

Statistical Inference and Machine Learning in Earth Science **SIMLES**

Module 2 Statistical Inferences

Lecture 1 Maximum Likelihood Estimator

Probability Models: Moments

Probability models: allow us to model a random process

Random variables allow us to simulate the outcome of random processes.

Distributions: summarize the outcome of a random process.

- **pmfs, pdfs** describe the underlying *process*.
- **Histograms** summarize a finite sample.

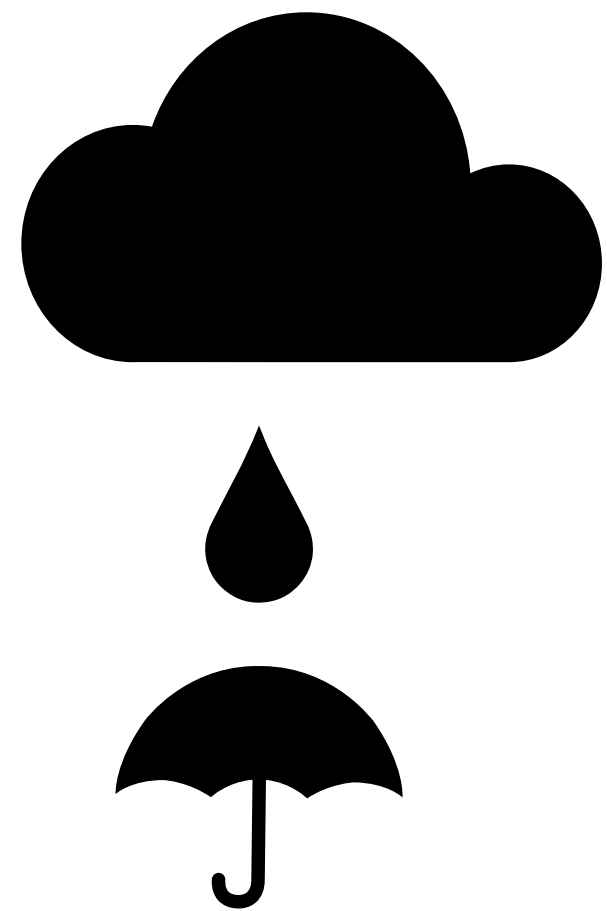
Moments: another useful summary of the outcome of a random process.

Probability Models: Random Variable

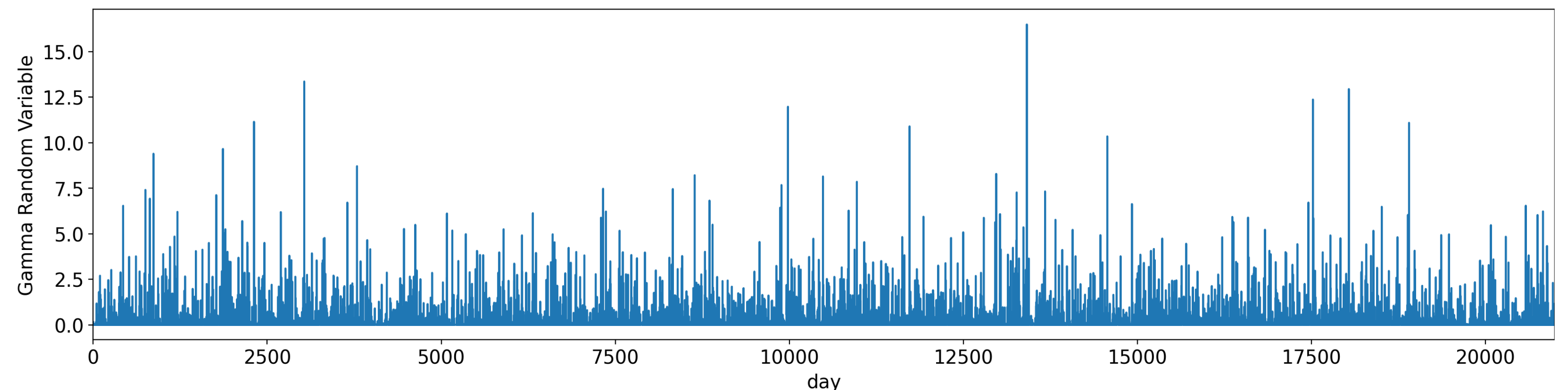
Probability models: allow us to model a random process

Random variables allow us to simulate the outcome of random processes.

- dice roll
- coin flip
- precipitation



$$X \sim \Gamma(\alpha, \beta)$$



Probability ↔ Statistical inference

Probability

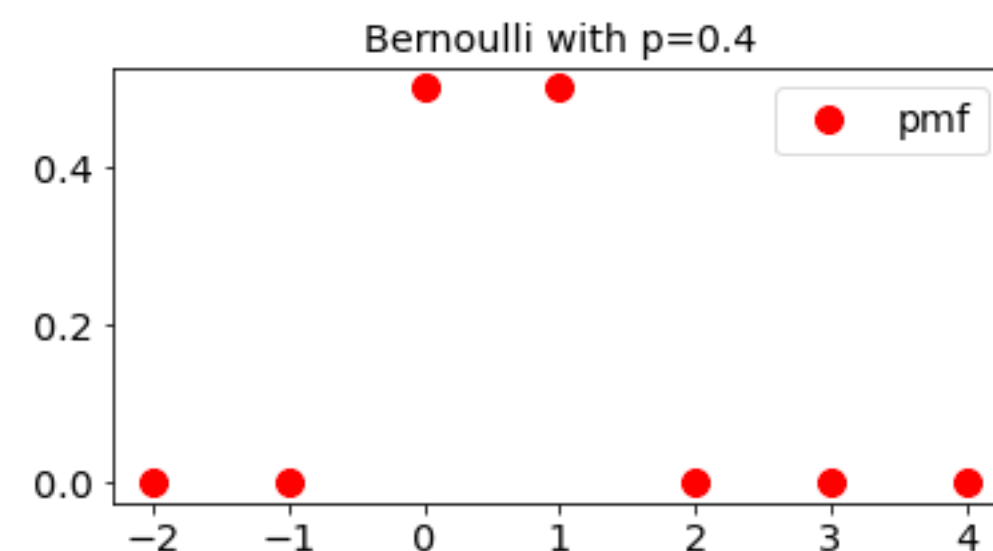
Process



Distribution $p(X = 1) = p$



Data



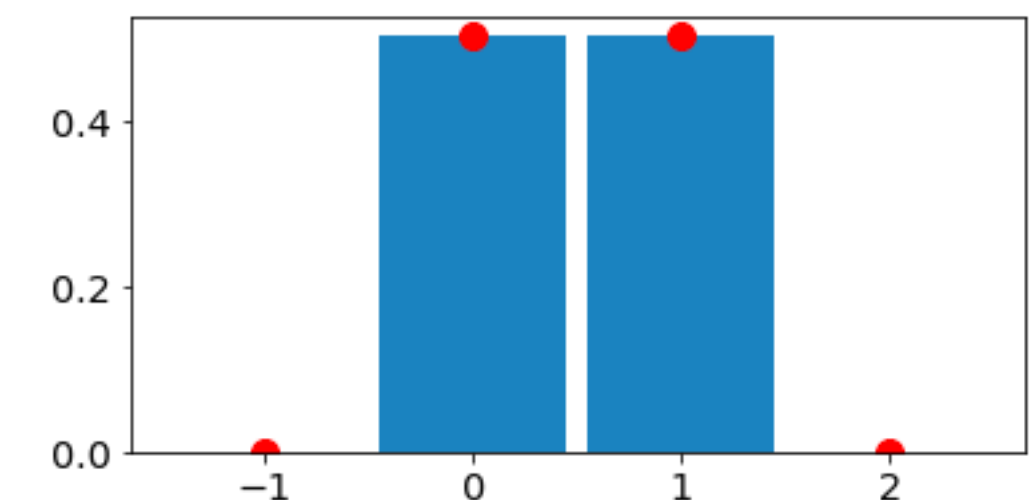
Statistics

Data

[1 1 1 1 0 1 1 1 1 0 1 0 1 1 1 1 1 1 1 1 1 0 1 1 1 1 1 0 1 1 1 1]



