## Ch 2 Bootstrap confidence intervals

November-13-17 9:21 AM

```
#Visualizing bootstrap samples
for _ in range(50):
  # Generate bootstrap sample: bs_sample
  bs sample = np.random.choice(rainfall, size=len(rainfall))
  # Compute and plot ECDF from bootstrap sample
  x, y = ecdf (bs_sample)
  _ = plt.plot(x, y, marker='.', linestyle='none',
         color='gray', alpha=0.1)
# Compute and plot ECDF from original data
x, y = ecdf (rainfall)
_ = plt.plot(x, y, marker='.')
# Make margins and label axes
plt.margins(0.02)
_ = plt.xlabel('yearly rainfall (mm)')
_ = plt.ylabel('ECDF')
# Show the plot
plt.show()
```

#Generating many bootstrap replicates
def draw\_bs\_reps(data, func, size=1):
 """Draw bootstrap replicates."""

# Initialize array of replicates: bs\_replicates
bs\_replicates = np.empty (size)

# Generate replicates
for i in range(size):
 bs\_replicates[i] = bootstrap\_replicate\_1d(data, func)
return bs\_replicates

#Bootstrap replicates of the mean and the SEM # Take 10,000 bootstrap replicates of the mean: bs\_replicates bs\_replicates = draw\_bs\_reps (rainfall, np.mean, 10000)

# Compute and print SEM sem = np.std(rainfall) / np.sqrt(len(rainfall)) print(sem)

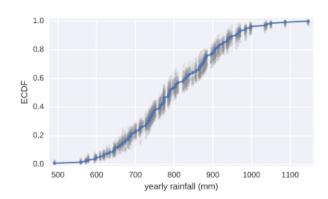
# Compute and print standard deviation of bootstrap replicates bs\_std = np.std (bs\_replicates) print(bs\_std)

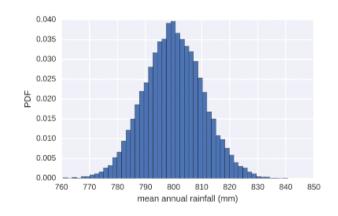
# Make a histogram of the results
\_ = plt.hist(bs\_replicates, bins=50, normed=True)
\_ = plt.xlabel('mean annual rainfall (mm)')
\_ = plt.ylabel('PDF')

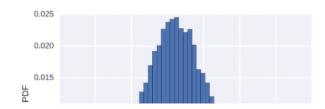
# Show the plot plt.show()

Conf\_int = np.percentile (bs\_replicates, [2.5, 97.5])

#Bootstrap replicates of other statistics
# Generate 10,000 bootstrap replicates of the variance: bs\_replicates
bs\_replicates = draw\_bs\_reps\_(rainfall\_np\_var\_10000)







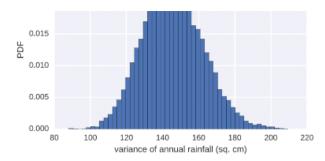
#Bootstrap replicates of other statistics
# Generate 10,000 bootstrap replicates of the variance: bs\_replicates
bs\_replicates = draw\_bs\_reps (rainfall, np.var, 10000)

# Put the variance in units of square centimeters bs\_replicates = bs\_replicates/100

# Make a histogram of the results

- \_ = plt.hist(bs\_replicates, normed = True, bins = 50)
- \_ = plt.xlabel('variance of annual rainfall (sq. cm)')
- \_ = plt.ylabel('PDF')

# Show the plot plt.show()



#Confidence interval on the rate of no-hitters
# Draw bootstrap replicates of the mean no-hitter time (equal to tau):
bs\_replicates
bs\_replicates = draw\_bs\_reps (nohitter\_times, np.mean, 10000)

# Compute the 95% confidence interval: conf\_int conf\_int = np.percentile (bs\_replicates, [2.5, 97.5])

# Print the confidence interval print('95% confidence interval =', conf\_int, 'games')

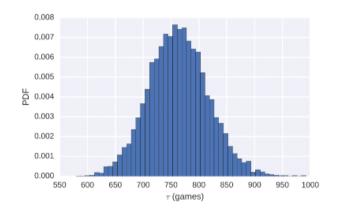
# Plot the histogram of the replicates \_ = plt.hist(bs\_replicates, bins=50, normed=True)

\_ = plt.xlabel(r'\$\tau\$ (games)')

#A function to do pairs bootstrap

\_ = plt.ylabel('PDF')

# Show the plot plt.show()



def draw\_bs\_pairs\_linreg(x, y, size=1):
 """Perform pairs bootstrap for linear regression."""

# Set up array of indices to sample from: inds
 inds = np.arange(len(x))

# Initialize replicates: bs\_slope\_reps, bs\_intercept\_reps
 bs\_slope\_reps = np.empty (size)
 bs\_intercept\_reps = np.empty (size)

# Generate replicates
for i in range(size):
 bs\_inds = np.random.choice(inds, size=len (inds))
 bs\_x, bs\_y = x[bs\_inds], y[bs\_inds]
 bs\_slope\_reps[i], bs\_intercept\_reps[i] = np.polyfit (bs\_x, bs\_y, 1)

#Pairs bootstrap of literacy/fertility data
# Generate replicates of slope and intercept using pairs bootstrap
bs\_slope\_reps, bs\_intercept\_reps = draw\_bs\_pairs\_linreg (illiteracy,
fertility, 1000)

# Compute and print 95% CI for slope print(np.percentile(bs\_slope\_reps, [2.5, 97.5]))

return bs\_slope\_reps, bs\_intercept\_reps

# Plot the histogram



```
# Compute and print 35% circl slope
print(np.percentile(bs_slope_reps, [2.5, 97.5]))

# Plot the histogram
_ = plt.hist(bs_slope_reps, bins=50, normed=True)
_ = plt.xlabel('slope')
_ = plt.ylabel('PDF')
plt.show()
```

