The Law of Large Numbers

Lecturer: John Guttag

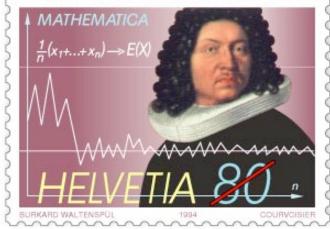
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Law of Large Numbers

Law of Large Numbers

In repeated **independent tests** with the same **actual probability** p of a particular outcome in each test, the chance that the **fraction of times** that outcome occurs differs from p converges to zero as the number

of trials goes to infinity.



Gambler's Fallacy

If deviations from expected behavior occur, these deviations are likely to be evened out by opposite deviations in the future.



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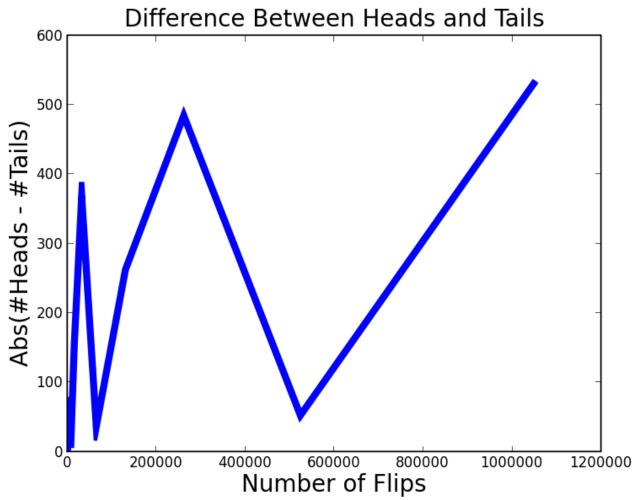
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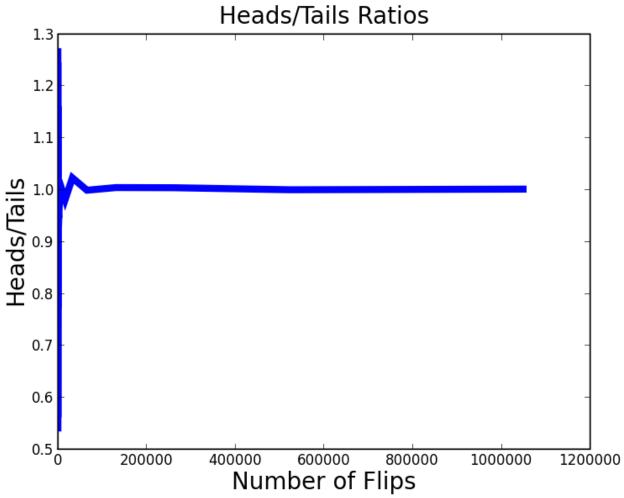
```
for numFlips in xAxis:
    numHeads = 0
    for n in range(numFlips):
        if random.random() < 0.5:
            numHeads += 1
    numTails = numFlips - numHeads
    ratios.append(numHeads/float(numTails))
    diffs.append(abs(numHeads - numTails))</pre>
```

. . .

```
pylab.title('Difference Between Heads and Tails')
    pylab.xlabel('Number of Flips')
    pylab.ylabel('Abs(#Heads - #Tails)')
    pylab.plot(xAxis, diffs)
    pylab.figure()
    pylab.title('Heads/Tails Ratios')
    pylab.xlabel('Number of Flips')
    pylab.ylabel('Heads/Tails')
    pylab.plot(xAxis, ratios)
```

6.00x





How Much Is Enough?

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6.00x

How Much Is Enough?



6.00x How Much Is Enough?

How Much is Enough?

How many samples do we need to look at in order to have a justified confidence that something that is true about the population of samples is also true about the population from which the samples were drawn?

Depends upon the variance in the underlying distribution

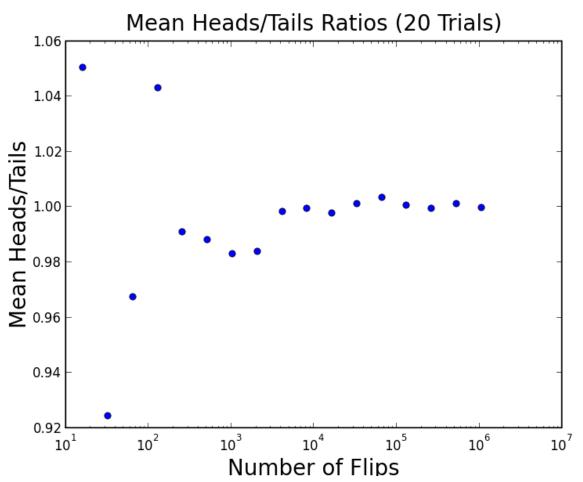
Variance

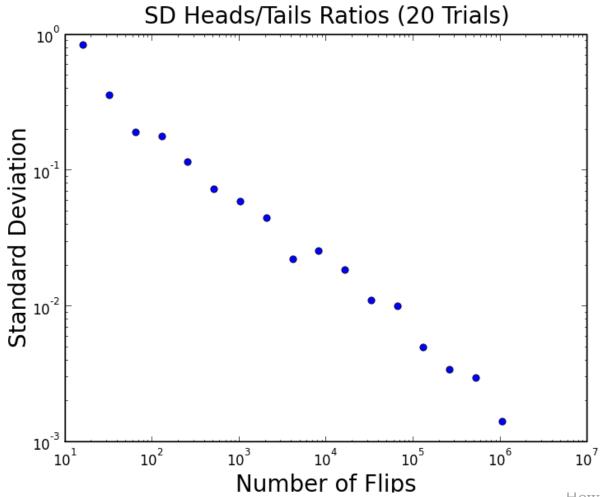
We measure the amount of variance in the outcomes of multiple trials.