Distribution



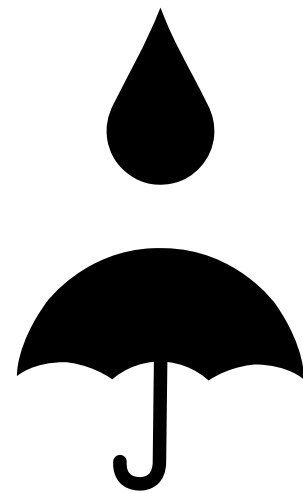
Process



Probability ↔ Statistical inference

Probability

Process

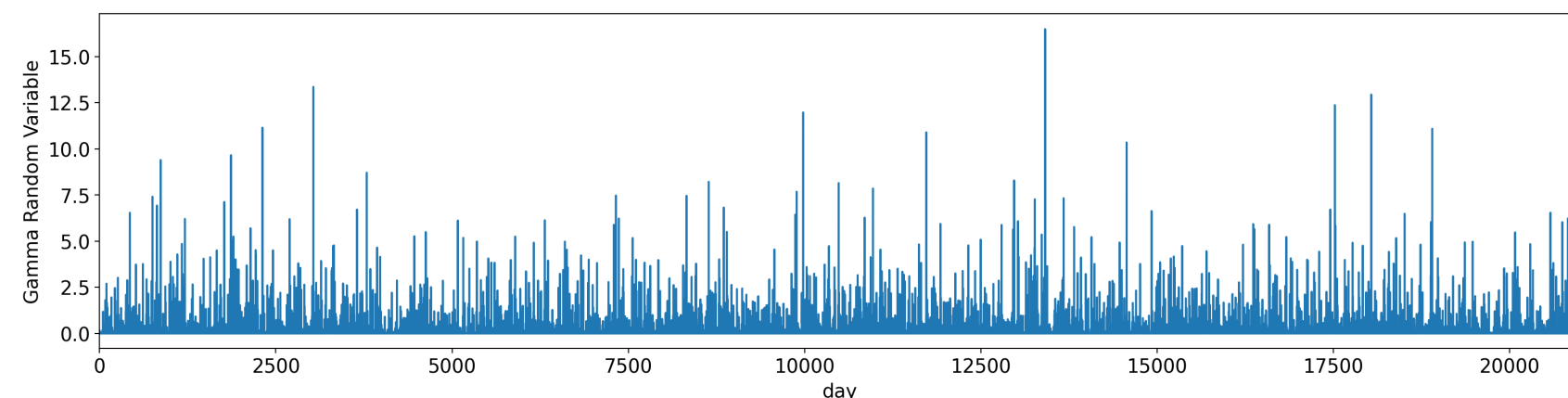


Distribution

$$X \sim \Gamma(\alpha, \beta)$$

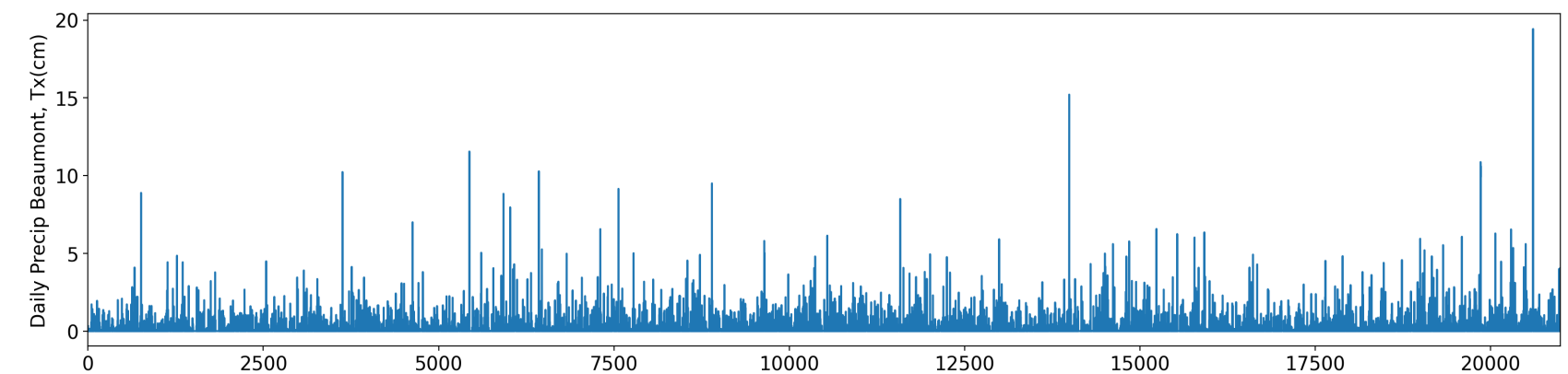


Data



Statistics

Data



Distribution

$$X \sim \Gamma(\alpha, \beta)$$



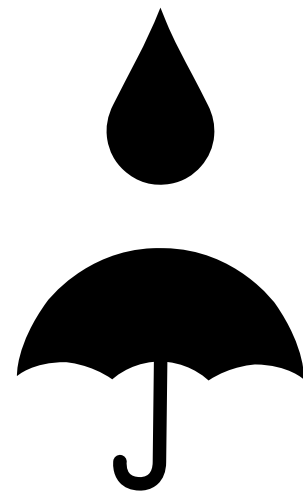
Process



Probability ↔ Statistical inference

Probability

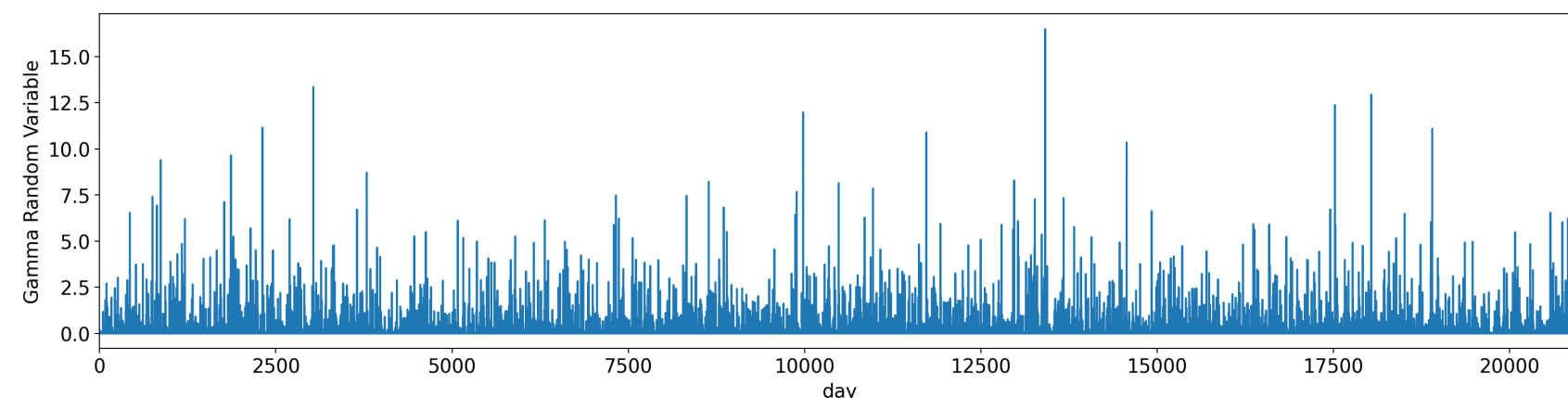
Process



Distribution

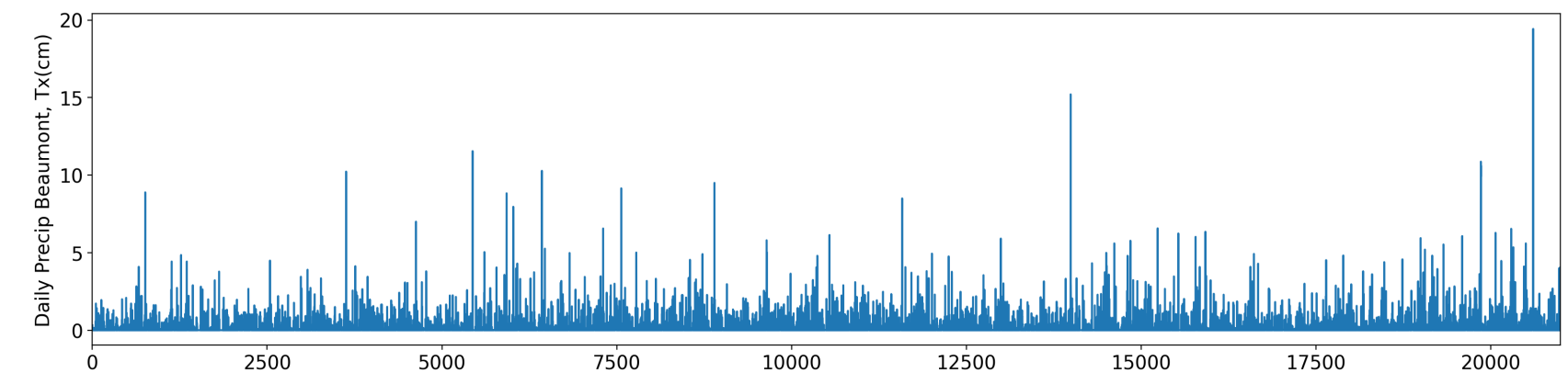
$$X \sim \Gamma(\alpha, \beta)$$

Data



Statistics

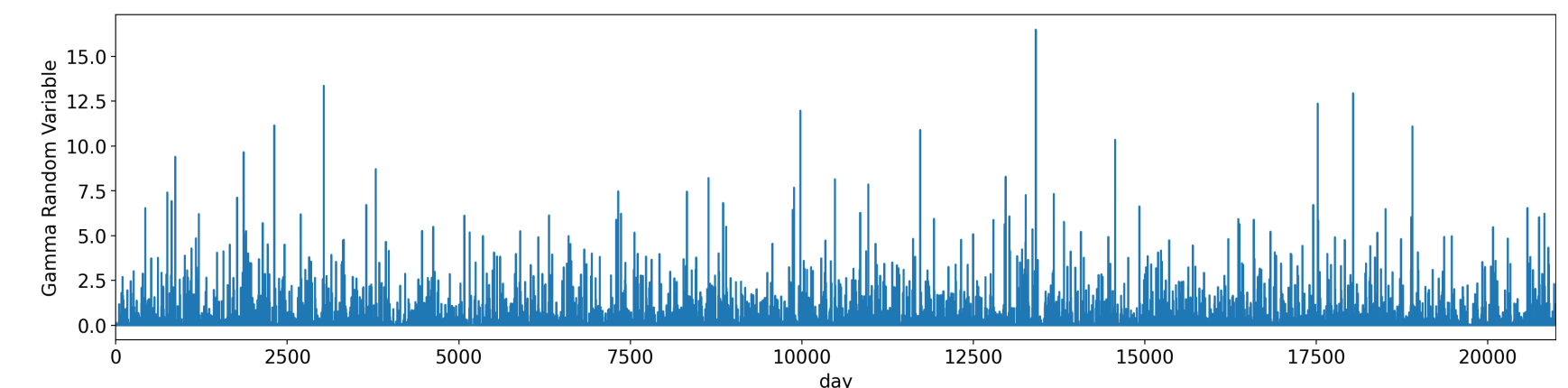
Data



Distribution

$$X \sim \Gamma(\alpha, \beta)$$

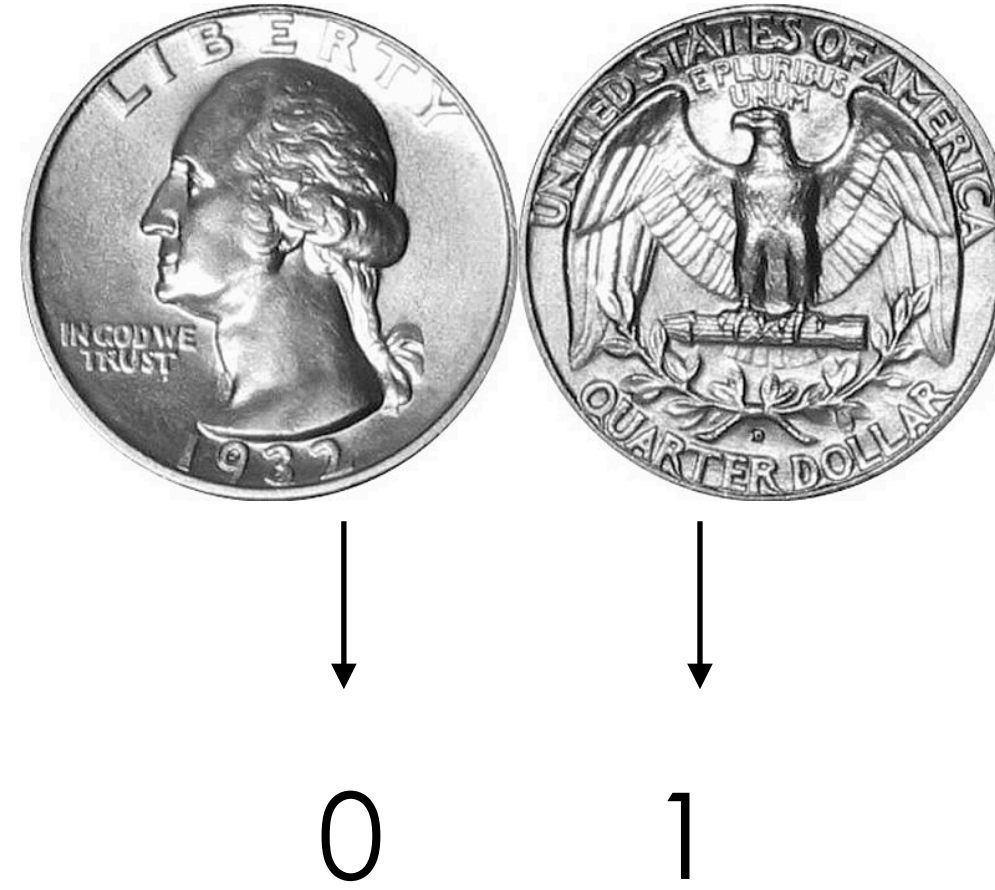
Simulation



Statistical Inference: Likelihood

Data

$$\mathbf{x} = [x_1, x_2, \dots, x_n]$$



$$x = [0, 1, 0, 1, 1, 1, 0, 0, 1, 1]$$

Statistical Inference: Likelihood

Data

