

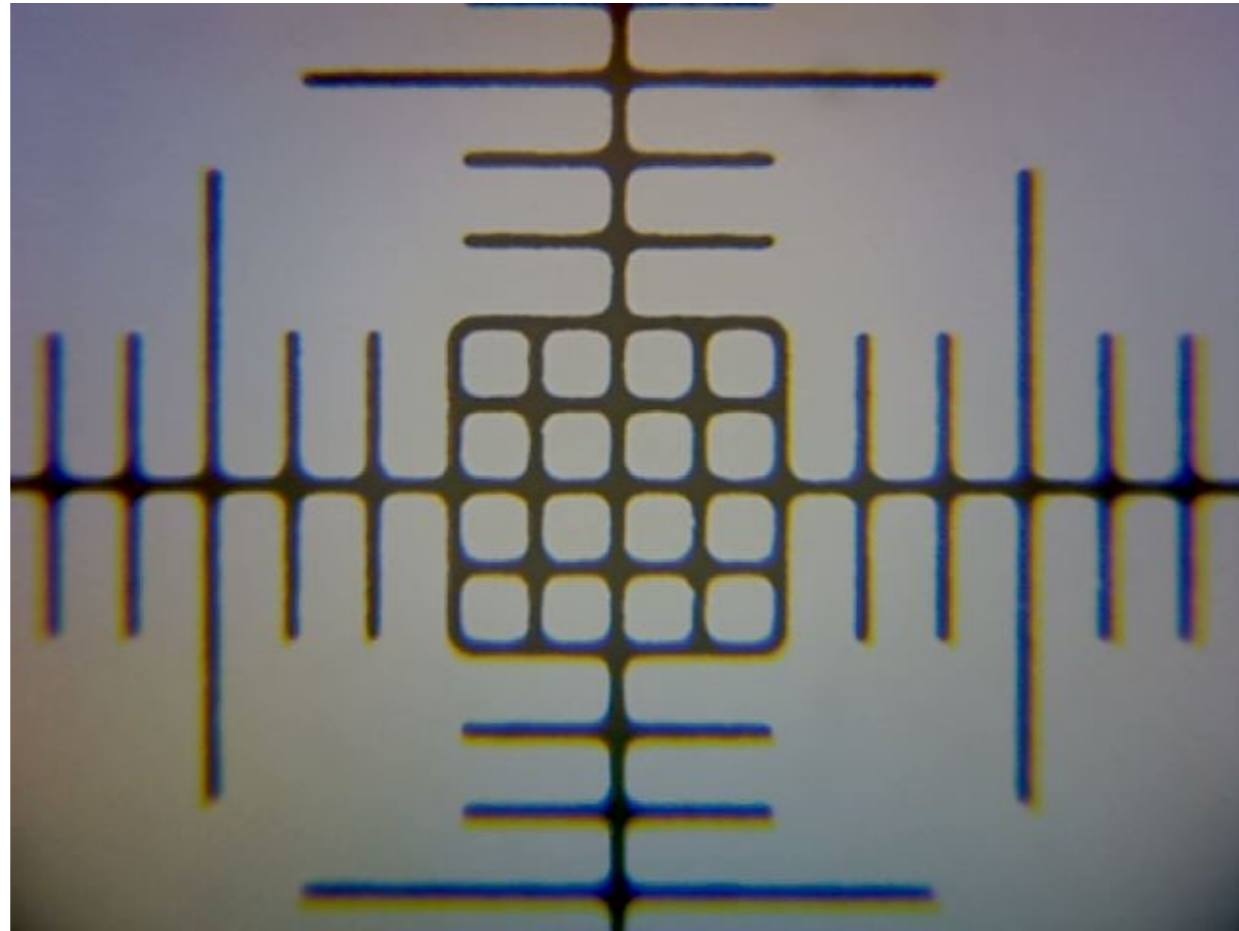
# PHYS 391 - Day 12

Lab 3 discussion  
Hypothesis Testing

<http://pages.uoregon.edu/torrence/391/class/day12.pdf>

# Lab 3 discussion

# Calibration

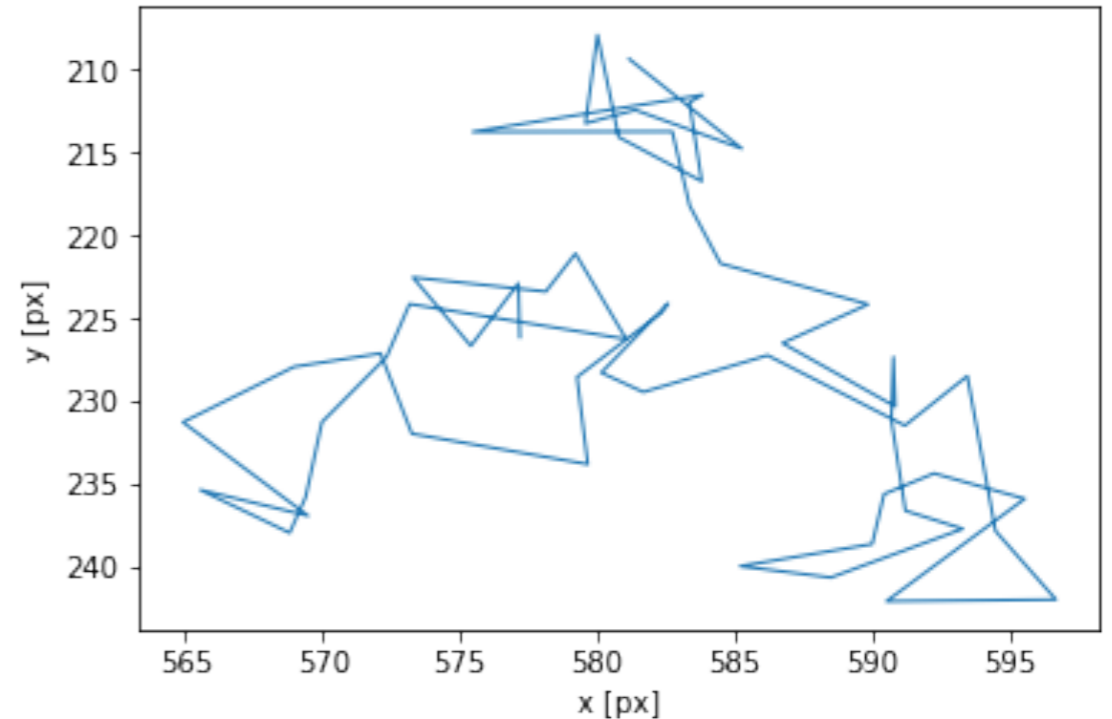


You must calibrate the images measured in microns

Determine  $\mu\text{m}$  per pixel

