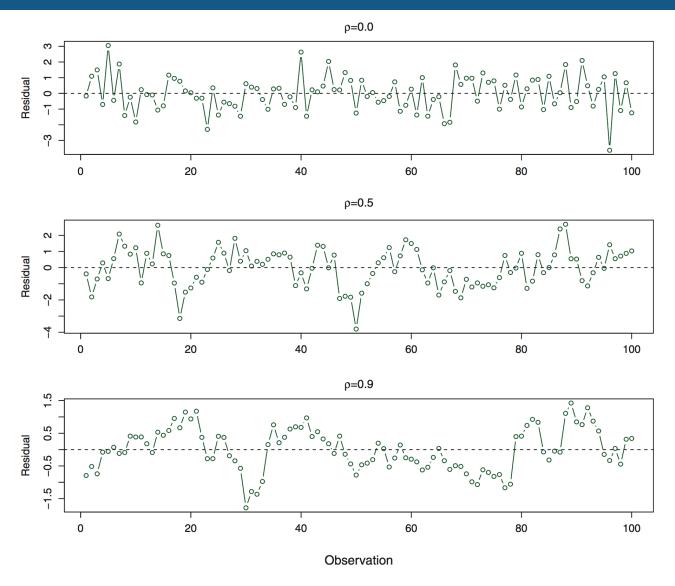
- How to report confidence intervals in those distributions?
- How to report confidence intervals of the <u>means</u> of those distributions?

# Go to notebook!

#### Two main tasks

- Identifying when my simulations have become stationary.
- Identifying how many simulation points are independent.

#### Identifying time correlation in simulation data



**FIGURE 3.10.** Plots of residuals from simulated time series data sets generated with differing levels of correlation  $\rho$  between error terms for adjacent time points.

# Go to notebook!

## How many independent points do I have?

The autocorrelation function is defined as:

$$A(\tau) = \frac{1}{Var(A(t))} \int_{0}^{\infty} A(t)A(t+\tau)dt$$
 Where:  $Var(A(t)) = \left\langle (A - \langle A \rangle)^{2} \right\rangle$ 

- We subtract mean from A so that <A> = 0
- We divide by Var(A(t)) to make sure that the autocorrelation functions starts at 1 when t=0.
- We usually have discrete samples, so we use:

$$A(\tau) = \frac{1}{Var(A(t))} \frac{1}{N} \sum_{t=0}^{N} A(t) A(t+\tau)$$
 Only meaningful for stationary series!

# Three ways to estimate when samples are independent

- Three ways to determine if they are not correlated anymore:
  - Determine when the autocorrelation function crosses to zero
  - Integrate the time under the autocorrelation function, use that as the correlation time.
  - Fit the autocorrelation to an exponential, estimate the characteristic time from  $\exp(-t/\tau)$

# Go to notebook!