$$\mathbf{x} = [x_1, x_2, \dots, x_n]$$



$$x = [0, 1, 0, 1, 1, 1, 0, 0, 1, 1]$$

Model Parameters

$$\theta = [\theta_1, \theta_2, \dots, \theta_n]$$

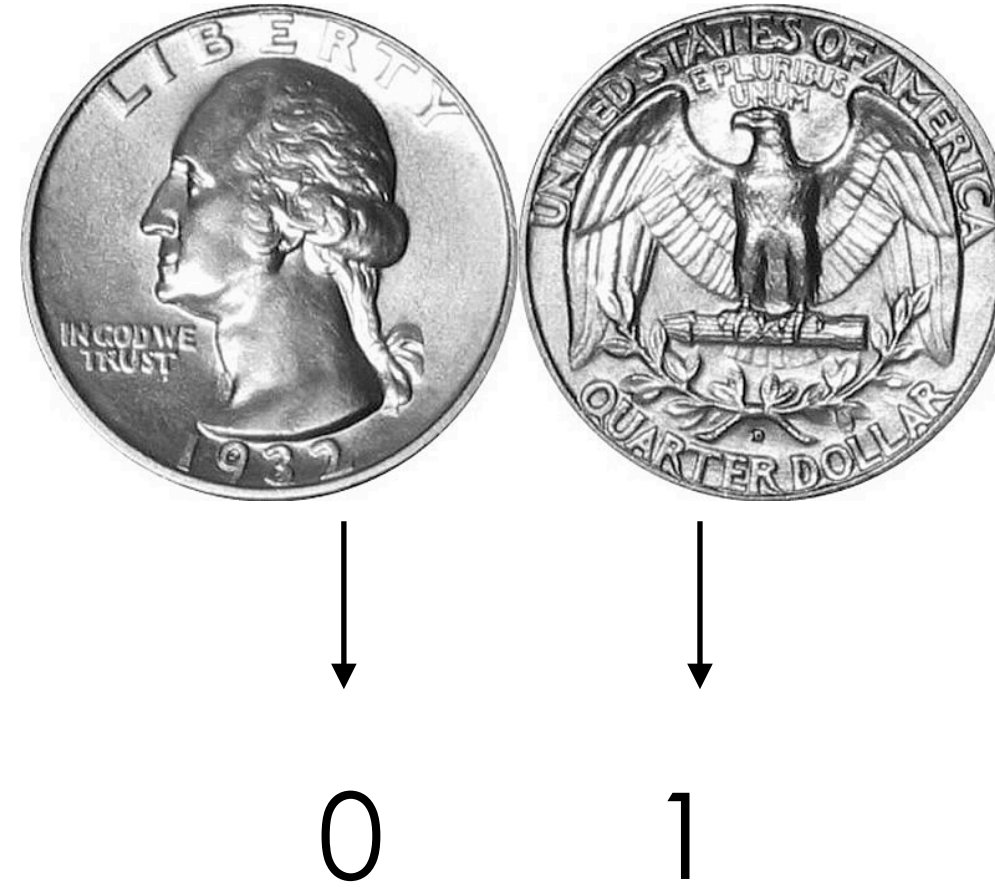
Model Parameters

$$X \sim \text{Bern}(p) \Rightarrow \theta = [p]$$

Statistical Inference: Likelihood

Data

$$\mathbf{x} = [x_1, x_2, \dots, x_n]$$



$$x = [0, 1, 0, 1, 1, 1, 0, 0, 1, 1]$$

Model Parameters

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Model Parameters

$$X \sim \text{Bern}(p) \Rightarrow \theta = [p]$$

Likelihood

$$\mathcal{L} = p(x | \theta)$$

What is the probability that
a coin with parameter θ
generated this data?

Gaussian

Data

$$\mathbf{X} = [x_1, x_2, \dots, x_n]$$

```
[5.73541792 4.59128669 5.67946042 5.66998442 6.2729418 7.11995521
5.59525591 6.13115623 1.95810532 6.40934121 4.98045955 6.24052625
3.23506281 7.29882568 6.69172193 5.13690828 5.23322762 3.53067353
5.1935833 5.92951912]
```

