

Classical statistical inference

Uncertainties on arbitrary RV

Part 4

Associated notebook:

[04-Basic_statistical_inference_frequentists_2/Frequentist_Monte_Carlo.ipynb](#)

Uncertainty calculation

How to calculate **stde** (or simply **std**) on a RV if it is a function of one or multiple RV ?

Case 1: $y = \phi(x)$ $p(x)$ known

Case 2: $z = \phi(x, y)$ $p(x)$ and $p(y)$ known

Uncertainty calculation

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Example: $y = -2.5 * \log(x)$

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Uncertainty calculation

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$$x = \phi^{-1}(y) = 10^{-0.4*y}$$

$$p(x) \sim U(0, 1) \begin{cases} 1 & \text{if } x \in [0, 1] \\ 0 & \text{if } x \notin [0, 1] \end{cases}$$

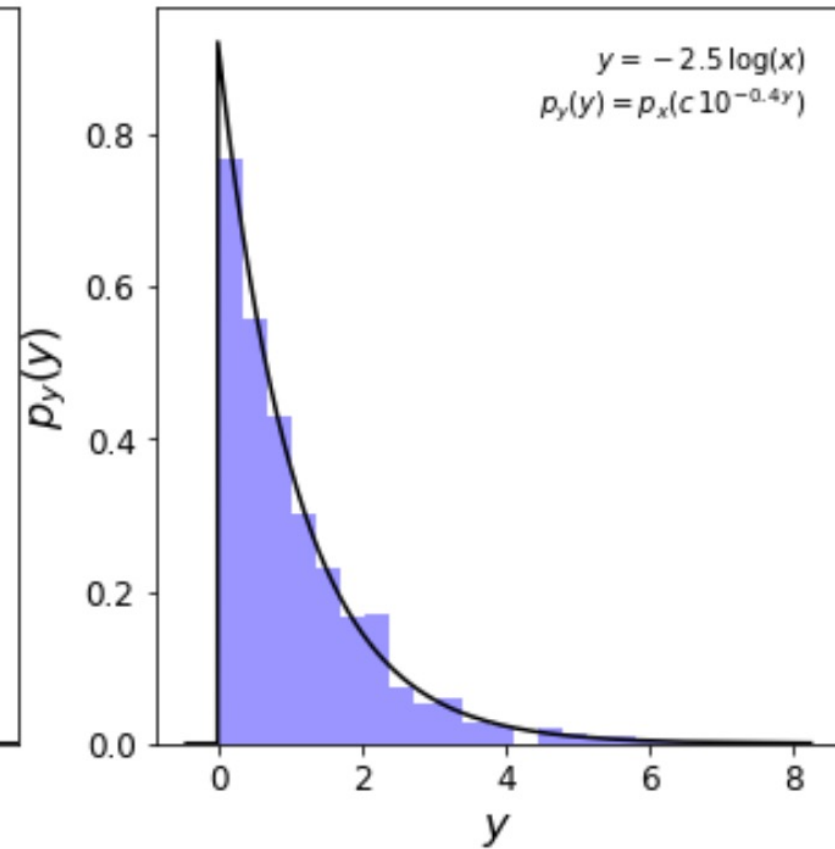
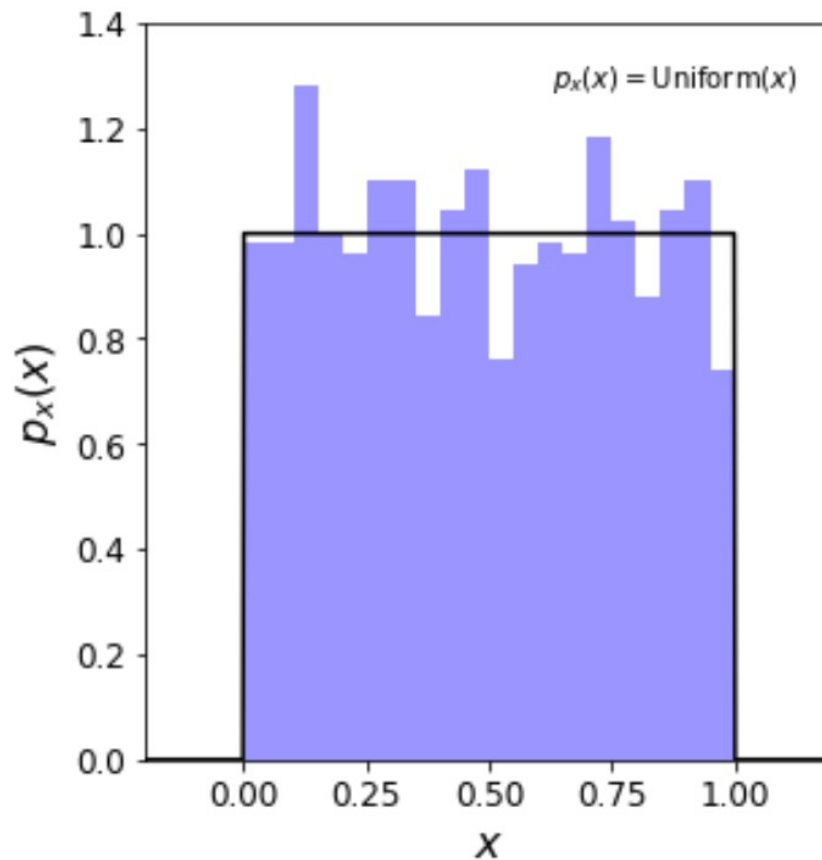
$$p(y) = 1 \times 0.4 \ln(10) 10^{-0.4 y}$$

