

# Probability density functions

STATISTICAL THINKING IN PYTHON (PART 1)



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# Continuous variables

- Quantities that can take any value, not just discrete values

# Michelson's speed of light experiment



Image: public domain, Smithsonian

Data: Michelson, 1880

# Michelson's speed of light experiment



measured speed of light (1000 km/s)

```
299.85 299.74 299.90 300.07 299.93  
299.85 299.95 299.98 299.98 299.88  
300.00 299.98 299.93 299.65 299.76  
299.81 300.00 300.00 299.96 299.96  
299.96 299.94 299.96 299.94 299.88  
299.80 299.85 299.88 299.90 299.84  
299.83 299.79 299.81 299.88 299.88  
299.83 299.80 299.79 299.76 299.80  
299.88 299.88 299.88 299.86 299.72  
299.72 299.62 299.86 299.97 299.95  
299.88 299.91 299.85 299.87 299.84  
299.84 299.85 299.84 299.84 299.84  
299.89 299.81 299.81 299.82 299.80  
299.77 299.76 299.74 299.75 299.76  
299.91 299.92 299.89 299.86 299.88  
299.72 299.84 299.85 299.85 299.78  
299.89 299.84 299.78 299.81 299.76  
299.81 299.79 299.81 299.82 299.85  
299.87 299.87 299.81 299.74 299.81  
299.94 299.95 299.80 299.81 299.87
```

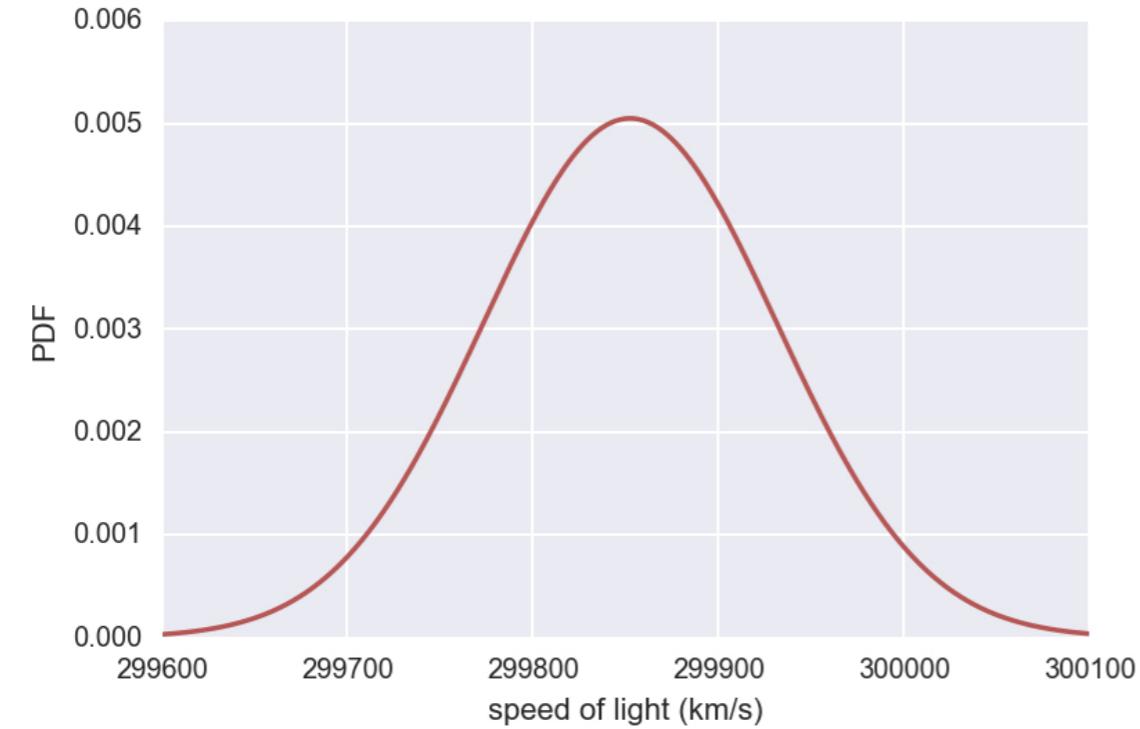
Image: public domain, Smithsonian

Data: Michelson, 1880

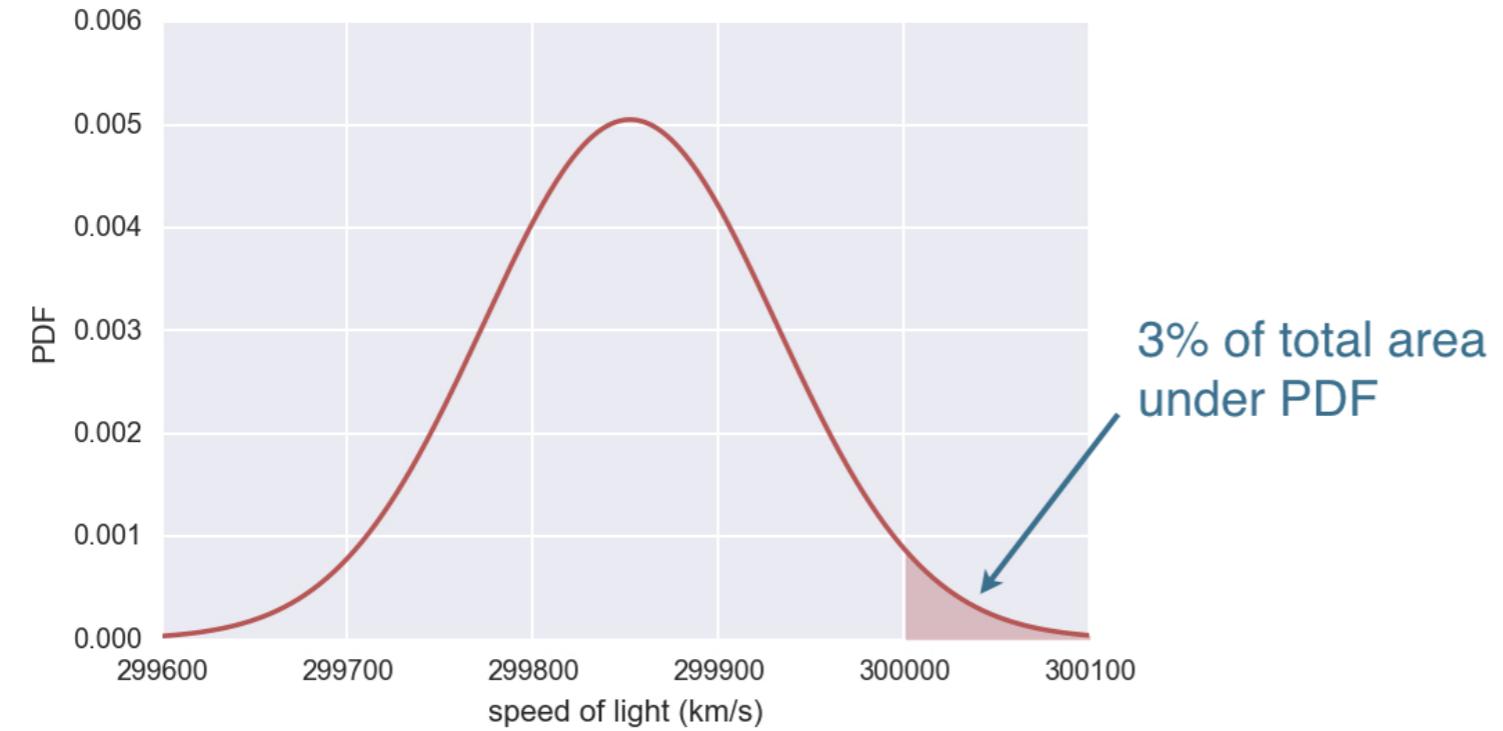
# Probability density function (PDF)