Model Parameters

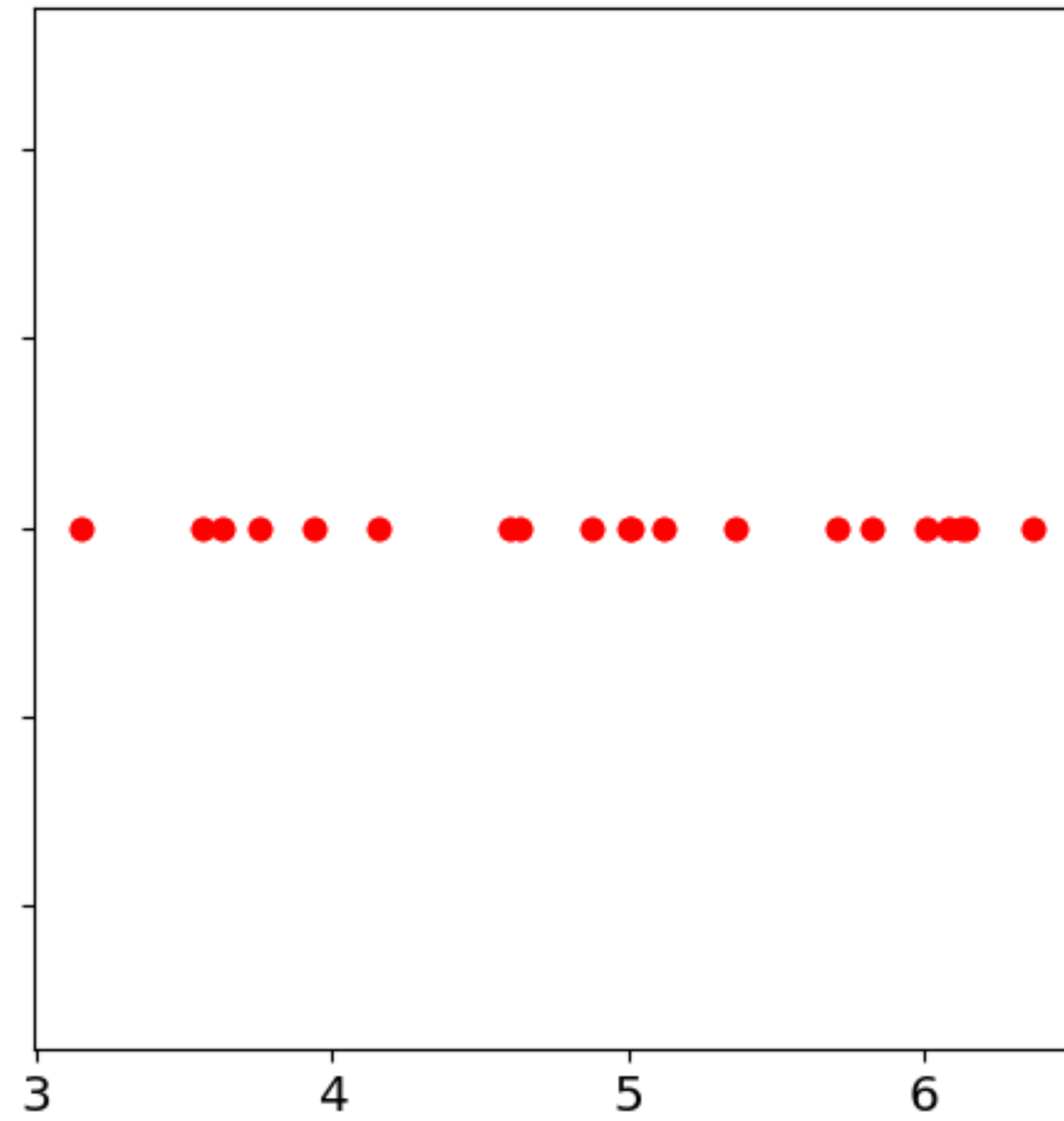
$$\theta = [\theta_1, \theta_2, \dots, \theta_n]$$

$$\begin{array}{ccc} \text{Model} & & \text{Parameters} \\ X \sim \mathcal{N}(\mu, \sigma) & \Rightarrow & \theta = [\mu, \sigma] \end{array}$$