The marks on this slide are 10 microns apart  
(assume this has negligible uncertainty)

# Particle Tracking



Do describe briefly the settings used  
in the trackpy process  
(mass cut, radius, ...)

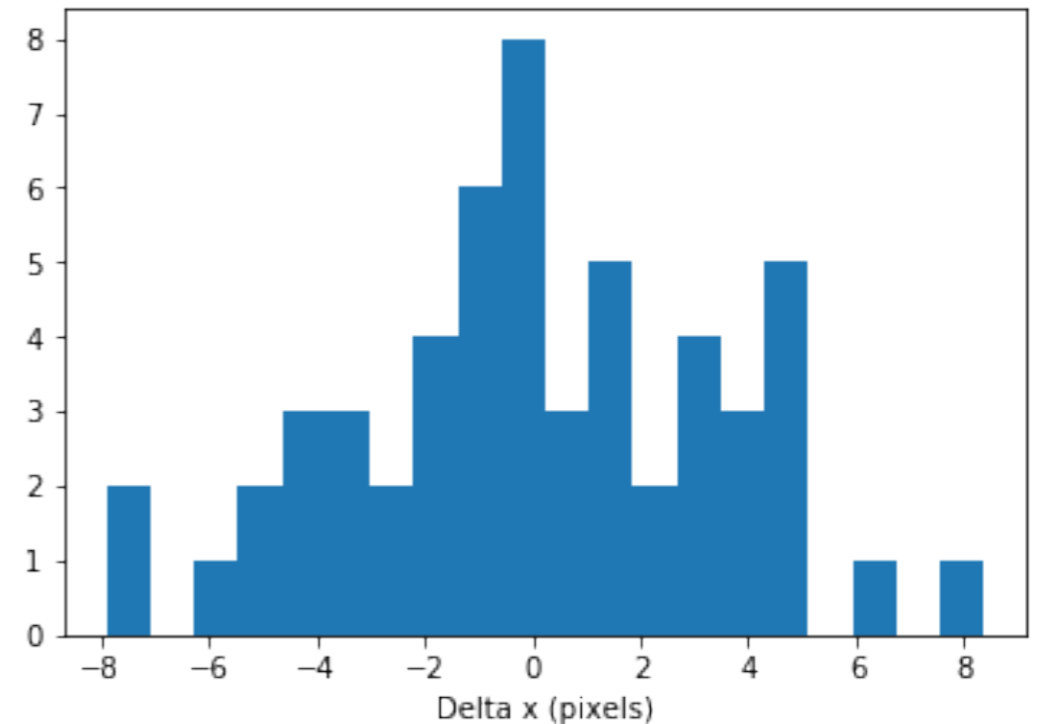
along with the movie you started from  
(bead diameter, frame rate, x40 or x100)