$$0 < y < \infty$$

Uncertainty calculation

Go to: Sect. II.4.1 of the notebook

Example: $y = -2.5 * \log(x)$



Uncertainty calculation

Case 2: $z = \phi(x, y)$ $p(x)$ and $p(y)$ known

Error propagation formula

$$\sigma_z^2 = \left(\frac{\partial \phi}{\partial x} \right)_{\bar{x}}^2 \sigma_x^2 + \left(\frac{\partial \phi}{\partial y} \right)_{\bar{y}}^2 \sigma_y^2 + 2 \left(\frac{\partial \phi}{\partial x} \right)_{\bar{x}} \left(\frac{\partial \phi}{\partial y} \right)_{\bar{y}} \sigma_{xy}$$

Results from Taylor expanding z around \bar{x} and \bar{y} \Rightarrow neglects some high order terms

Go to: Sect. II.4.2 of the notebook for the demo

Uncertainty calculation

Monte-Carlo

$c = \Phi(a, b): \sigma_c ?$

The expectation (“average”) of a function $f(x)$ of a RV x can be approximated by drawing a virtually infinite sample from x

$$E(f(x)) = \int_{-\infty}^{+\infty} f(x)h(x)dx \rightarrow \frac{1}{N} \sum_i^N f(x_i),$$

You can specialise $f()$ to the calculation of the mean, or the variance of a RV.

1. Draw random samples from $h(a; \mu_a, \sigma_a), h(b, ; \mu_b, \sigma_b)$
2. Construct a random (*monte-carlo*) sample of $c_i = \Phi(a_i, b_i)$
3. Derive σ_c from your Monte-Carlo sample

