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In [1]: # DSC530-T302
         # Stephen Smitshoek
         # Week06
         # Exercise 8-1 and 8-2
In [2]: from __future__ import print_function, division
         import thinkstats2
         import thinkplot
         import math
         import random
         import numpy as np
        def MeanError(estimates, actual):
In [3]:
             """Computes the mean error of a sequence of estimates.
             estimate: sequence of numbers
             actual: actual value
             returns: float mean error
             errors = [estimate-actual for estimate in estimates]
             return np.mean(errors)
In [4]: def RMSE(estimates, actual):
             """Computes the root mean squared error of a sequence of estimates.
             estimate: sequence of numbers
             actual: actual value
             returns: float RMSE
             e2 = [(estimate-actual)**2 for estimate in estimates]
             mse = np.mean(e2)
             return math.sqrt(mse)
In [5]: def Estimate1(n=7, m=1000):
             """Evaluates RMSE of sample mean and median as estimators.
             n: sample size
             m: number of iterations
             \mathbf{m} \mathbf{m} \mathbf{m}
             mu = 0
             sigma = 1
             means = []
             medians = []
             for _ in range(m):
                 xs = [random.gauss(mu, sigma) for _ in range(n)]
                 xbar = np.mean(xs)
                 median = np.median(xs)
                 means.append(xbar)
                 medians.append(median)
             print('Experiment 1')
```

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print('mean error xbar', MeanError(means, mu))
            print('mean error median', MeanError(medians, mu))
In [6]: def Estimate2(n=7, m=1000):
            """Evaluates S and Sn-1 as estimators of sample variance.
            n: sample size
            m: number of iterations
            mu = 0
            sigma = 1
            estimates1 = []
            estimates2 = []
            for in range(m):
                xs = [random.gauss(mu, sigma) for _ in range(n)]
                biased = np.var(xs)
                unbiased = np.var(xs, ddof=1)
                estimates1.append(biased)
                estimates2.append(unbiased)
            print('Experiment 2')
            print('rmse biased', RMSE(estimates1, sigma**2))
            print('rmse unbiased', RMSE(estimates2, sigma**2))
In [7]: | def Estimate3(n=7, m=1000):
            """Evaluates L and Lm as estimators of the exponential parameter.
            n: sample size
            m: number of iterations
            lam = 2
            means = []
            medians = []
            for _ in range(m):
                xs = np.random.exponential(1/lam, n)
                L = 1 / np.mean(xs)
                Lm = math.log(2) / np.median(xs)
                means.append(L)
                medians.append(Lm)
            print('Experiment 3')
            print('rmse L', RMSE(means, lam))
            print('rmse Lm', RMSE(medians, lam))
            print('mean error L', MeanError(means, lam))
            print('mean error Lm', MeanError(medians, lam))
In [8]: def SimulateSample(mu=90, sigma=7.5, n=9, m=1000):
            """Plots the sampling distribution of the sample mean.
            mu: hypothetical population mean
            sigma: hypothetical population standard deviation
            n: sample size
            m: number of iterations
            def VertLine(x, y=1):
                thinkplot.Plot([x, x], [0, y], color='0.8', linewidth=3)
```

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means = []
              for _ in range(m):
                  xs = np.random.normal(mu, sigma, n)
                  xbar = np.mean(xs)
                  means.append(xbar)
              stderr = RMSE(means, mu)
              print('standard error', stderr)
              cdf = thinkstats2.Cdf(means)
              ci = cdf.Percentile(5), cdf.Percentile(95)
              print('confidence interval', ci)
              VertLine(ci[0])
              VertLine(ci[1])
              # plot the CDF
              thinkplot.Cdf(cdf)
              thinkplot.Save(root='estimation1',
                             xlabel='sample mean',
                             ylabel='CDF',
                             title='Sampling distribution')
 In [9]: def simulate experiment(n=10, lam=2, m=1000):
              def VertLine(x):
                  thinkplot.Plot([x, x], [0, 1], color='0.8', linewidth=3)
              L_estimates = []
              for _ in range(m):
                  xs = np.random.exponential(1.0/lam, n)
                  L = 1 / np.mean(xs)
                  L estimates.append(L)
              cdf = thinkstats2.Cdf(L estimates)
              ci = cdf.Percentile(5), cdf.Percentile(95)
              stderr = RMSE(L estimates, lam)
              print('The 90% Confidence Interval is {} to {}'.format(round(ci[0], 2), round(ci[1
              print('The standard error is {}'.format(round(stderr, 2)))
              VertLine(ci[0])
              VertLine(ci[1])
              thinkplot.Cdf(cdf)
In [10]: | print("m=10")
          Estimate1(m=10)
          print("\nm=10,000")
          Estimate1(m=100000)
          print("\nBoth the mean and median trend towards zero as the number of experiements inc
               "\nlikley that neither of them are biased.")
```

m = 10

Experiment 1

mean error xbar 0.13822209624935694 mean error median 0.25140235875528316

m=10,000

Experiment 1

mean error xbar 0.002117238023748525 mean error median 0.00301923665666006

Both the mean and median trend towards zero as the number of experiements increases, therefore it is

likley that neither of them are biased.

## In [11]: Estimate2()

print("\nBoth S^2 and S(n-1)^2 yield similar results with S^2 being slightly smaller.

Experiment 2

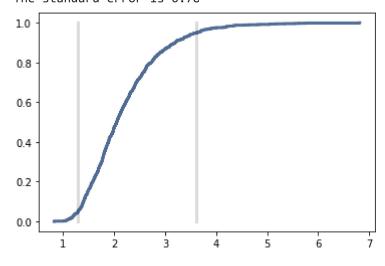
rmse biased 0.5008320018577254 rmse unbiased 0.5570055288254314

Both  $S^2$  and  $S(n-1)^2$  yield similar results with  $S^2$  being slightly smaller.

## In [12]:

simulate experiment(n=10)

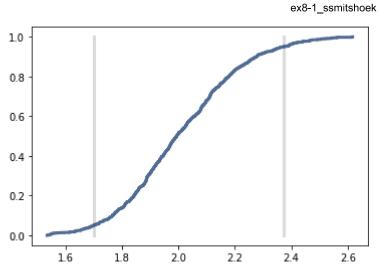
The 90% Confidence Interval is 1.3 to 3.62 The standard error is 0.78



## In [13]:

simulate\_experiment(n=100)

The 90% Confidence Interval is 1.7 to 2.37 The standard error is 0.2



simulate\_experiment(n=1000) In [14]:

The 90% Confidence Interval is 1.9 to 2.1 The standard error is 0.06

