

# Statistical Thinking in Python Part 1

## 4). Thinking probabilistically-- Continuous variables

### a). The Normal PDF

```
# Draw 100000 samples from Normal distribution with stds of interest: samples_std1, samples_std3, samples_std10
```

```
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')
```

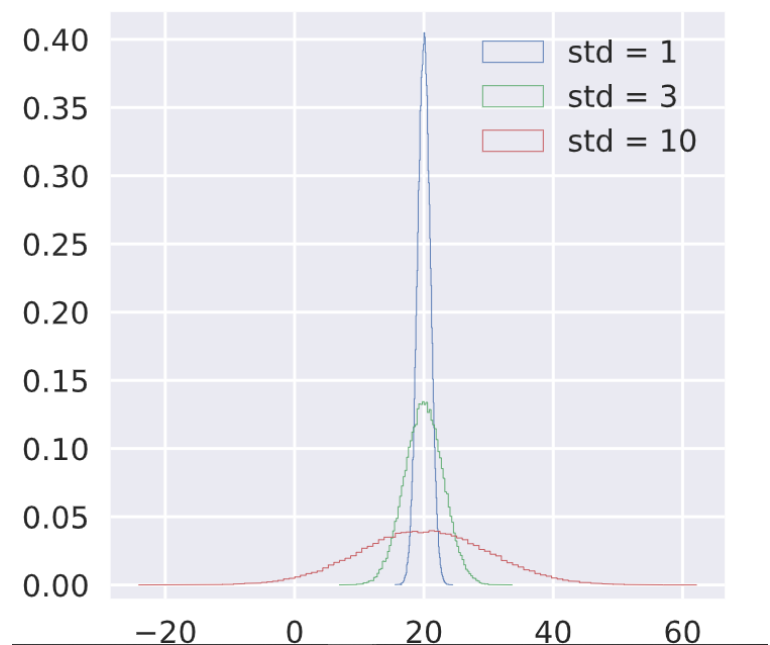
```
_ = plt.hist(samples_std10, bins=100, normed=True, histtype='step')
```

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

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

```
plt.ylim(-0.01, 0.42)
```

```
plt.show()
```



**b). 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)
```

```
# Plot CDFs
```

```
plt.plot(x_std1, y_std1, marker='.', linestyle='none')
```

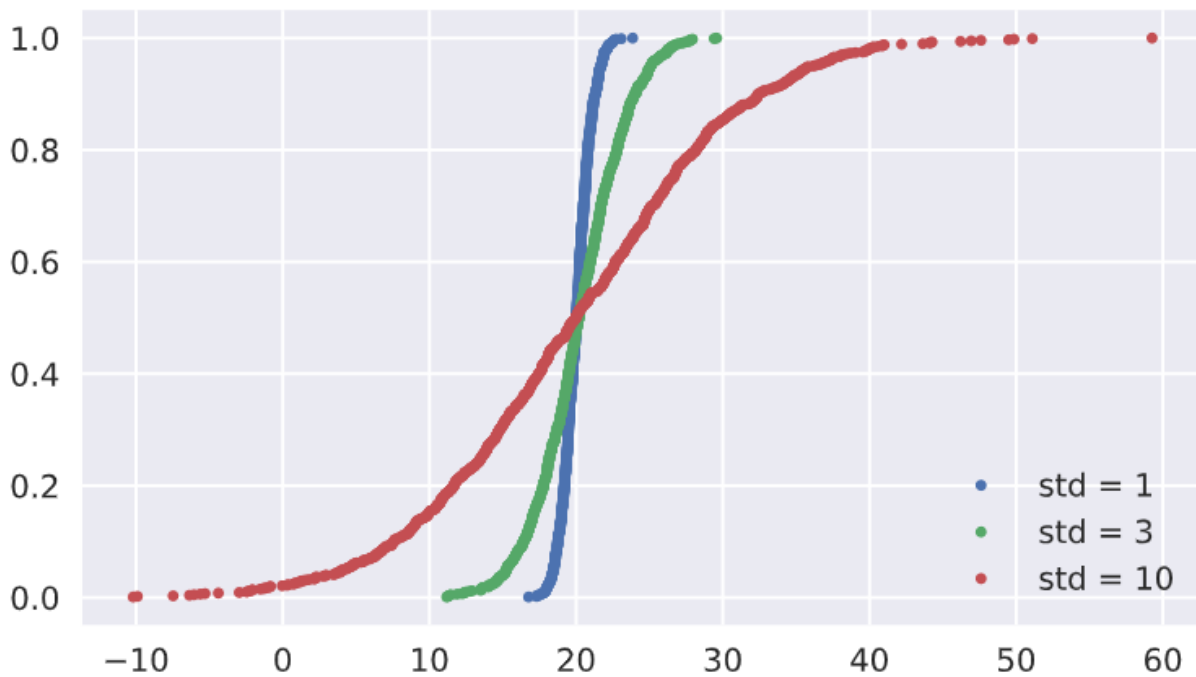
```
plt.plot(x_std3, y_std3, marker='.', linestyle='none')
```