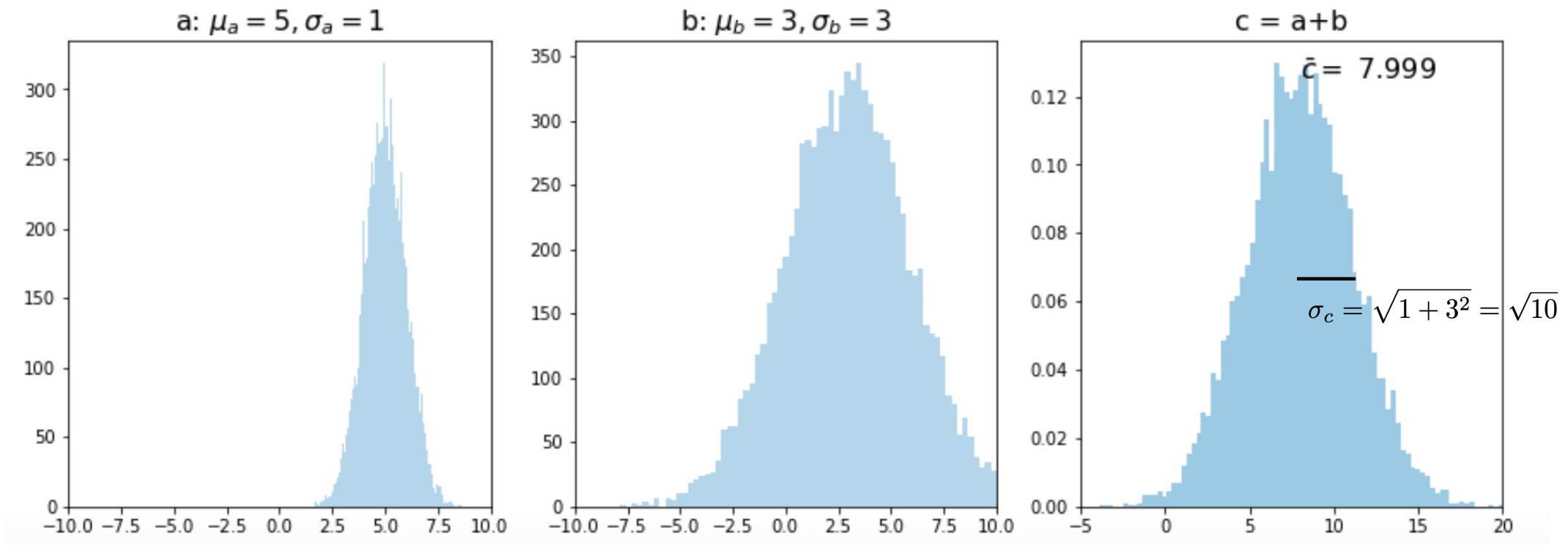
Uncertainty calculation

Monte-Carlo

$$c = a + b$$

Go to: Sect. II.4.3 of the notebook

1. Draw random samples from $h(a; \mu_a, \sigma_a)$, $h(b; \mu_b, \sigma_b)$
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Classical statistical inference

Bootstrap and Jackknife

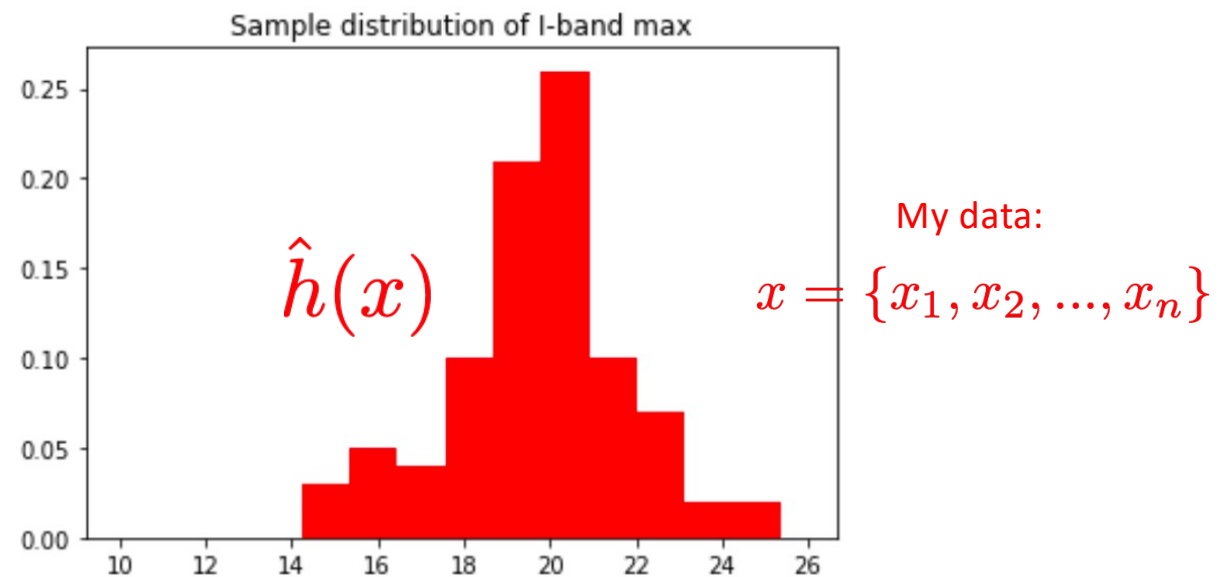
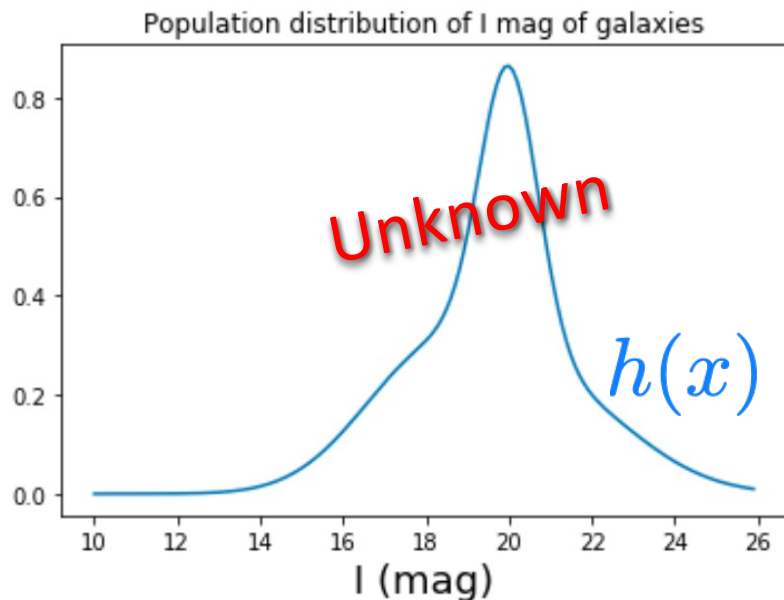
Part 5