- Continuous analog to the PMF
- Mathematical description of the relative likelihood of observing a value of a continuous variable

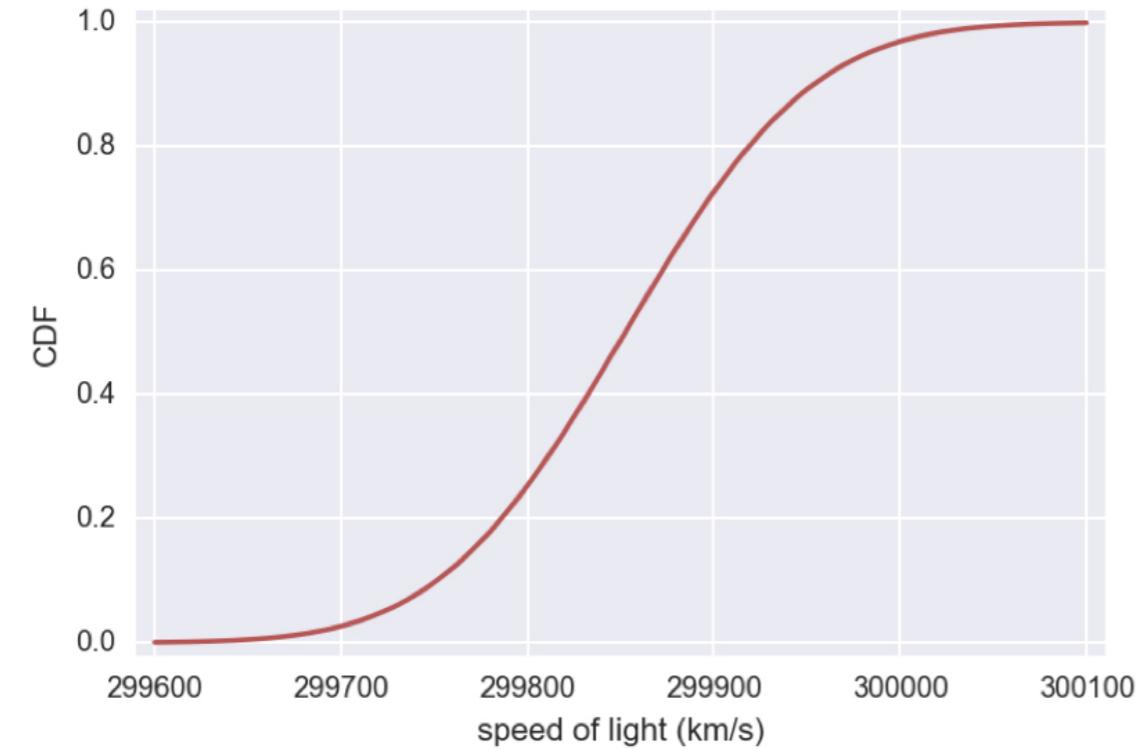
# Normal PDF



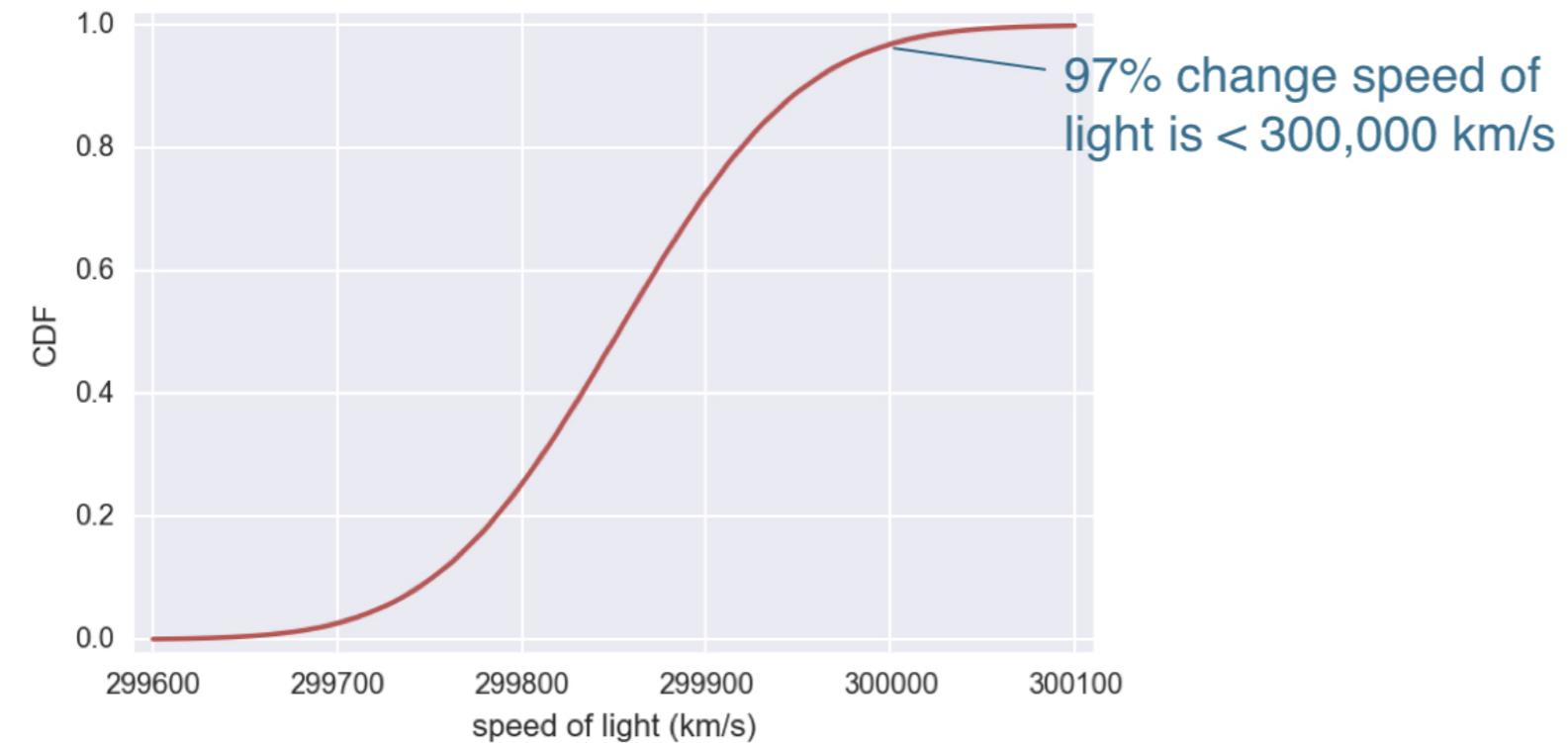
# Normal PDF



# Normal CDF



# Normal CDF

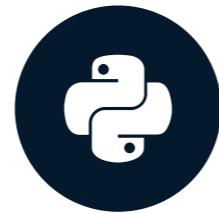


# **Let's practice!**

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# Introduction to the Normal distribution