```
plt.plot(x_std10, y_std10, marker='.', linestyle='none')
```

```
# Make a legend and show the plot
```

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

```
plt.show()
```



**c). 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
```

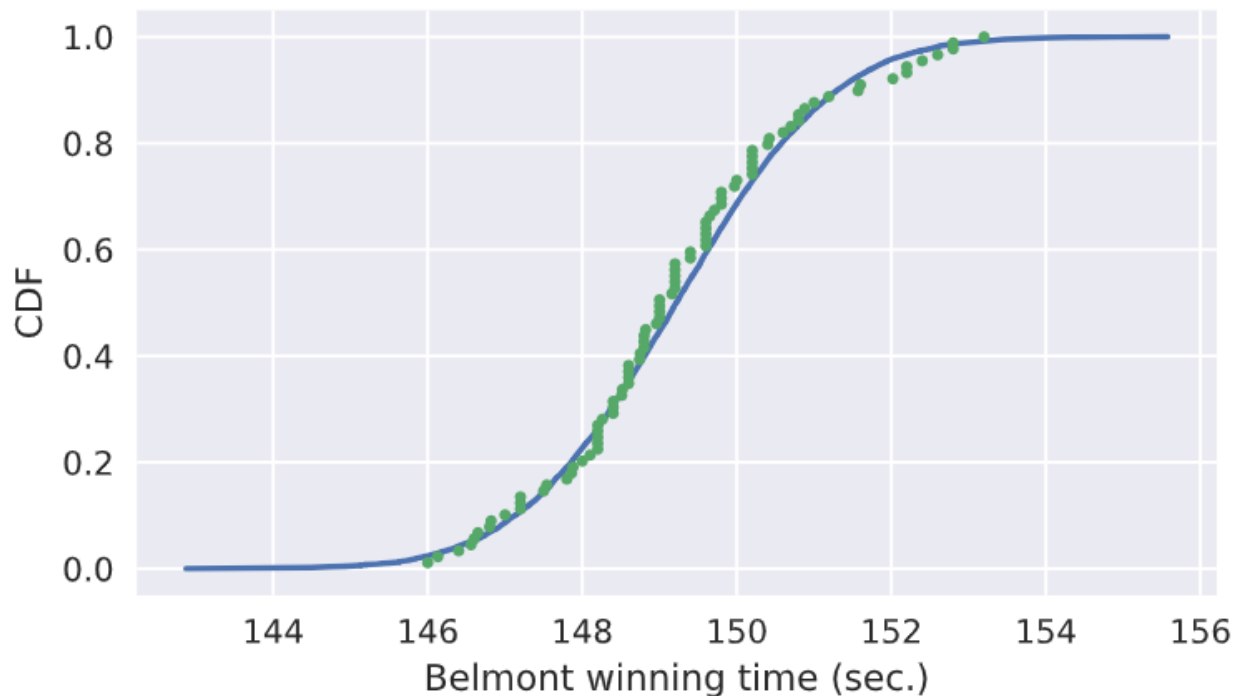
```
_ = plt.plot(x_theor, y_theor)
```

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

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

```
_ = plt.ylabel('CDF')
```

```
plt.show()
```



**d). 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)**

**<script.py> output:**

**Probability of besting Secretariat: 0.000635**

e). **If you have a story, you can simulate it!**

```
def successive_poisson(tau1, tau2, size=1):  
    """Compute time for arrival of 2 successive Poisson processes."""  
    # Draw samples out of first exponential distribution: t1  
    t1 = np.random.exponential(tau1, size=size)  
  
    # Draw samples out of second exponential distribution: t2  
    t2 = np.random.exponential(tau2, size=size)  
  
    return t1 + t2
```

**f). Distribution of no-hitters and cycles**

```
# Draw samples of waiting times: waiting_times
```

```
waiting_times= successive_poisson(764, 715, size=100000)
```

```
# Make the histogram
```

```
plt.hist(waiting_times, bins=100, normed=True, histtype='step')
```

```
# Label axes
```

```
plt.xlabel('X')
```

```
plt.ylabel('Y')
```

```
# Show the plot
```

```
plt.show()
```

