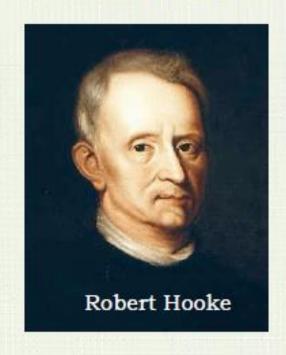
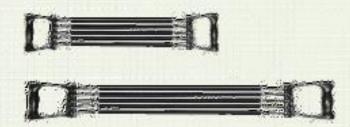
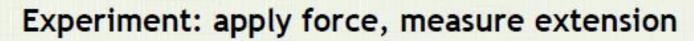
# A Common Pattern in Science and Engineering

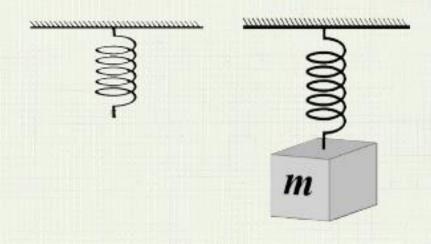
### Hypothesis: Hooke's Law for Springs



"The power of any springy body is in the same proportion with the extension."



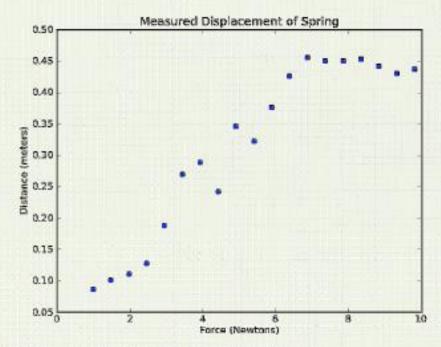




### **Process Observations**

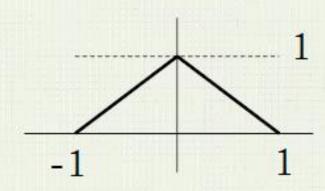
Distance(m)	Mass(kg)
0.0865	0.1
0.1015	0.15
0.1106	0.2
0.1279	0.25
0.4416	0.9
0.4304	0.95
0.437	1.0

### Hypothesis: linear relationship between F and x

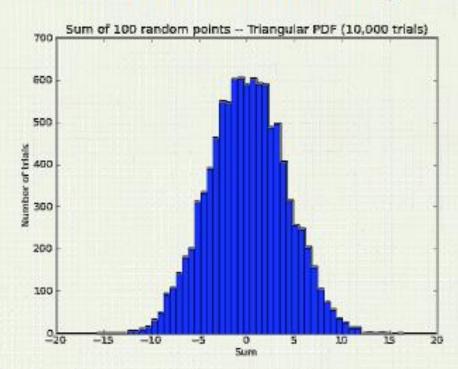


# observation; = prediction; ± error

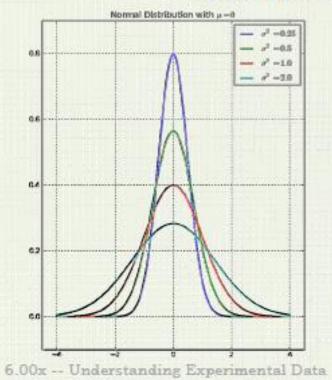
# Accumulation of many small random errors



# Accumulation of many small random errors



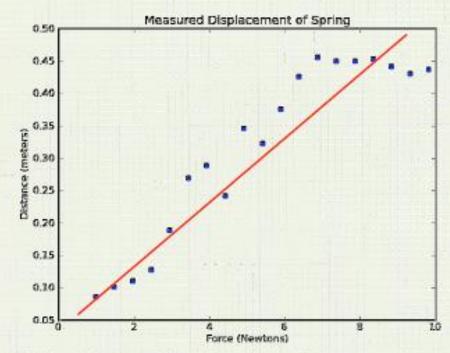
### The Normal distribution



### Gaussian model for observed errors

So when observation errors are due to the accumulation of many small random perturbations:

### Have: observations, Want: most likely line



# Log Likelihood

$$\begin{array}{l} \operatorname{Maximize} \prod_{i=0}^{len(obs)-1} L_{err}(obs_i - pred_i) \end{array}$$

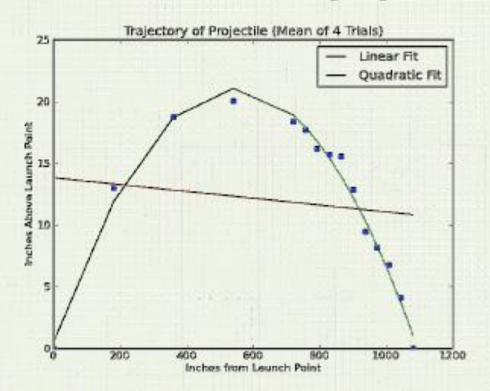
# Least Squares 6.00x -- Understanding Experimental Data

### pylab.polyfit(xvals, yvals, degree)

```
# find a, b that minimize
# sum((yvals-(a*xvals + b))**2)
a,b = pylab.polyfit(xvals,yvals,1)

# find a, b, c that minimize
# sum((yvals-(a*xvals**2 + b*xvals + c))**2)
a,b,c = pylab.polyfit(xvals,yvals,2)
```

# Measuring "goodness" of fit



### R2: Coefficient of Determination

Fraction of variability not explained by model

Fraction of variability explained by model

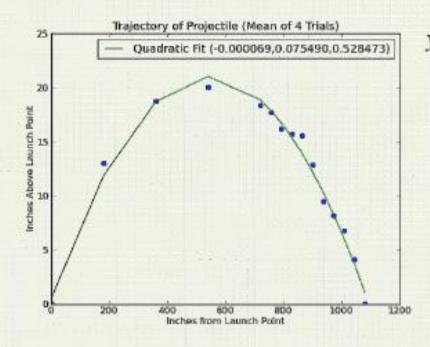
### Using a model: How thick a shield?



Use the model to make predictions when experiments are impractical or inadvisable.

Want to know the speed of the arrow as it reaches the target.

### Arrow speed when reaching target



$$y = (-.000069)x^2 + (.0755)x + .528$$