#### Computational Bayesian statistics

# Computational solutions

Bayes Theorem ⇒ posterior distribution

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How can one estimate the *posteriori* distribution ?

- ⇒ sample according to this posterior distribution
  - direct sampling
  - Markov chain Monte Carlo (MCMC)

# Monte Carlo method

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- ⇒ use random numbers to compute quantities whose analytical computation is hard (or impossible)
  - Law of Large Numbers (LLN)
  - so-called "Monte Carlo sample"
- compute various functions from that sample distribution

**Example :** One wants to compute  $\mathbb{E}[f(X)] = \int f(x)p_X(x)dx$ 

If 
$$x_i \stackrel{iid}{\sim} p_X$$
,  $\mathbb{E}[f(X)] = \frac{1}{N} \sum_{i=1}^{N} f(x_i)$  (LLN)

 $\Rightarrow$  if one knows how to sample from  $p_X$ , one can then estimate  $\mathbb{E}[f(X)]$ 

. . .

Computational Bayesian statistics

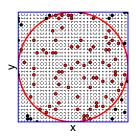
## Monte Carlo method: illustration

### $\pi$ estimation:

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### $\pi$ estimation:





A casino roulette (in Monte Carlo ?)

A 36×36 grid

- 1 The probability of being inside the disk while in the square:  $p_C = \frac{\pi R^2}{(2R)^2} = \frac{\pi}{4}$
- 2 n points  $\{(x_{11}, x_{21}), \dots, (x_{1n}, x_{2n})\} = \{P_1, \dots, P_n\}$  on the  $36 \times 36$  grid (generated with the *roulette*)
- 3 Count the number of points inside the disk
- $\Rightarrow$  Compute the ratio (estimated probability of being inside the disk while in the square):  $\hat{p}_C = \frac{\sum P_i \in circle}{\dots}$

## Monte Carlo method: illustration

### $\pi$ estimation:



A casino roulette (in Monte Carlo ?)



A 36×36 grid

If n = 1000 and 786 points are inside the disk:  $\hat{\pi} = 4 \times \frac{786}{1000} = 3.144$ 

One can improve the estimate by increasing:

- the grid resolution, and also
- the number of points sampled n:  $\lim_{n \to +\infty} \widehat{p}_C = p_C = \pi/4$

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Monte Carlo sample ⇒ compute various functions e.g.  $\pi = 4 \times$  the probability of being inside the disk

## Your turn!

Intro



Practical: exercise 1