Associated notebook:

[04-Basic statistical inference frequentists 2/
Frequentist inference Bootstrap.ipynb](#)

Bootstrap

Go to: Sect. II.5 of the notebook



Bootstrap \equiv Draw samples from the sample PDF $\hat{h}(x)$, allowing from replacements

$$B = \{x_1^*, x_2^*, \dots, x_n^*\}$$

· w. x^* from $\{x_1, x_2, \dots, x_n\}$

e.g. $B1 = \{x_1, x_1, x_7, \dots, x_{28}\}$

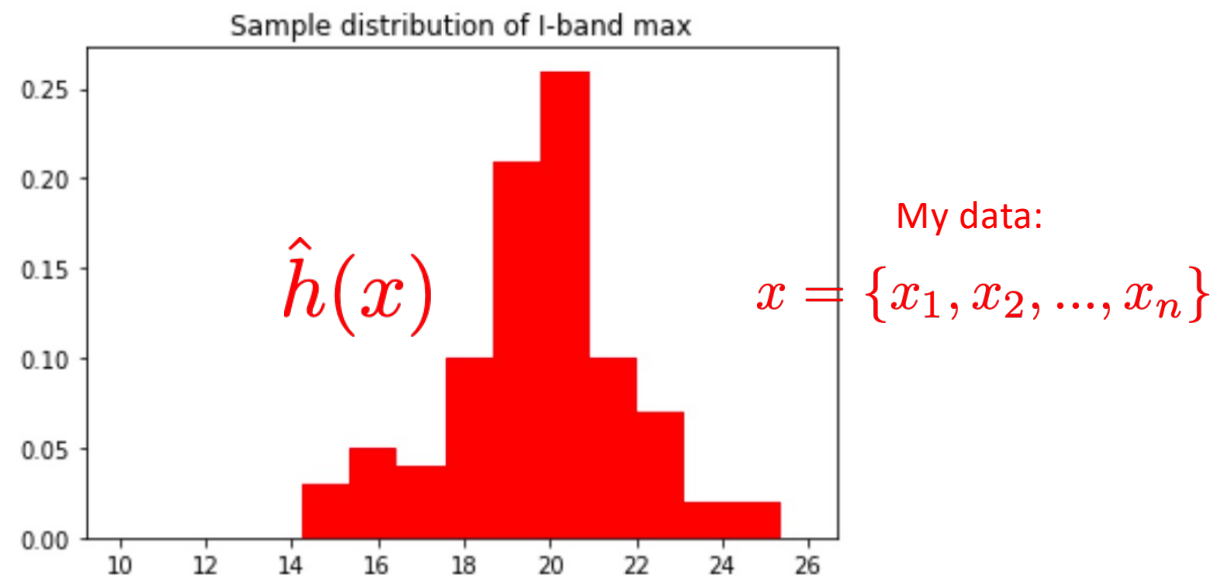
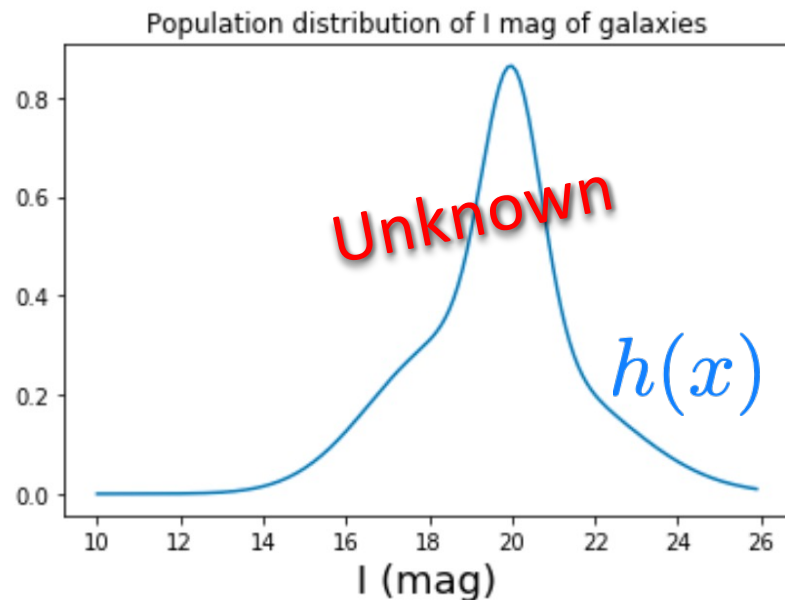
$B2 = \{x_2, x_{36}, x_9, \dots, x_8\}$

