# Ch 4 Thinking probabilistically-- Continuous variables

November-12-17 10:58 PM

## #The Normal PDF

# Draw 100000 samples from Normal distribution with stds of interest: samples std1, samples std3, samples std10 #mean = 20, std = 1, 3, 10

samples std1 = np.random.normal(20, 1, size = 100000)

samples std3 = np.random.normal (20, 3, size = 100000)

samples\_std10 = np.random.normal (20, 10, size = 100000)

## # Make histograms

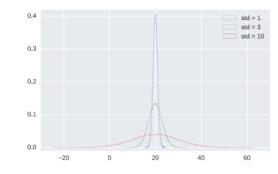
- \_=plt.hist (samples\_std1, bins = 100, normed=True, histtype = 'step')
- \_=plt.hist (samples\_std3, bins = 100, normed=True, histtype = 'step')
- \_=plt.hist (samples\_std10, bins = 100, normed = True, histtype = 'step')

## # Make a legend, set limits and show plot

= plt.legend(('std = 1', 'std = 3', 'std = 10'))

plt.ylim(-0.01, 0.42)

plt.show()



## #The Normal CDF

# Generate CDFs

x\_std1, y\_std1 = ecdf (samples\_std1)

x\_std3, y\_std3 = ecdf (samples\_std3)

x\_std10, y\_std10 = ecdf (samples\_std10)

\_=plt.plot (x\_std1, y\_std1, marker = ".", linestyle = 'none') \_=plt.plot (x\_std3, y\_std3, marker = ".", linestyle = 'none')

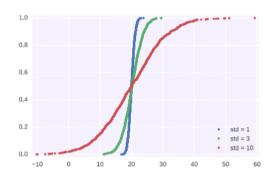
=plt.plot (x\_std10, y\_std10, marker = ".", linestyle = 'none')

# # Make 2% margin

plt.margins(0.02)

### # Make a legend and show the plot

= plt.legend(('std = 1', 'std = 3', 'std = 10'), loc='lower right') plt.show()



# #Are the Belmont Stakes results Normally distributed? # Compute mean and standard deviation: mu, sigma

mu = np.mean (belmont\_no\_outliers)

sigma = np.std (belmont\_no\_outliers)

# Sample out of a normal distribution with this mu and sigma: samples samples = np.random.normal (mu, sigma, size = 10000)

# # Get the CDF of the samples and of the data

x\_theor, y\_theor = ecdf (samples)

x, y = ecdf (belmont\_no\_outliers)

## # Plot the CDFs and show the plot

\_ = plt.plot(x\_theor, y\_theor)

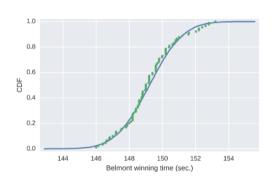
= plt.plot(x, y, marker='.', linestyle='none')

plt.margins(0.02)

\_ = plt.xlabel('Belmont winning time (sec.)')

= plt.ylabel('CDF')

plt.show()



#What are the chances of a horse matching or beating Secretariat's record? # Take a million samples out of the Normal distribution: samples samples = np.random.normal (mu, sigma, size = 1000000)

# Compute the fraction that are faster than 144 seconds: prob prob = np.sum (samples <= 144)/len(samples)

# Print the result

print('Probability of besting Secretariat:', prob)

#If you have a story, you can simulate it!

def successive\_poisson(tau1, tau2, size=1): # Draw samples out of first exponential distribution: t1 t1 = np.random.exponential (tau1, size)

# Draw samples out of second exponential distribution: t2 t2 = np.random.exponential (tau2, size)

return t1 + t2

#Distribution of no-hitters and cycles
# Draw samples of waiting times: waiting\_times
waiting\_times = successive\_poisson (764, 715, 100000)

# Make the histogram \_=plt.hist(waiting\_times, bins = 100, normed=True, histtype = 'step')

# Label axes \_=plt.xlabel ('Waiting Time (number of games)') \_=plt.ylabel ('PDF')

# Show the plot plt.show ()