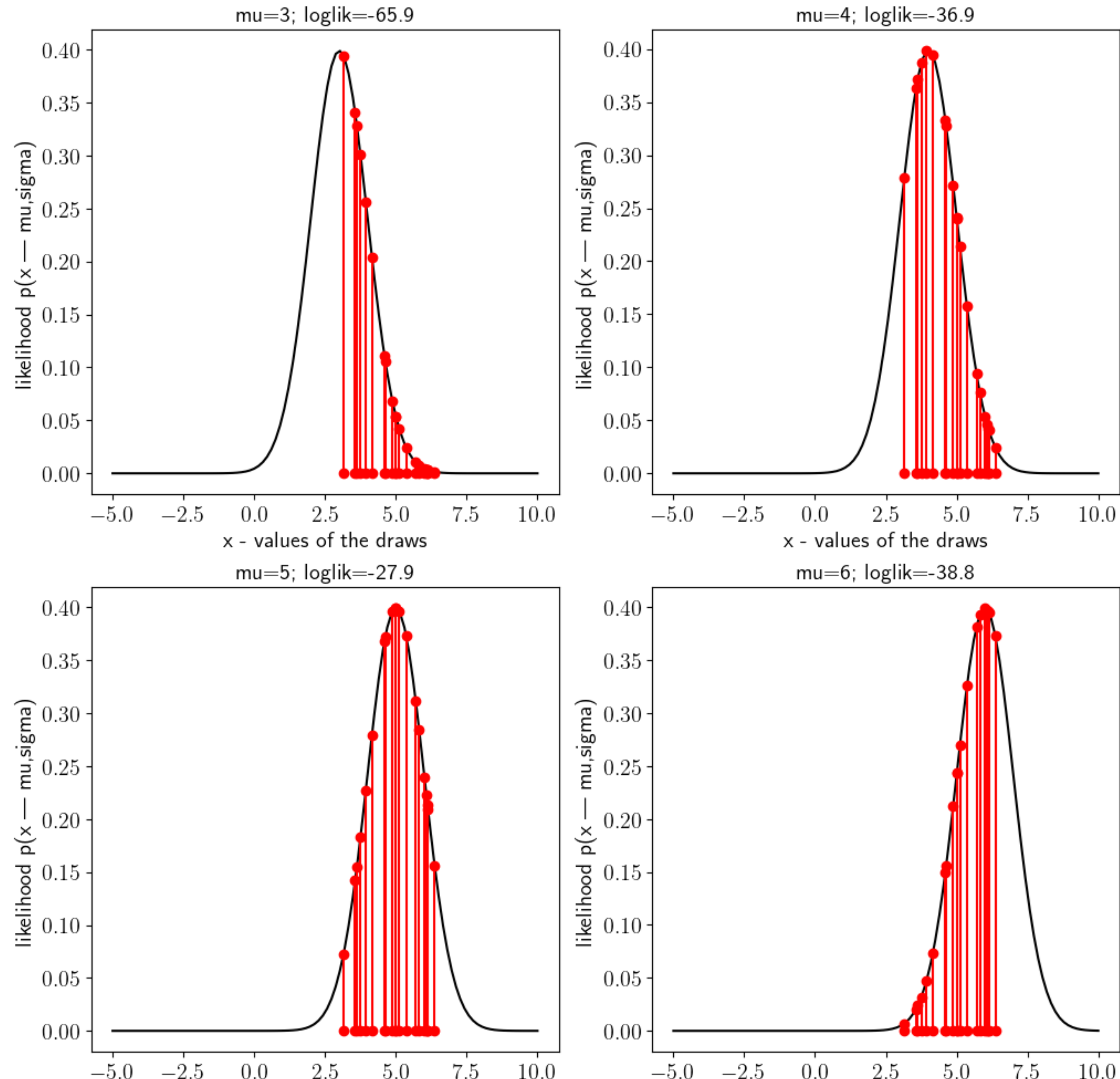
Likelihood

$$\mathcal{L} = p(x | \theta)$$