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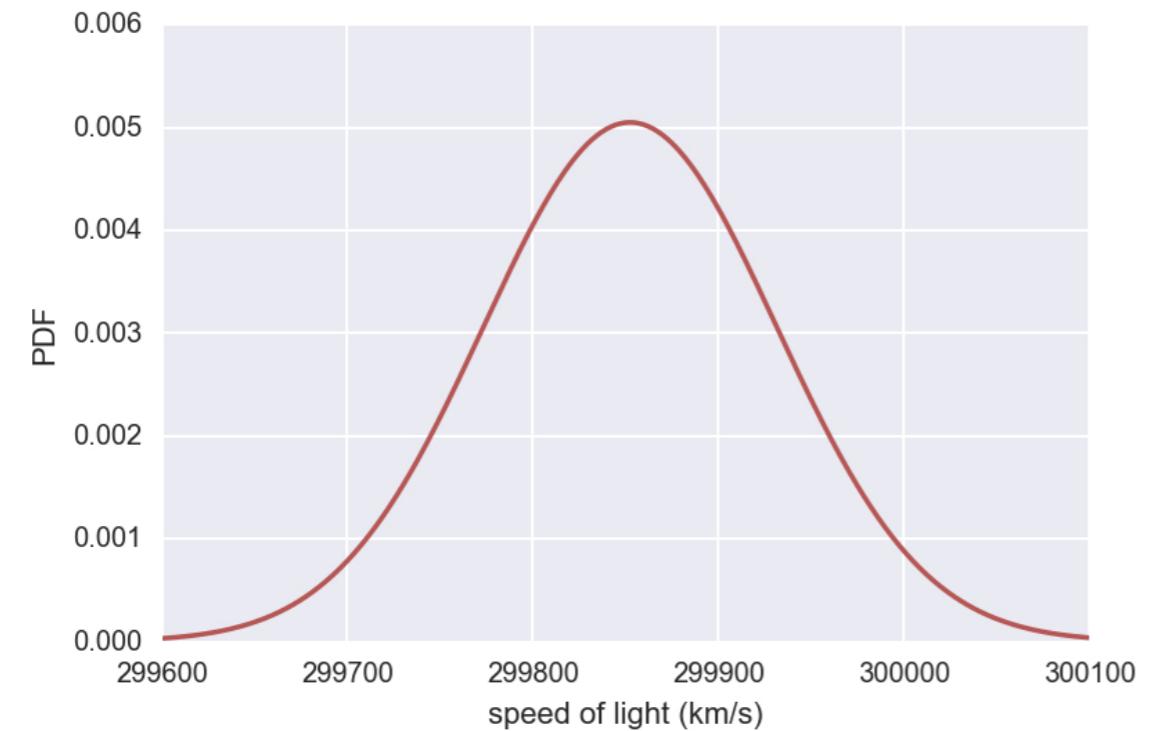
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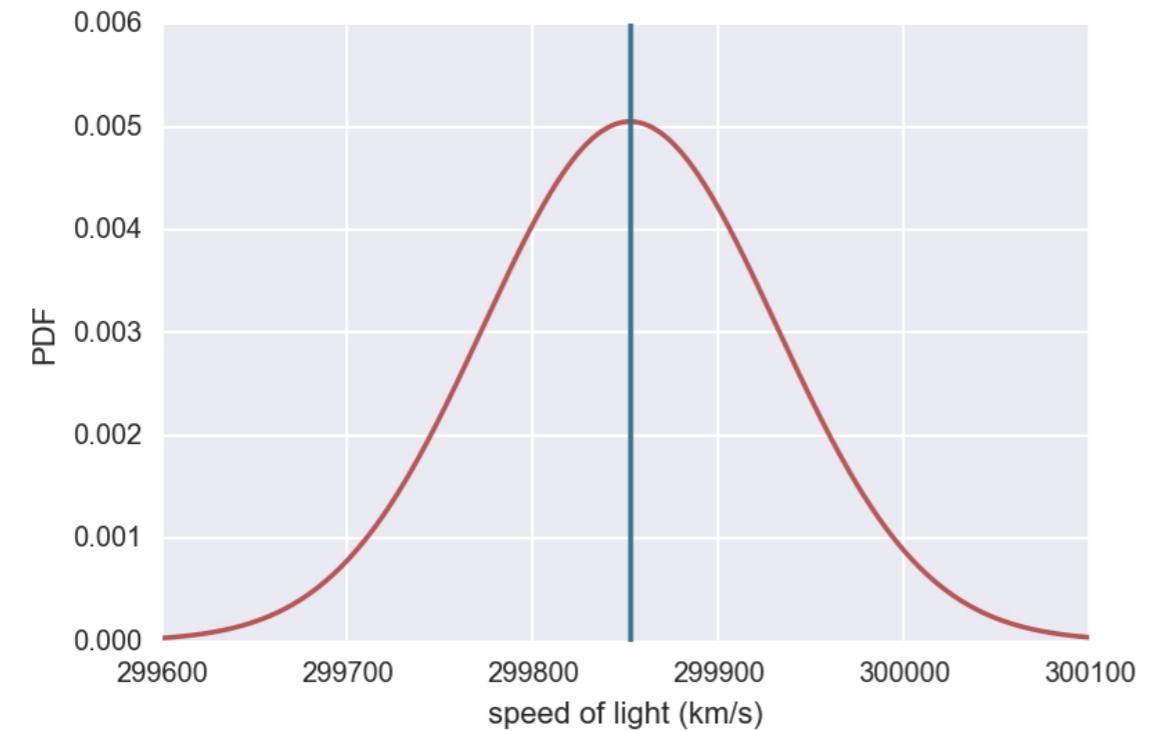
# Normal distribution

- Describes a continuous variable whose PDF has a single symmetric peak.