$(x_i, y_i)$  arrays

Your analysis  
starts here

# Consistency

$(x_i, y_i)$  arrays  
↓  
 $\Delta x_i$  and  $\Delta y_i$  arrays



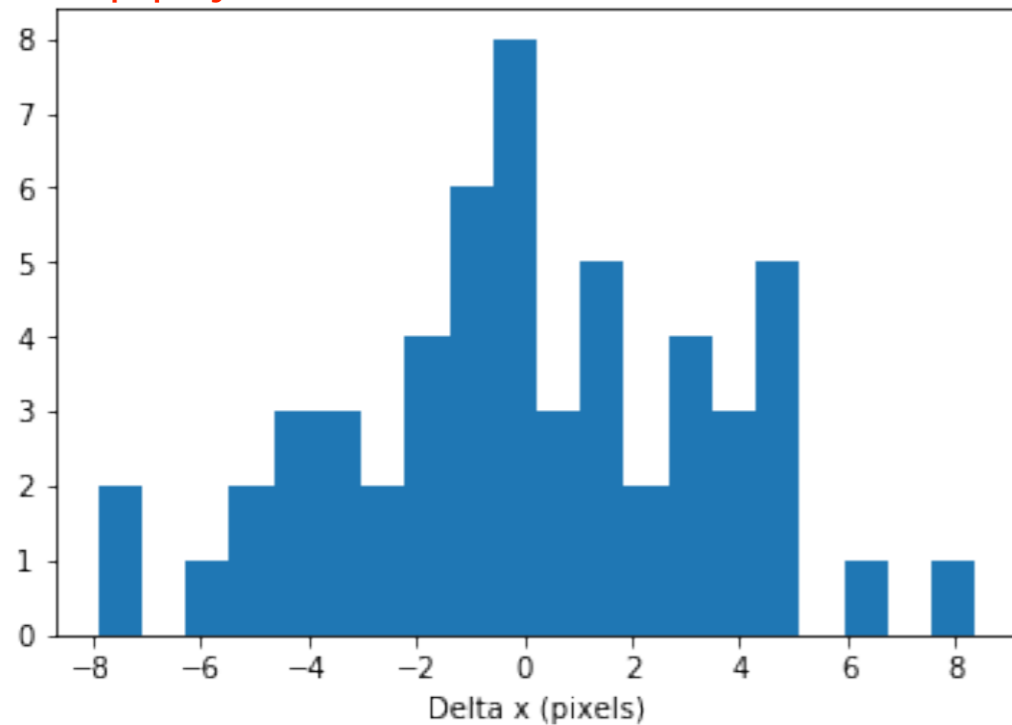
These  $\Delta x$  and  $\Delta y$  distributions should have  $\mu = 0$  and  $\sigma_x = \sigma_y$ . **Test this quantitatively!**

Remember, we talked about the stat error on the standard deviation! **Hwk 2 problem 5.28**

Don't worry if you are up to 2 sigma off...

# Diffusion

Apply calibration somewhere!



In each dimension:

$$D_x = \overline{\Delta x^2} / (2\Delta t)$$

But since mean is zero:

$$\overline{\Delta x^2} = \sigma_x^2$$

Variance is proportional to diffusion constant - **errors**

Note:  $D_x$  and  $D_y$  are measuring the same thing!