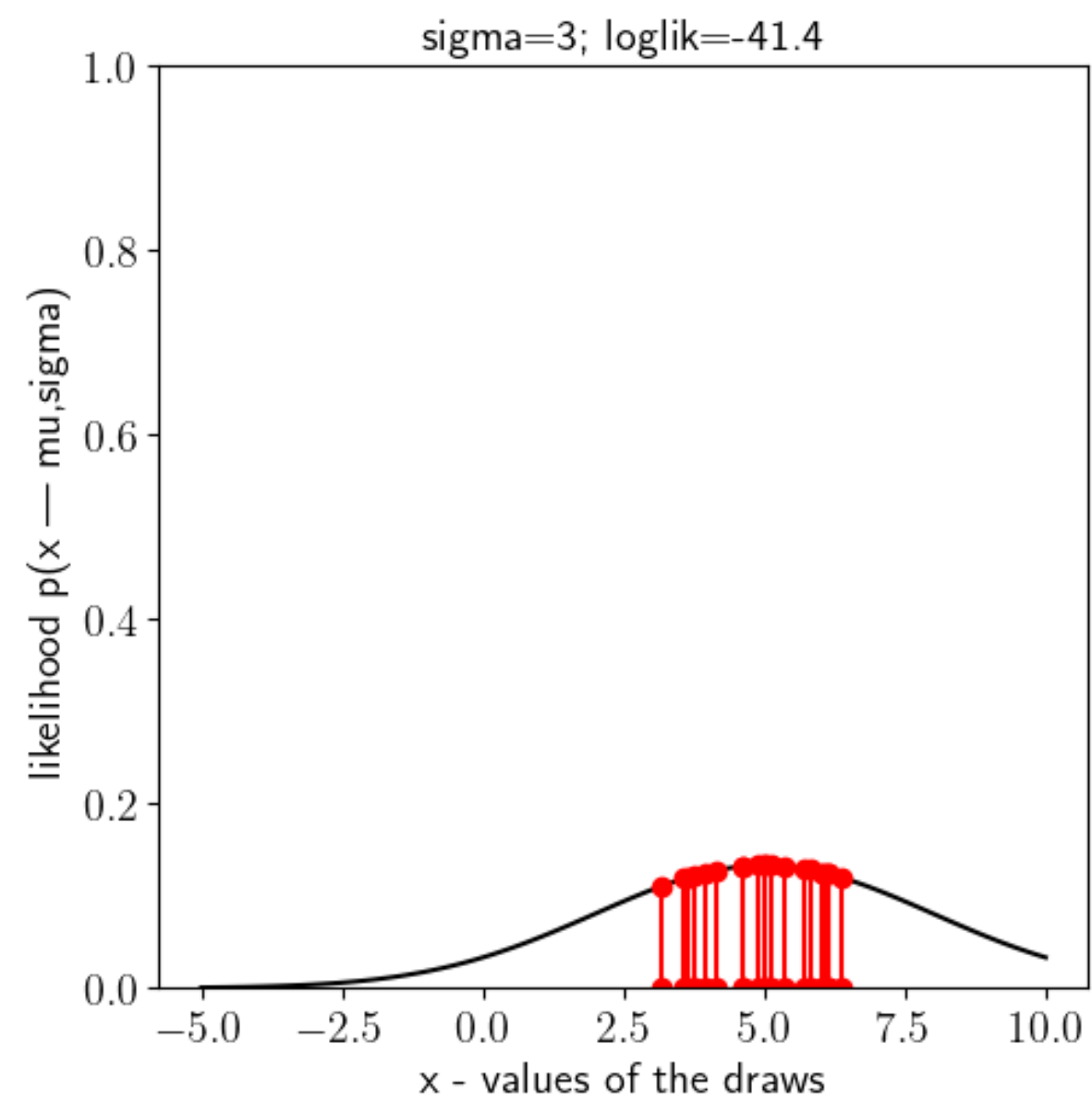
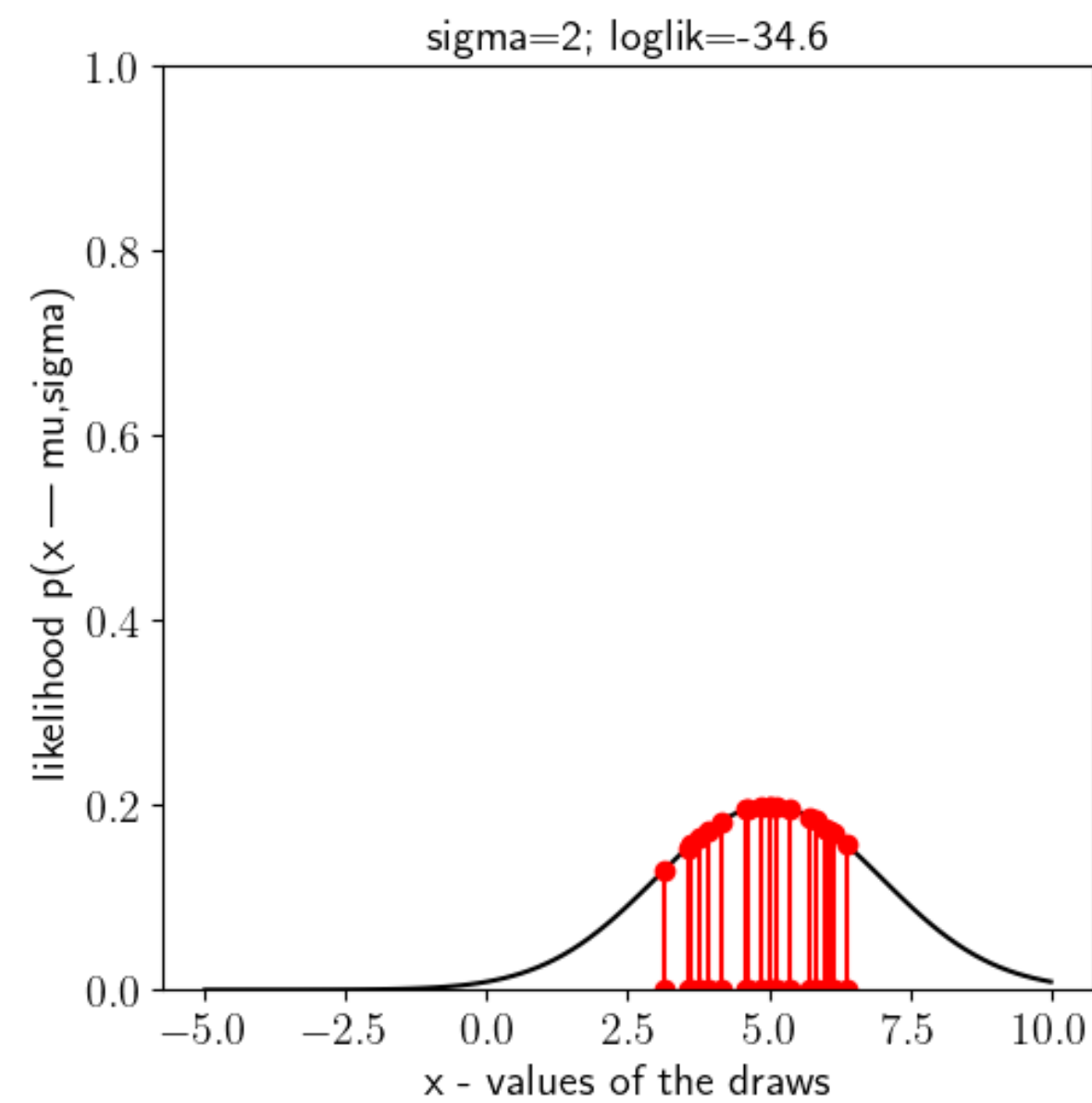