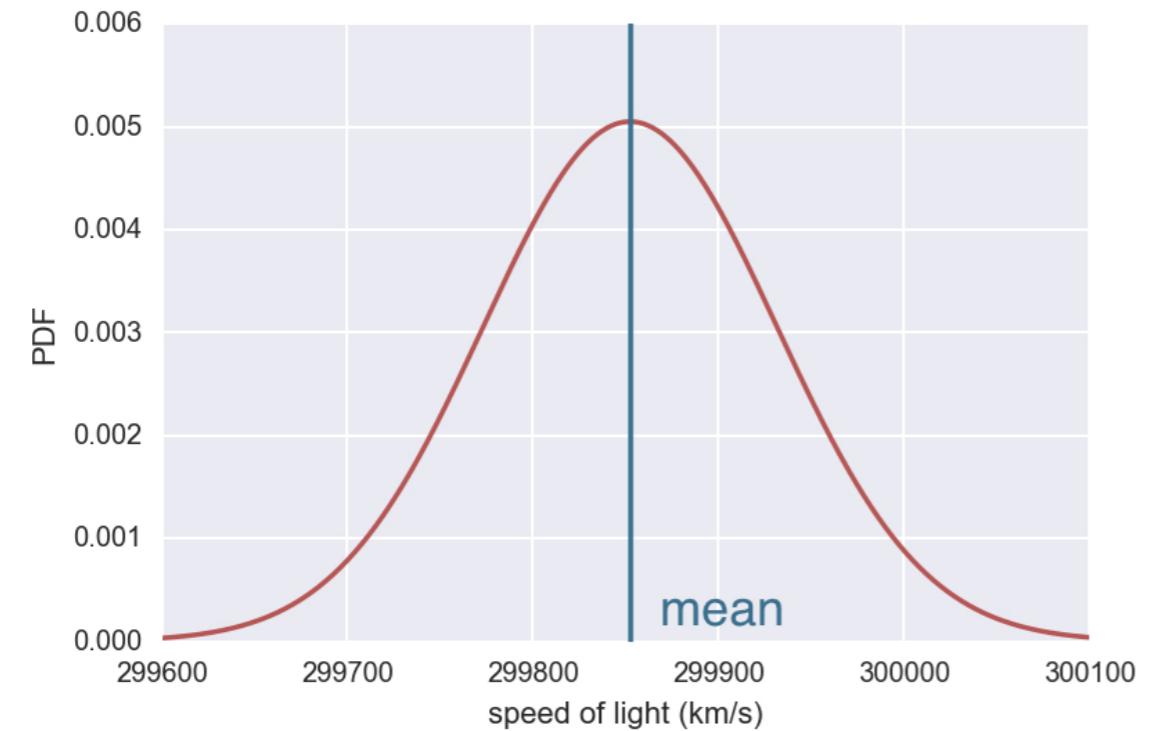
# Normal distribution



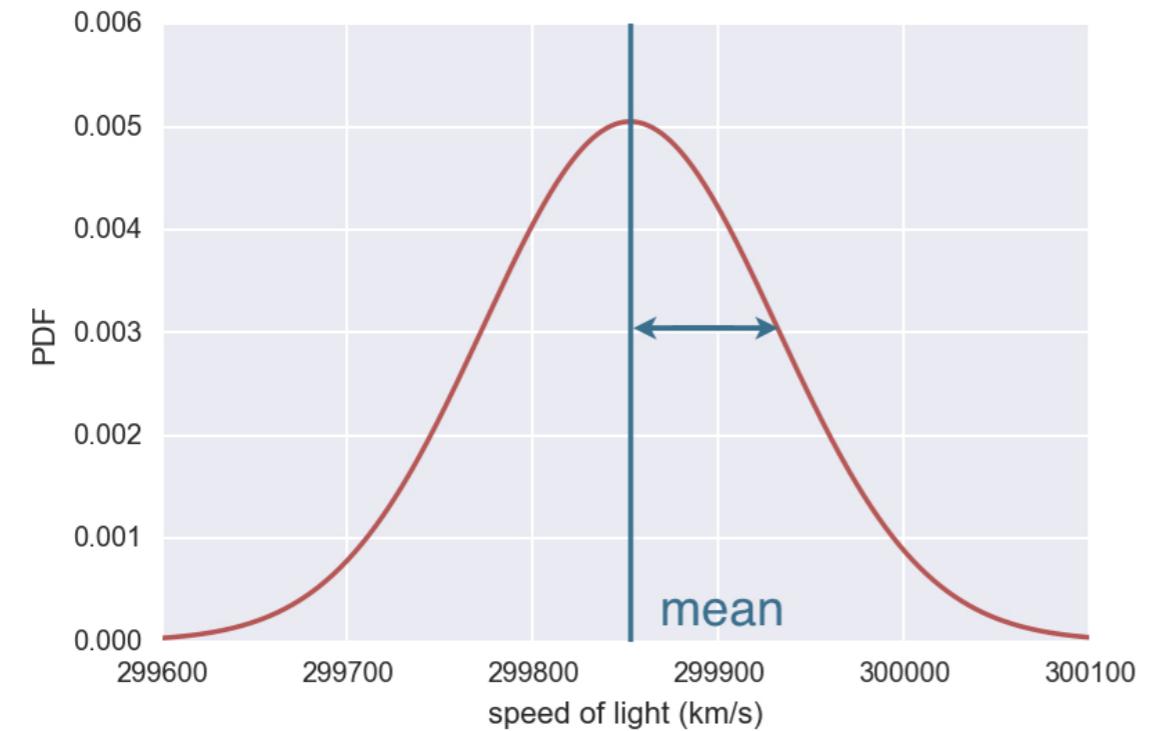
# Normal distribution



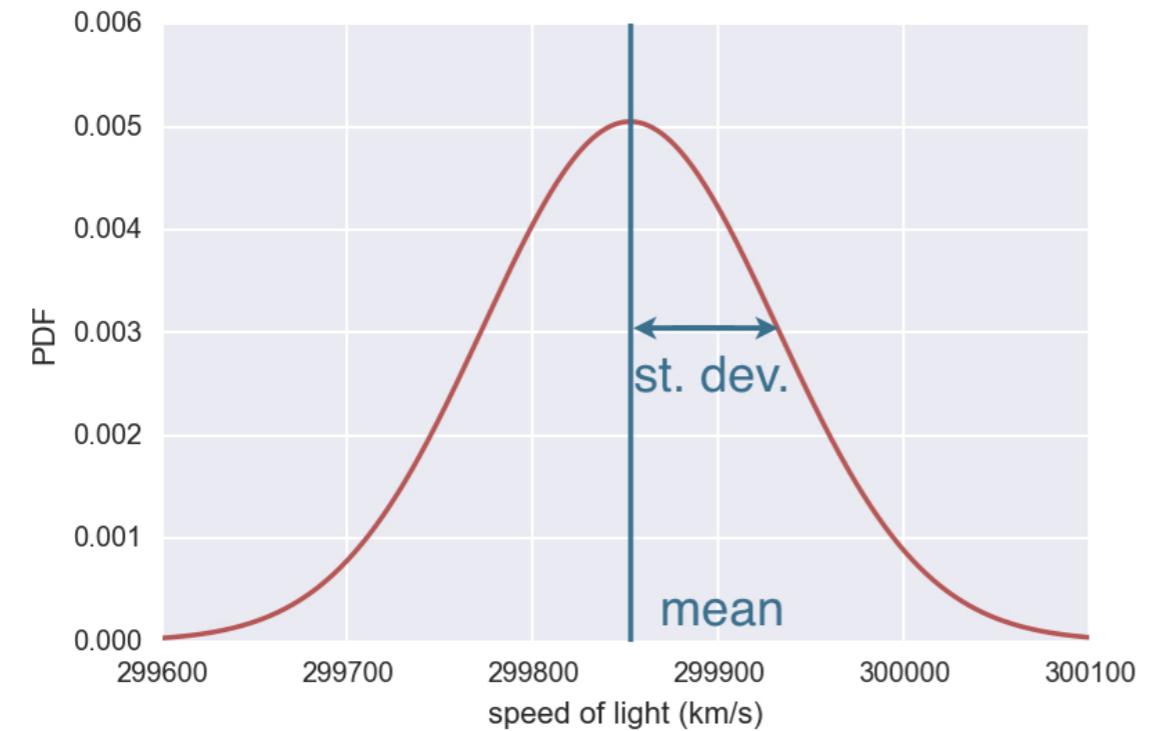
# Normal distribution



# Normal distribution



# Normal distribution



## Parameter

mean of a  
Normal distribution

$\neq$

## Calculated from data

mean computed  
from data

## Parameter

mean of a  
Normal distribution

$\neq$

## Calculated from data

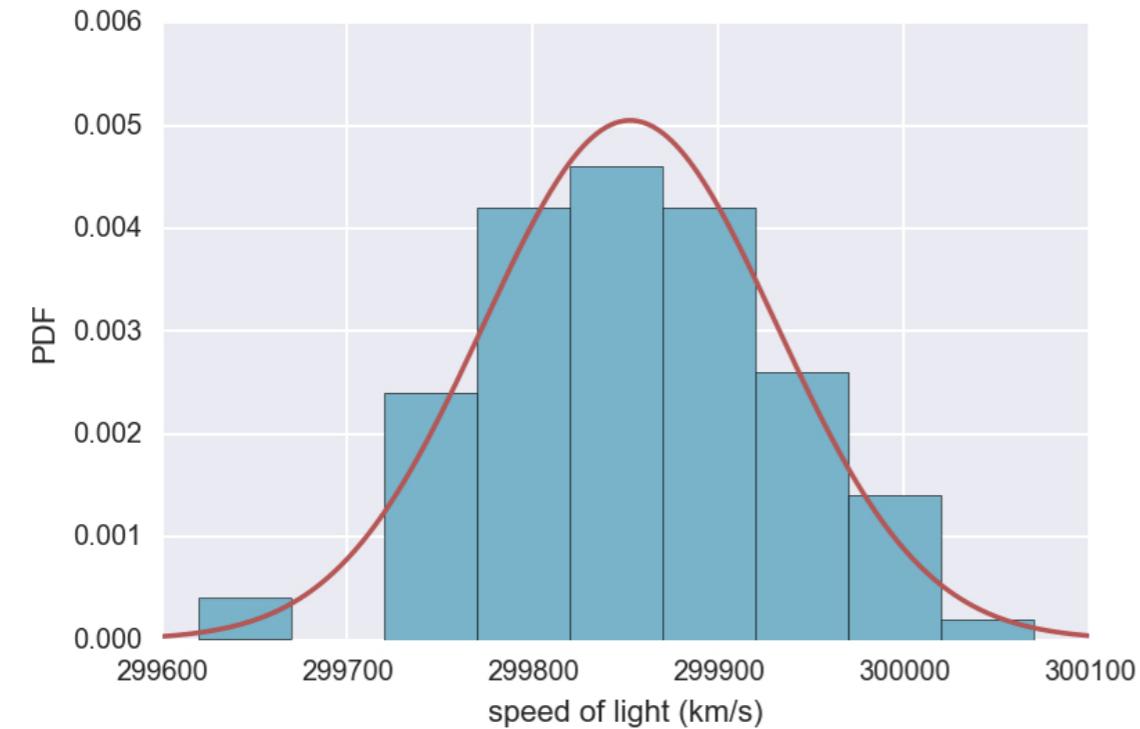
st. dev. of a  
Normal distribution

$\neq$

mean computed  
from data

standard deviation  
computed from data

# Comparing data to a Normal PDF



Data: Michelson, 1880

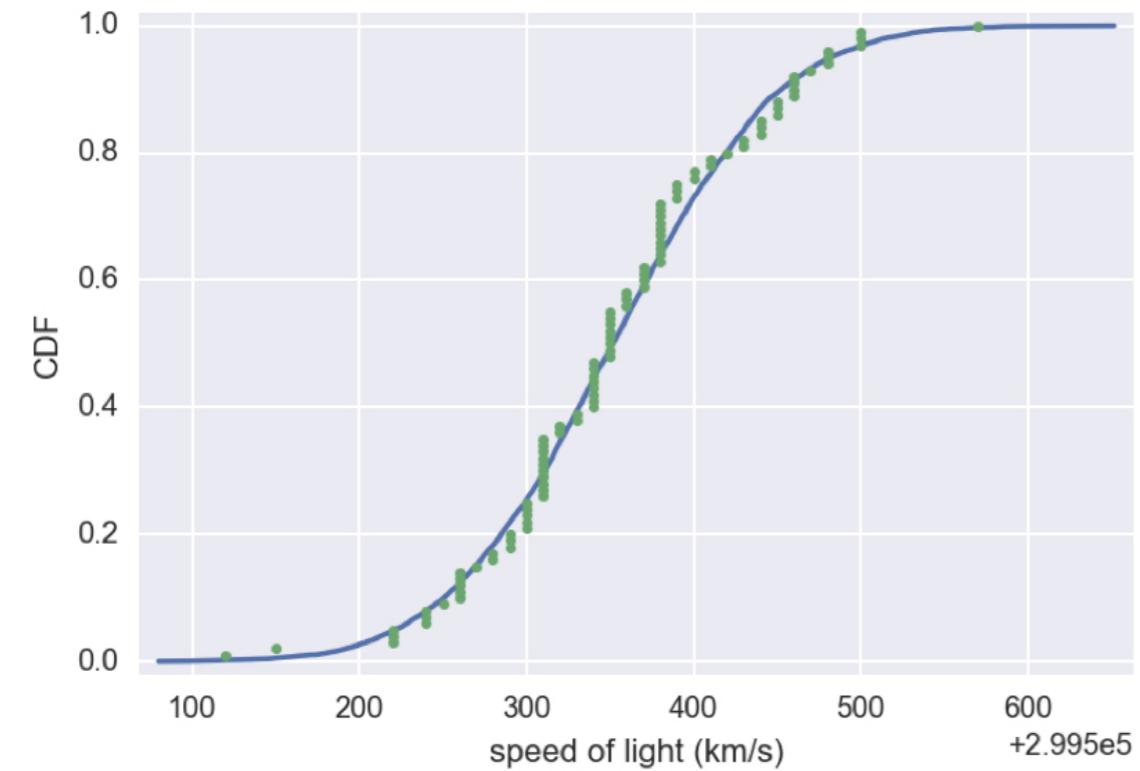
# Checking Normality of Michelson data

```
import numpy as np  
  
mean = np.mean(michelson_speed_of_light)  
std = np.std(michelson_speed_of_light)  