Either take average, or measure 2D diffusion:

$$D = (\sigma_x^2 + \sigma_y^2) / (4\Delta t)$$

# Boltzmann constant

- Diffusion Coefficient  $D$  is your primary experimental observable, make sure you state this clearly with (correct) **uncertainties**
- Can interpret this as the Boltzmann constant:

$$Df = k_B T \quad \text{where} \quad f = 6\pi\eta R$$

Beads have **diameters** of either 2.54  $\mu\text{m}$  or 1.06  $\mu\text{m}$

Error propagation problem, straight off HW1, **assumptions**

**Be careful with units!** You may be off by x2, if you are off by  $10^{12}$  you have forgotten to convert microns...

# Dependence on $\Delta t$

- Diffusion relation predicts variance  $\sigma^2 \sim \Delta t$

$$2\Delta t D_x = \overline{\Delta x^2}$$

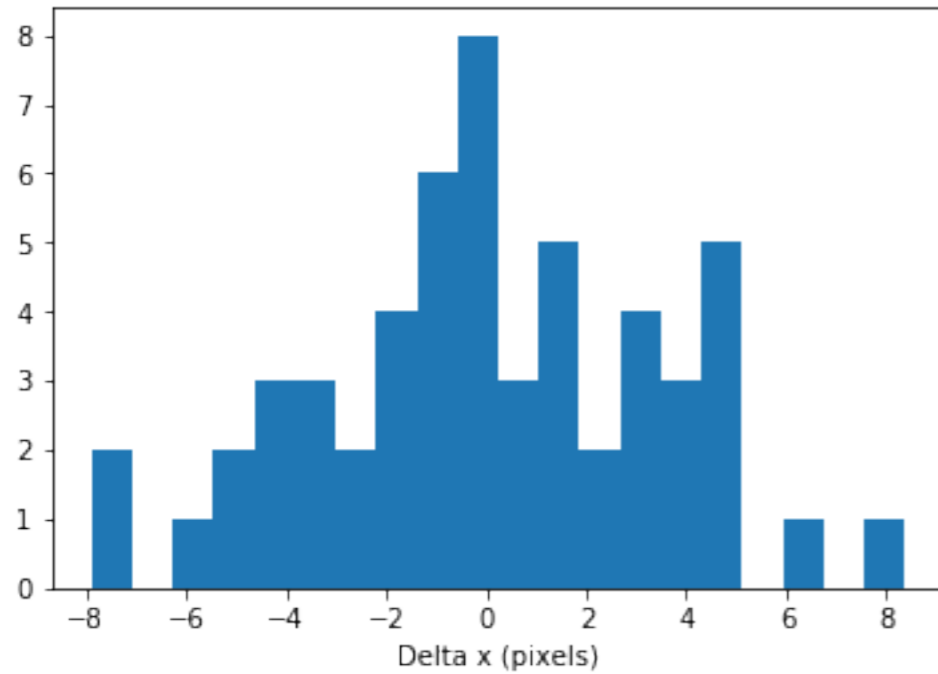
- Test this by changing  $\Delta t$ , and find  $\sigma^2$  (with error)
- Fit plot of  $\sigma^2$  vs.  $\Delta t$ , should show linear dependence, slope is  $2D$ , does this (qualitatively) agree with your previous value?

Don't need to go crazy here, just want to see that you can fit a line and get error out