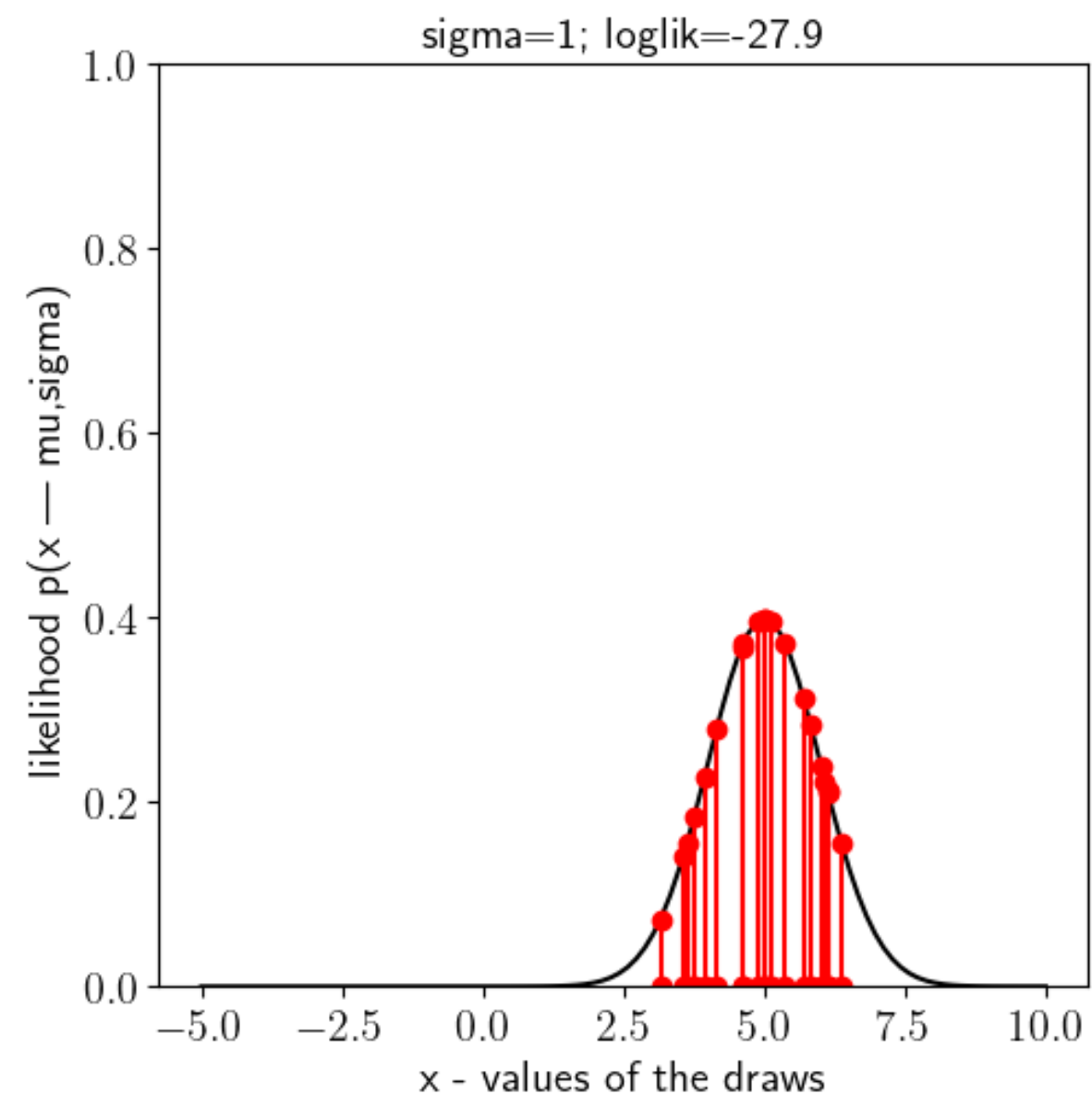
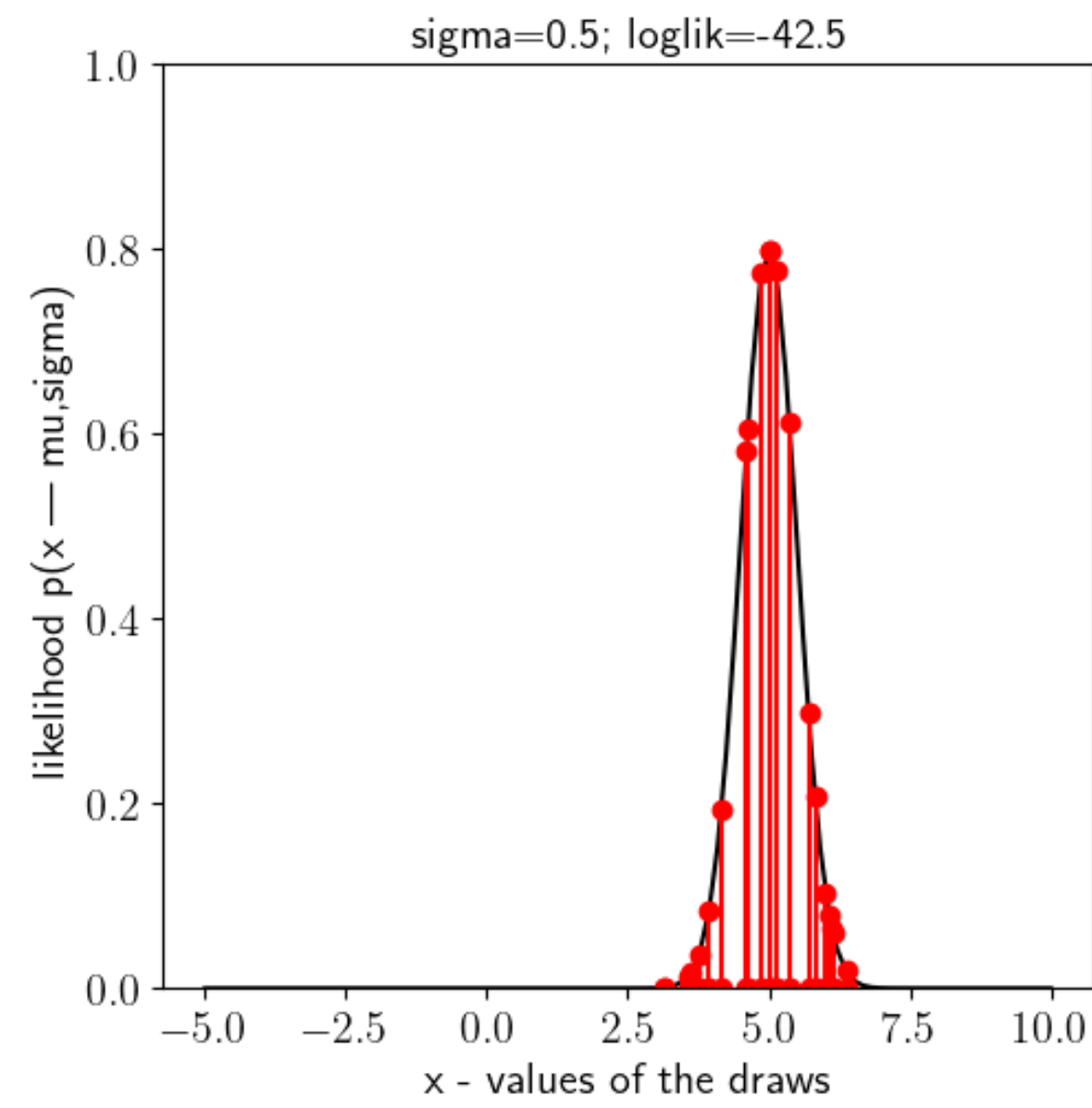
Gaussian