samples = np.random.normal(mean, std, size=10000)  
x, y = ecdf(michelson_speed_of_light)  
x_theor, y_theor = ecdf(samples)
```

# Checking Normality of Michelson data

```
import matplotlib.pyplot as plt  
import seaborn as sns  
  
sns.set()  
  
_ = plt.plot(x_theor, y_theor)  
_ = plt.plot(x, y, marker='.', linestyle='none')  
_ = plt.xlabel('speed of light (km/s)')  
_ = plt.ylabel('CDF')  
plt.show()
```

# Checking Normality of Michelson data



Data: Michelson, 1880

# **Let's practice!**

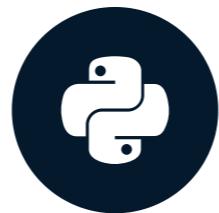
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# The Normal distribution: Properties and warnings

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# Image: Deutsche Bundesbank



Image: Deutsche Bundesbank

# The Gaussian distribution

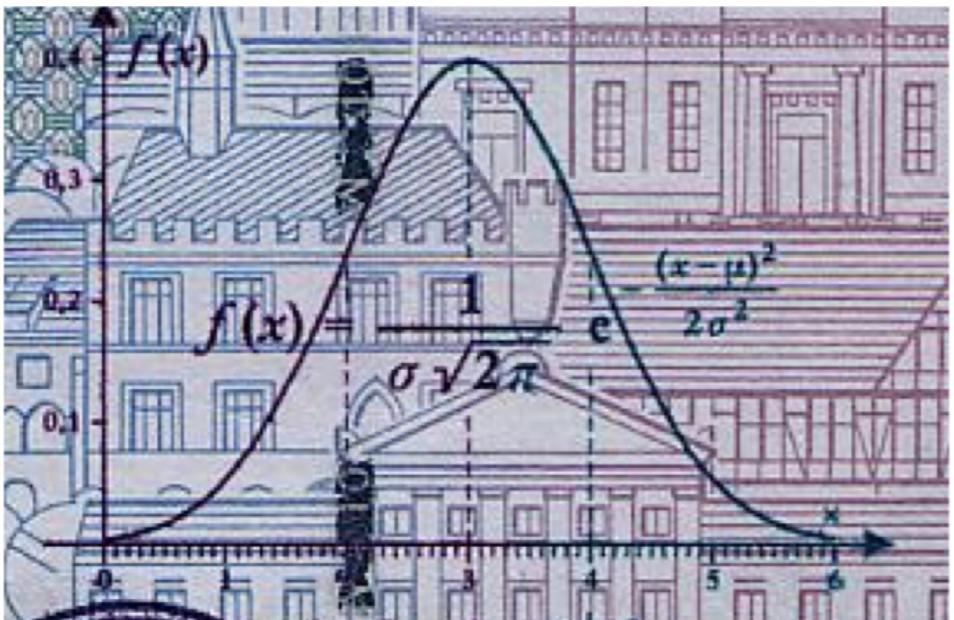
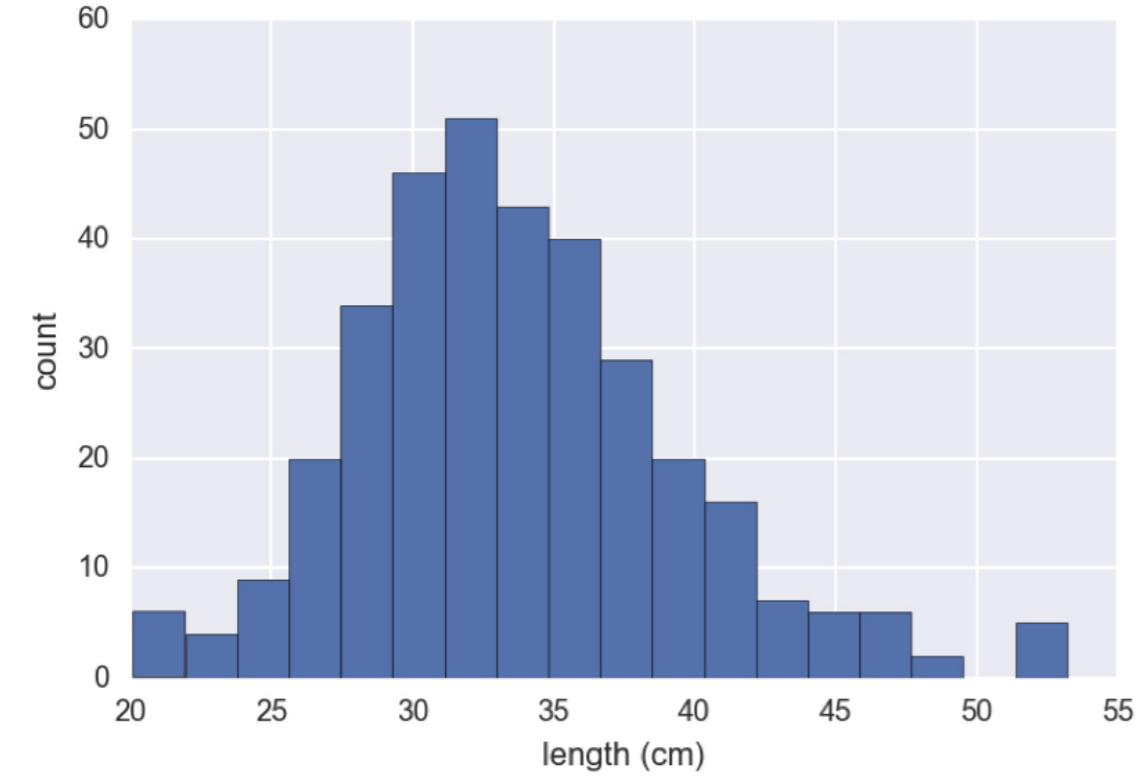