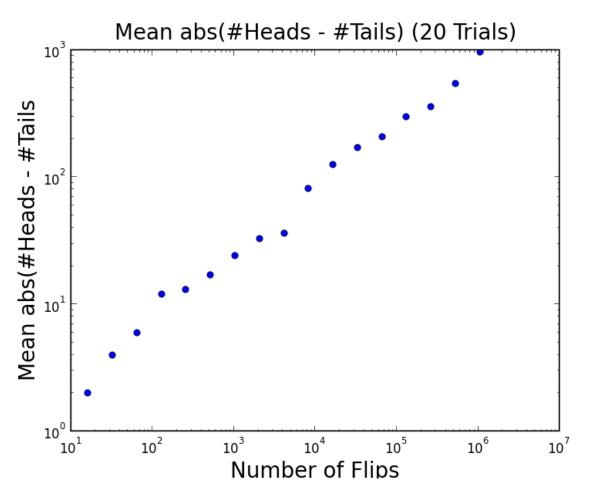
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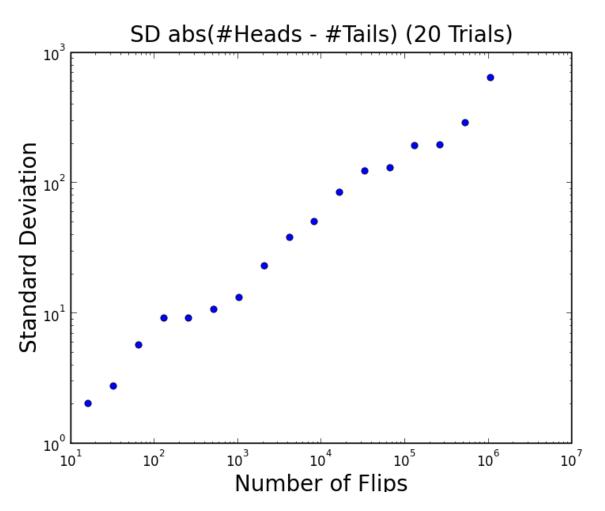
How Much Is Enough?

```
def stdDev(X):
    mean = sum(X)/float(len(X))
    tot = 0.0
    for x in X:
        tot += (x - mean)**2
    return (tot/len(X))**0.5
```



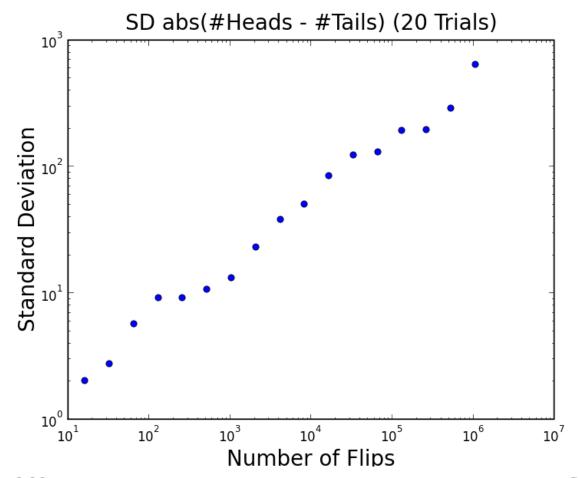






Standard Deviations and Histograms

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```
def CV(X):
    mean = sum(X)/float(len(X))
    try:
        return stdDev(X)/mean
    except ZeroDivisionError:
        return float('NaN')
```

```
def flip(numFlips):

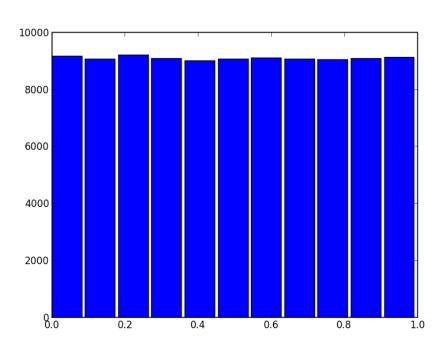
def flipSim(numFlipsPerTrial, numTrials):

def labelPlot(nf, nt, mean, sd):

def makePlots(nf1, nf2, nt):
   """nt = number of trials per experiment
   nf1 = number of flips 1st experiment
   nf2 = number of flips 2nd experiment"""
```

```
def makePlots(numFlips1, numFlips2, numTrials):
    val1, mean1, sd1 = flipSim(numFlips1, numTrials)
    pylab.hist(val1, bins = 20)
    xmin,xmax = pylab.xlim()
    ymin,ymax = pylab.ylim()
    labelPlot(numFlips1, numTrials, mean1, sd1)
    pylab.figure()
    val2, mean2, sd2 = flipSim(numFlips2, numTrials)
    pylab.hist(val2, bins = 20)
    pylab.xlim(xmin, xmax)
    ymin, ymax = pylab.ylim()
    labelPlot(numFlips2, numTrials, mean2, sd2)
```

```
vals = []
for i in range(100000):
    num = random.random()
    vals.append(num)
pylab.hist(vals, bins = 11)
```



Standard Deviations and Histograms

```
vals = []
for i in range(100000):
    num = random.random()
    vals.append(num)
pylab.hist(vals, bins = 11)
xmin, xmax = pylab.xlim()
ymin, ymax = pylab.ylim()
print 'x-range =', xmin, '-', xmax
print 'y-range =', ymin, '-', ymax
pylab.figure
pylab.hist(vals, bins = 11)
\#pylab.xlim(-1.0, 2.0)
```