Gaussian



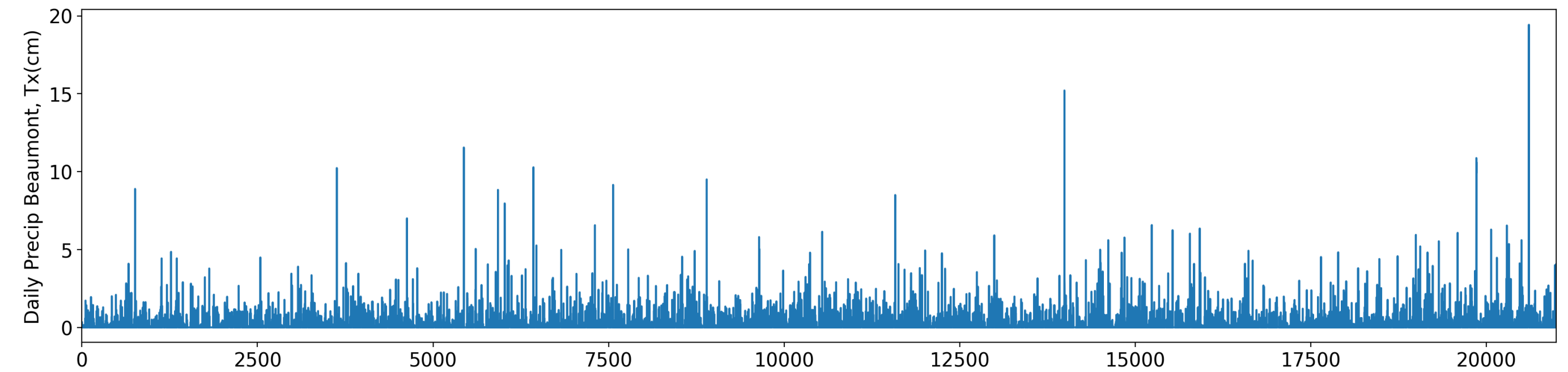
Gaussian



Statistical Inference: Likelihood

Data

$$\mathbf{X} = [x_1, x_2, \dots, x_n]$$

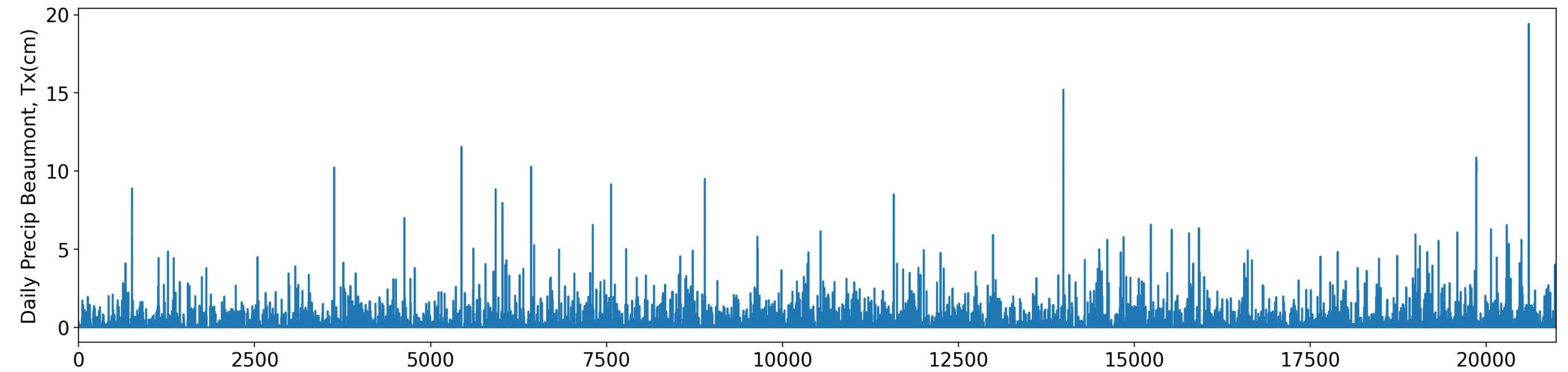


$$\mathbf{X} = [P(t_1), P(t_2), \dots, P(t_n)]$$

Statistical Inference: Likelihood

Data

$$\mathbf{X} = [x_1, x_2, \dots, x_n]$$



$$\mathbf{x} = [P(t_1), P(t_2), \dots, P(t_n)]$$

Model Parameters

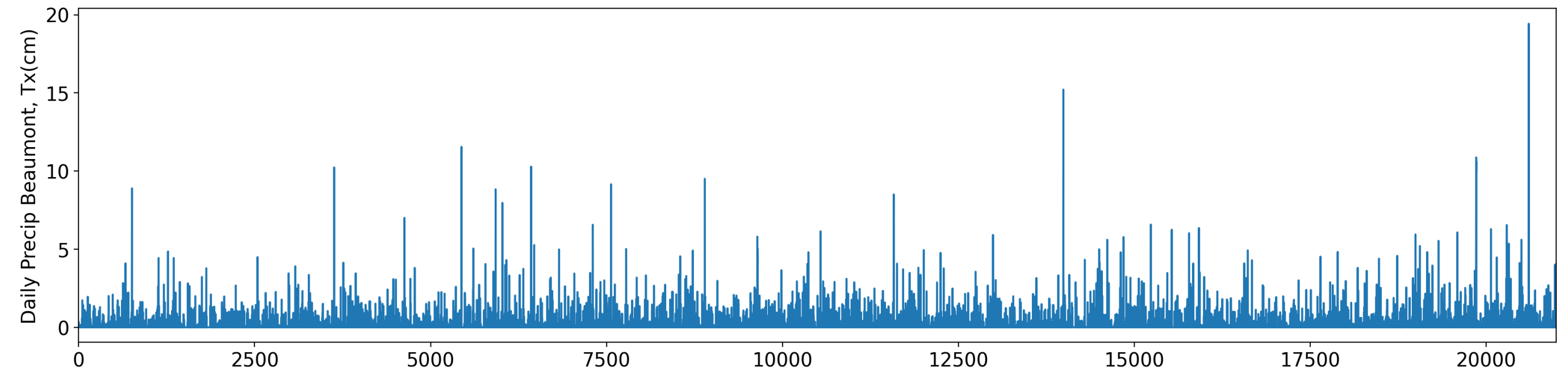
$$\theta = [\theta_1, \theta_2, \dots, \theta_n]$$

Model		Parameters
$X \sim \Gamma(\alpha, \beta)$	\Rightarrow	$\theta = [\alpha, \beta]$

Statistical Inference: Likelihood

Data

$$\mathbf{X} = [x_1, x_2, \dots, x_n]$$



$$\mathbf{x} = [P(t_1), P(t_2), \dots, P(t_n)]$$

Model Parameters

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Model		Parameters
$X \sim \Gamma(\alpha, \beta)$	\Rightarrow	$\theta = [\alpha, \beta]$

Likelihood

$$\mathcal{L} = p(x | \theta)$$

???