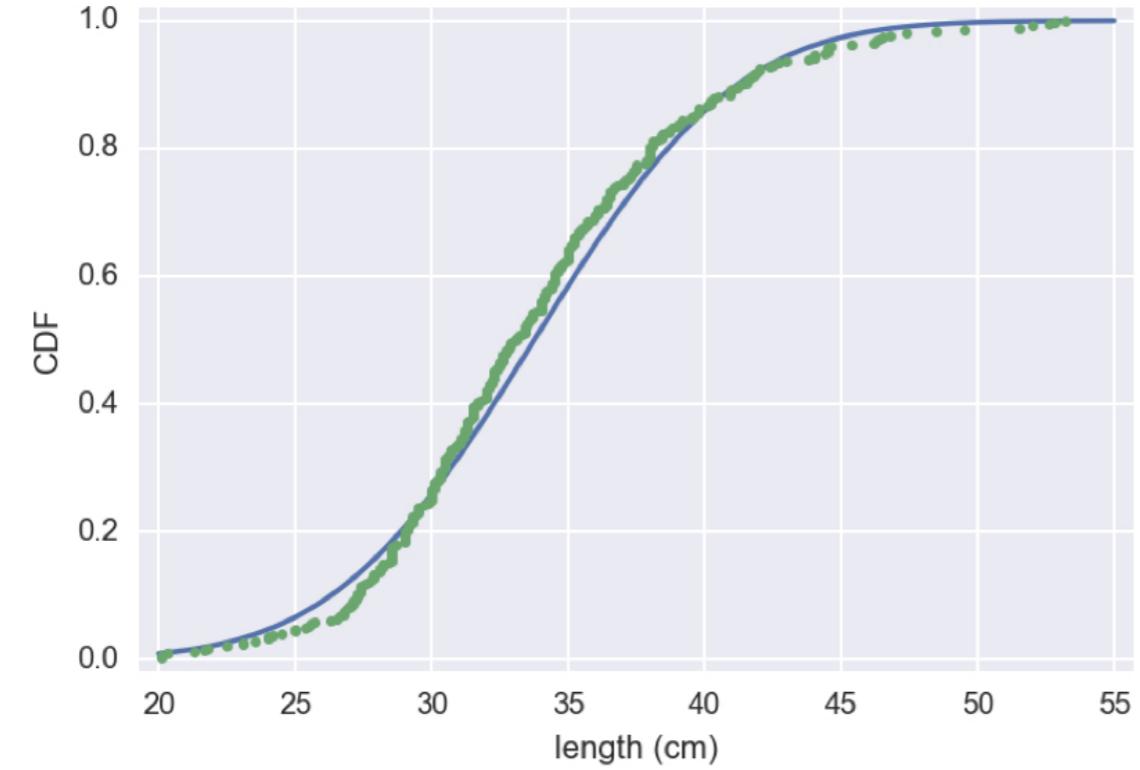
Image: Deutsche Bundesbank

# Length of MA large mouth bass



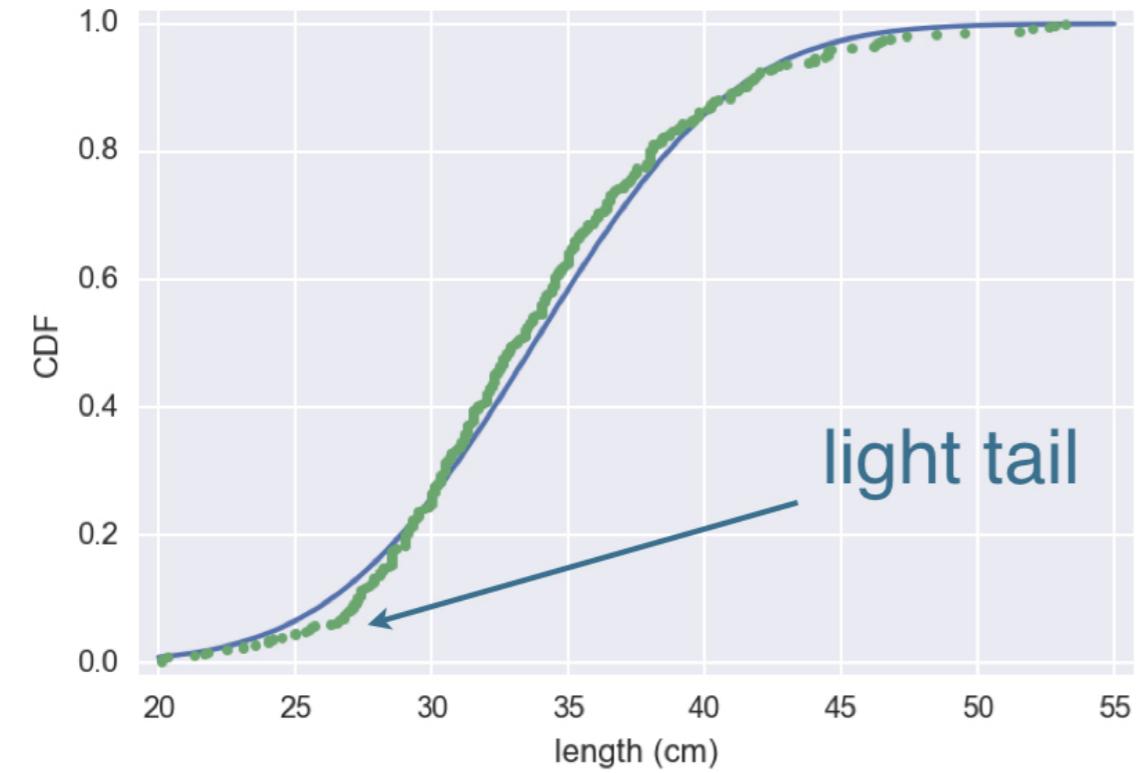
Source: Mass. Dept. of Environmental Protection