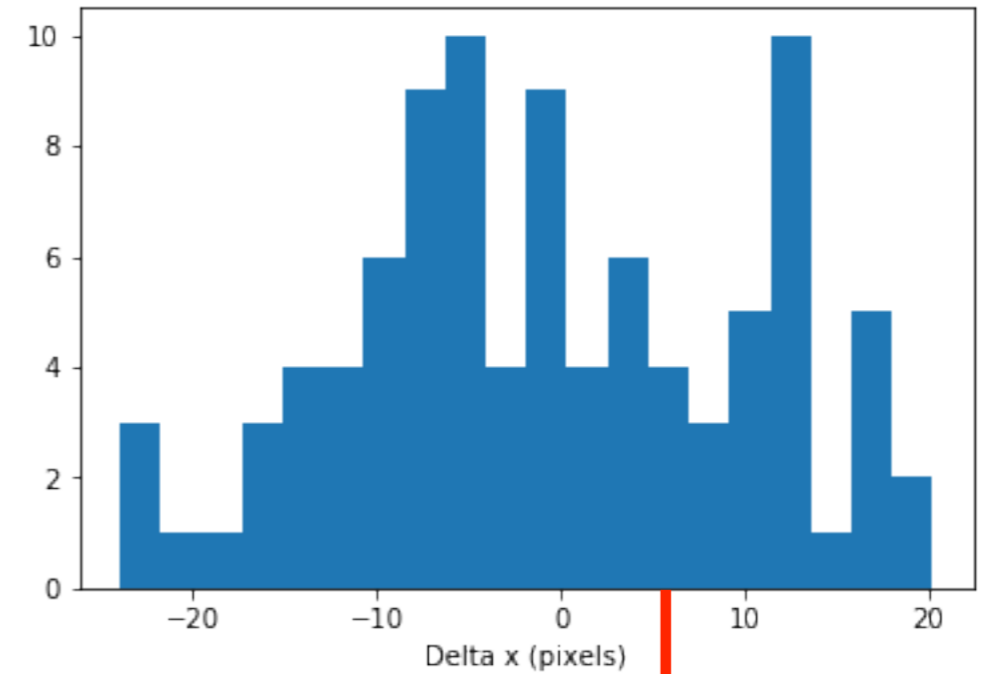


# Example

$\Delta t = 1\text{s}$

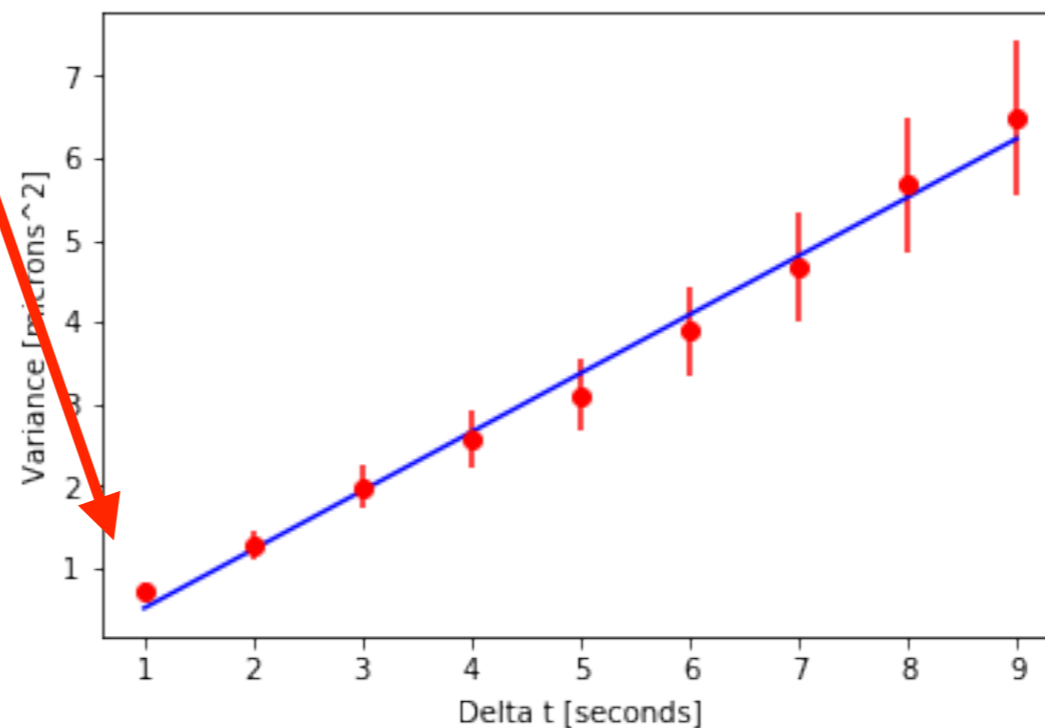


$\Delta t = 9\text{s}$



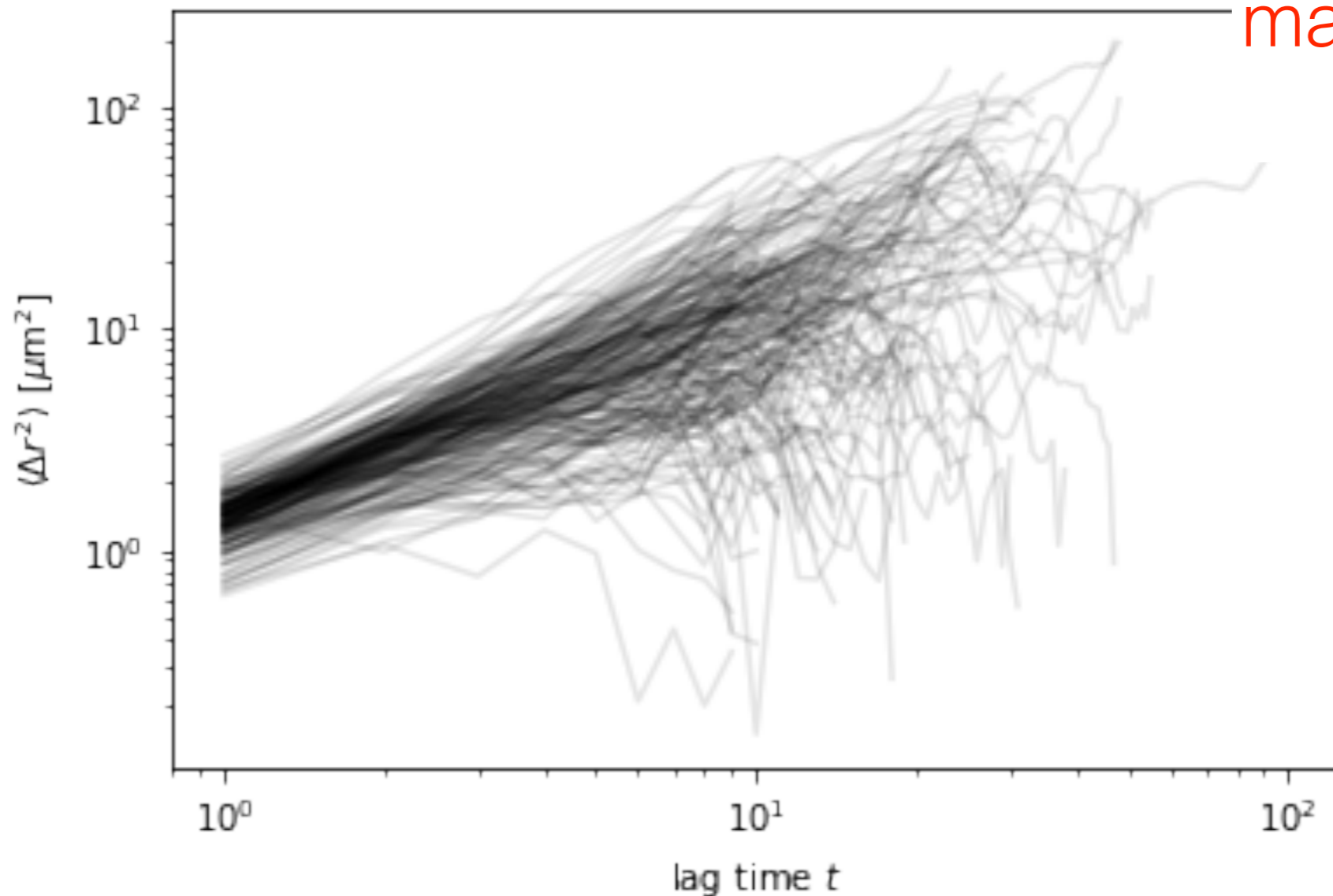
Find variance  
for different  $\Delta t$  values

variance vs.  $\Delta t$



# Your Data

You will not  
make this exact  
plot...



$\Delta t \lesssim 10\text{s}$   
should work

$\Delta r^2$  vs  $\Delta t$  for many trajectories  
Any individual path may be off, but on average,  
there is a nice linear relationship!

# Hypothesis Testing

# Question

- Some ropes are tested by hanging very heavy weights from them. For good ropes, 10% fail this test.
- After being used outside for a year, you test 5 ropes and find that 2 fail (40%). Is this significant evidence that sunlight is causing the ropes to fail?
- You test 5 more ropes and 2 more fail. Does this change your conclusion about the failure rate?

# Hypothesis Testing

- Frame the question as a null hypothesis
- Identify  $n$  (trials) and  $v$  (“successes”),  
for null hypothesis, identify  $p$  (expected “success” per trial)
- Check if you can use Gaussian approximation:  
Is  $np$  and  $n(1-p) \geq 10$ ?  
If so, Gaussian w/  $\mu = np$ ,  $\sigma^2 = np(1-p)$  works fine...
- Find statistical probability (P value) of observed  $v$  events or **worse** given  $n$  and  $p$ .
- Compare P value to arbitrary thresholds\*

# Binomial Distribution

$$B_{n,p}(\nu) = \frac{n!}{\nu!(n-\nu)!} p^\nu (1-p)^{(n-\nu)}$$

n - number of trials

p - probability of success / trial

v - number of successes

<https://stattrek.com/online-calculator/binomial.aspx>

Google: **binomial calculator**