...

$Bk = \{x_{16}, x_{12}, x_3, \dots, x_{10}\}$

Bootstrap Confidence Interval

Go to: Sect. II.5.1 of the notebook



Calculate the estimate of your statistics q from all the bootstrapped samples + its associated $stde^B(q)$

Normal CI:

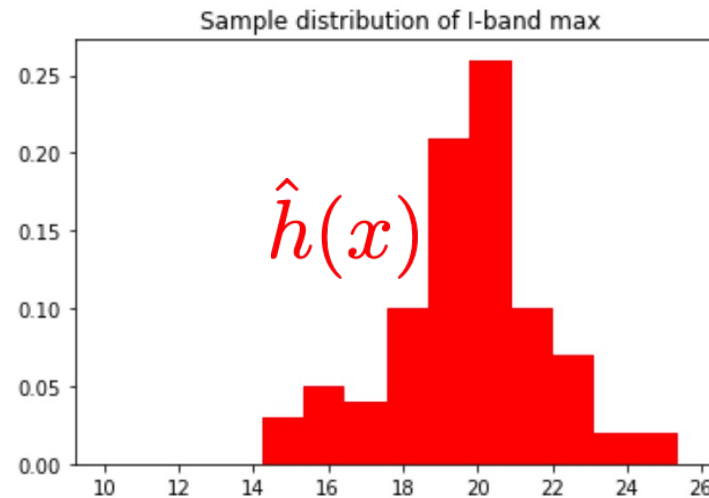
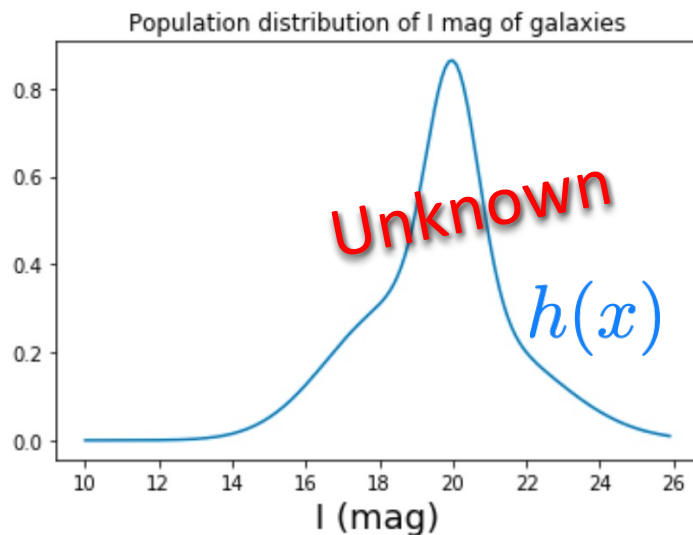
$$[\hat{q} - z_{\alpha/2} stde^B(q), \hat{q} + z_{\alpha/2} stde^B(q)]$$

Percentile CI:

$$[q_{\alpha/2}^*, q_{1-\alpha/2}^*]$$

Jackknife

Go to: Sect. II.5.2 of the notebook



My data:

$$x = \{x_1, x_2, \dots, x_n\}$$

$$J1 = \{x_1, \dots, x_{n-2}, x_{n-1}\}$$

$$J2 = \{x_1, \dots, x_{n-2}, x_n\}$$

...

$$Jn = \{x_2, \dots, x_{n-1}, x_n\}$$

With n-1 point per sample

Jackknife \equiv Remove 1 data point from your sample, n times

WARNING: statistics q_n from Jackknife is biased

$$q^J = n q_n - (n - 1) \bar{q}_n \quad \bar{q}_n = n^{-1} \sum_{i=1}^n q_i^*$$

$$\sigma_q^2 = \frac{n-1}{n} \sum_{i=1}^n (q_i^* - \bar{q}_n)^2$$