# Length of MA large mouth bass



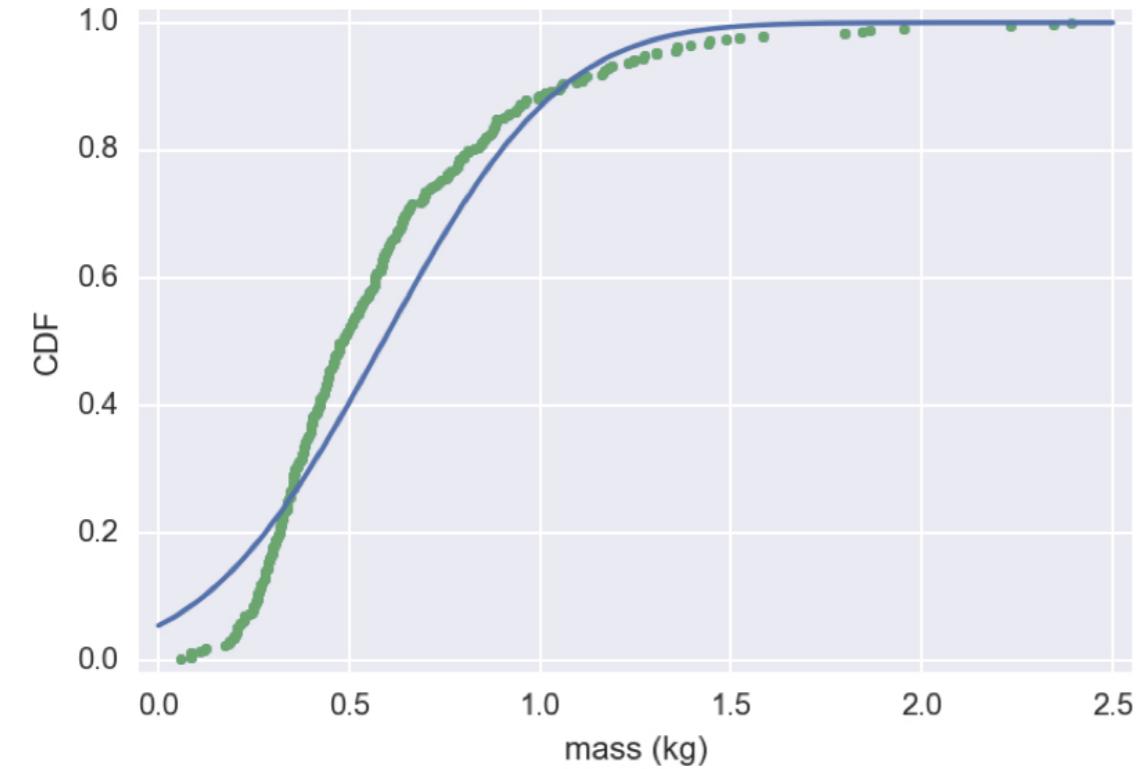
Source: Mass. Dept. of Environmental Protection

# Length of MA large mouth bass



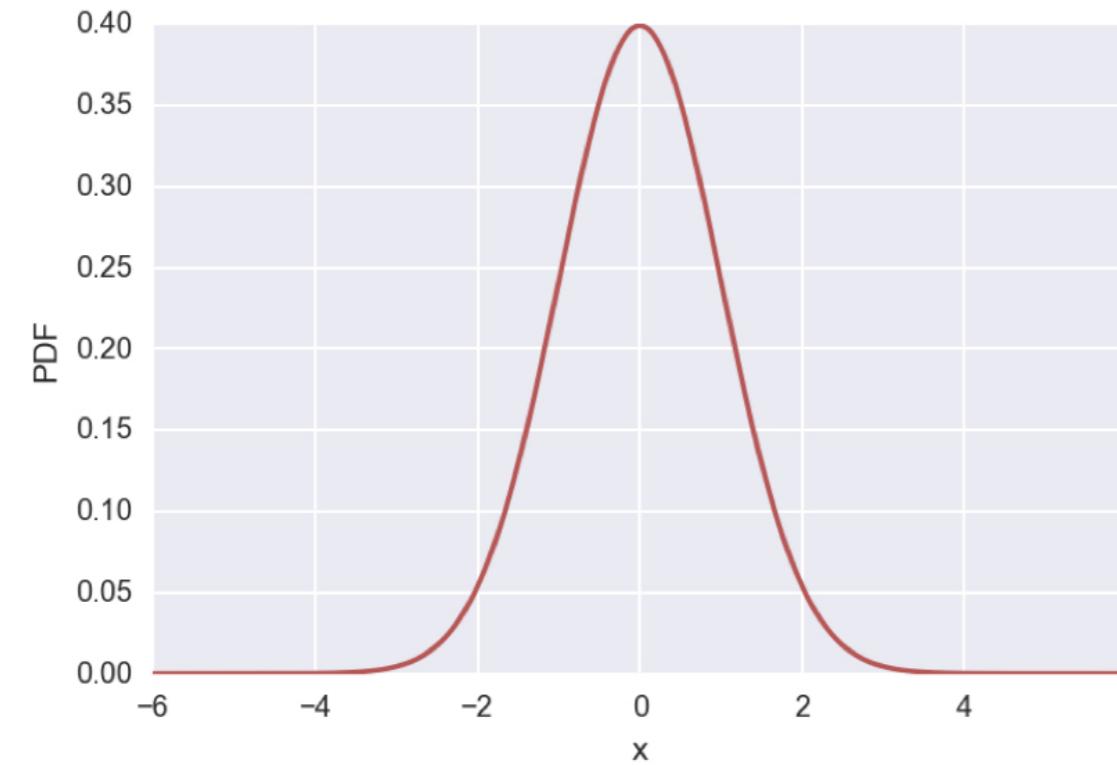
Source: Mass. Dept. of Environmental Protection

# Mass of MA large mouth bass

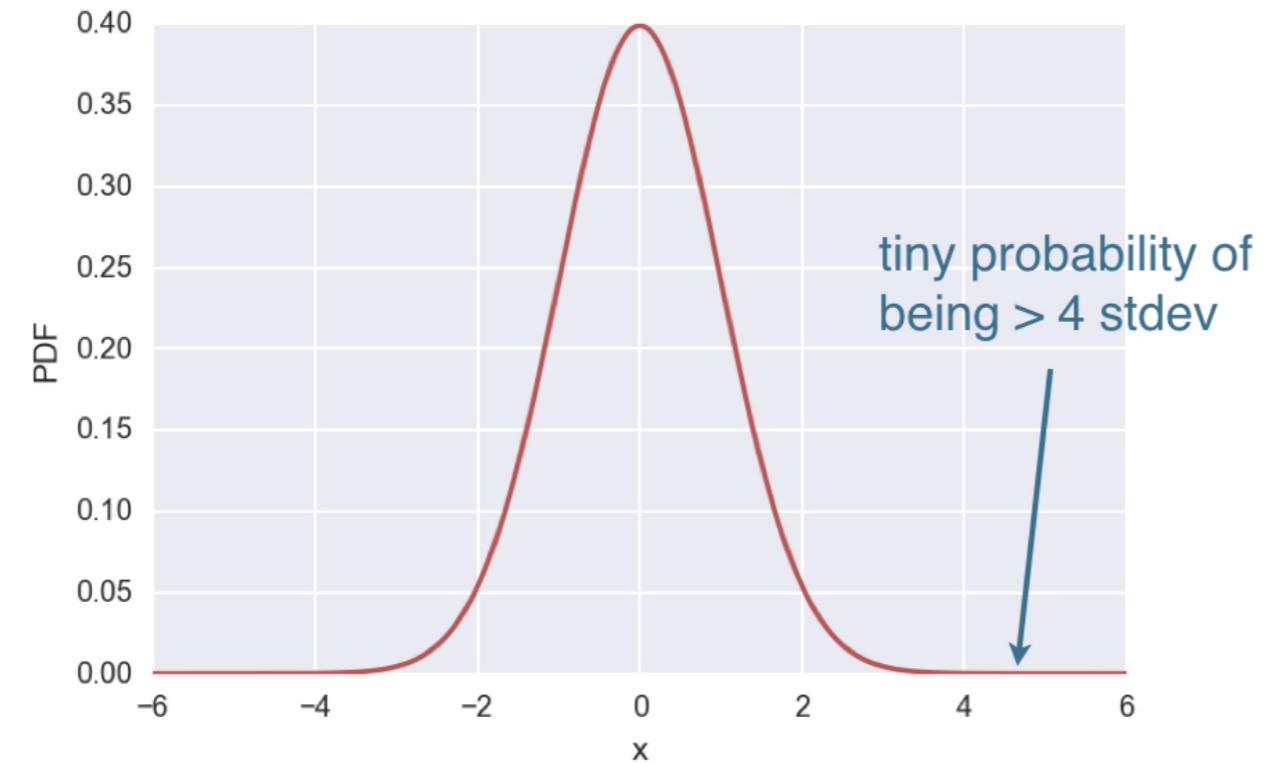


Source: Mass. Dept. of Environmental Protection

# Light tails of the Normal distribution



# Light tails of the Normal distribution



# **Let's practice!**

**STATISTICAL THINKING IN PYTHON (PART 1)**

# The Exponential distribution

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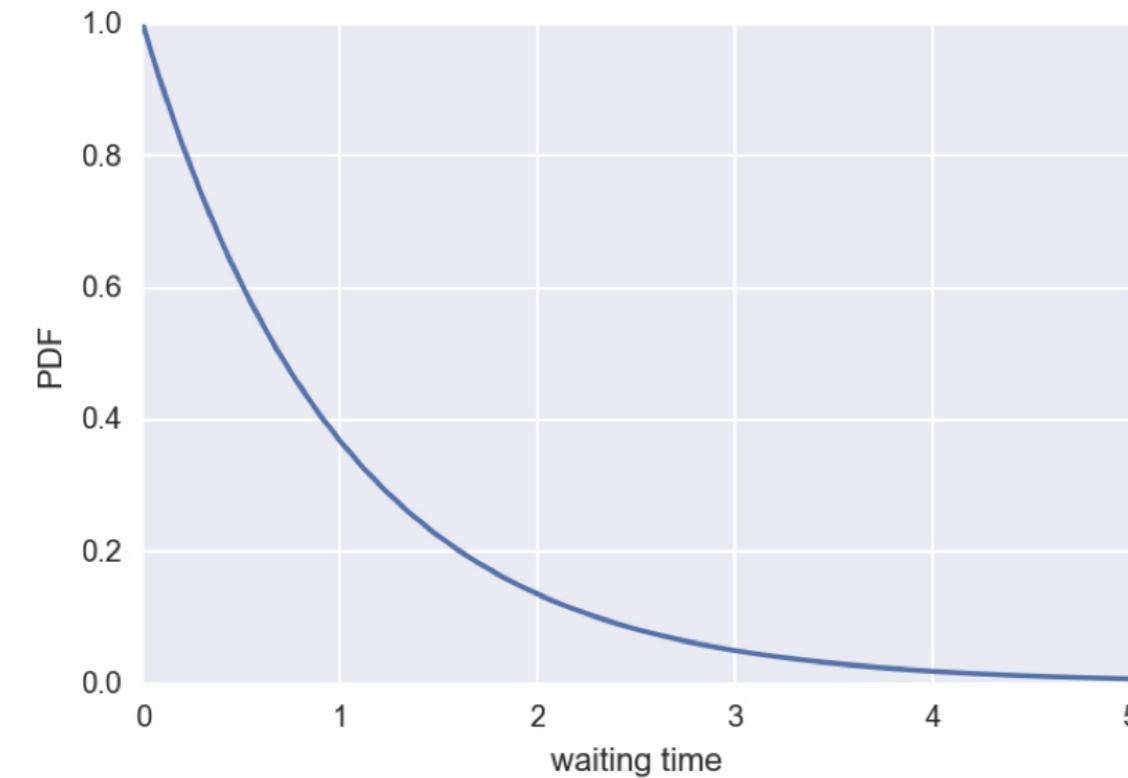
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# The Exponential distribution

- The waiting time between arrivals of a Poisson process is Exponentially distributed

# The Exponential PDF



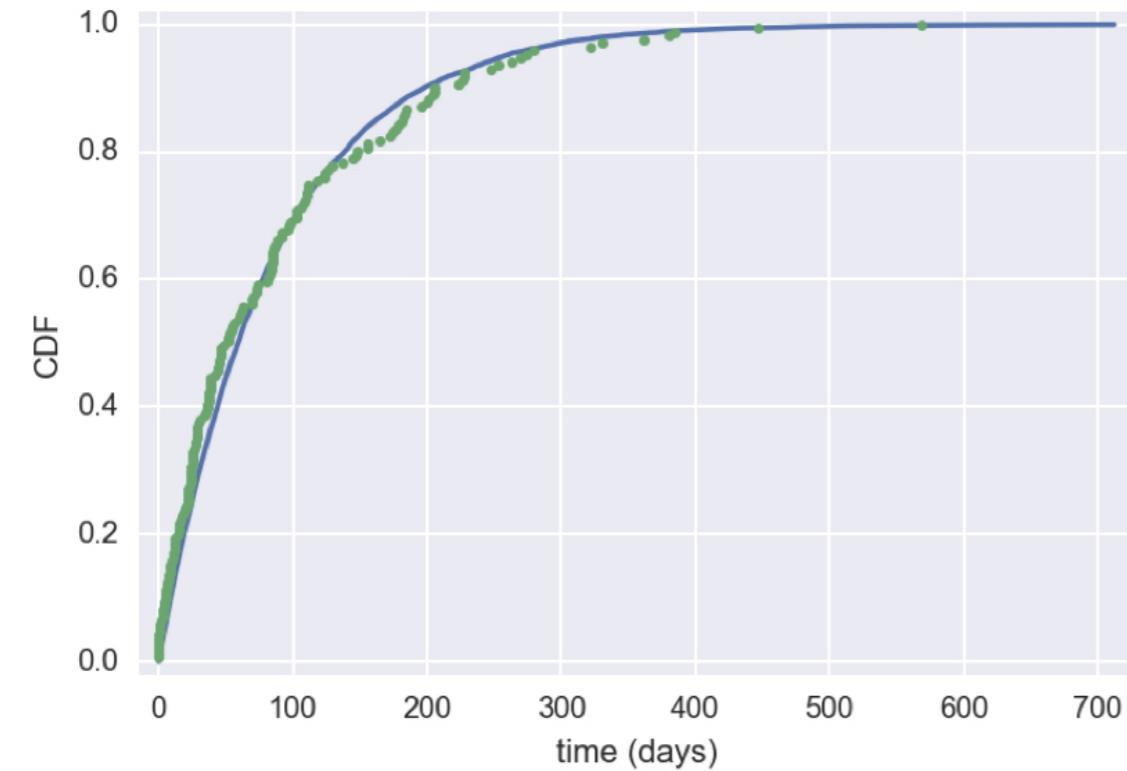
# Possible Poisson process

- Nuclear incidents:
  - Timing of one is independent of all others

# Exponential inter-incident times

```
mean = np.mean(inter_times)
samples = np.random.exponential(mean, size=1000)
x, y = ecdf(inter_times)
x_theor, y_theor = ecdf(samples)
_ = plt.plot(x_theor, y_theor)
_ = plt.plot(x, y, marker='.', linestyle='none')
_ = plt.xlabel('time (days)')
_ = plt.ylabel('CDF')
plt.show()
```

# Exponential inter-incident times



Data Source: Wheatley, Sovacool, Sornette, Nuclear Events Database

# **Let's practice!**

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# Final thoughts

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# You now can...

- Construct (beautiful) instructive plots
- Compute informative summary statistics
- Use hacker statistics
- Think probabilistically

# In the sequel, you will...

- Estimate parameter values
- Perform linear regressions
- Compute confidence intervals
- Perform hypothesis tests

# **Let's practice!**

**STATISTICAL THINKING IN PYTHON (PART 1)**