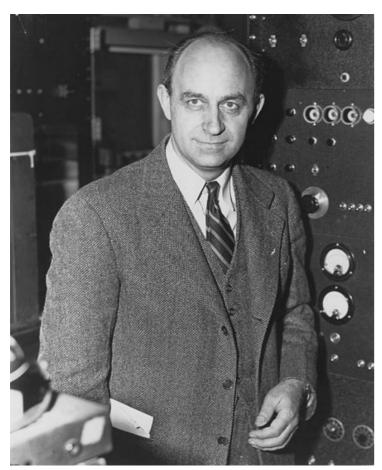
ENM531: Data-driven modeling and probabilistic scientific computing

Lecture #14: Sampling Methods



$$\mathbb{E}_{x \sim p(x)}[f(x)] = \int f(x)p(x)dx \approx \frac{1}{n} \sum_{i=1}^{n} f(x_i),$$

where  $x_i$  are drawn iid from p(x)







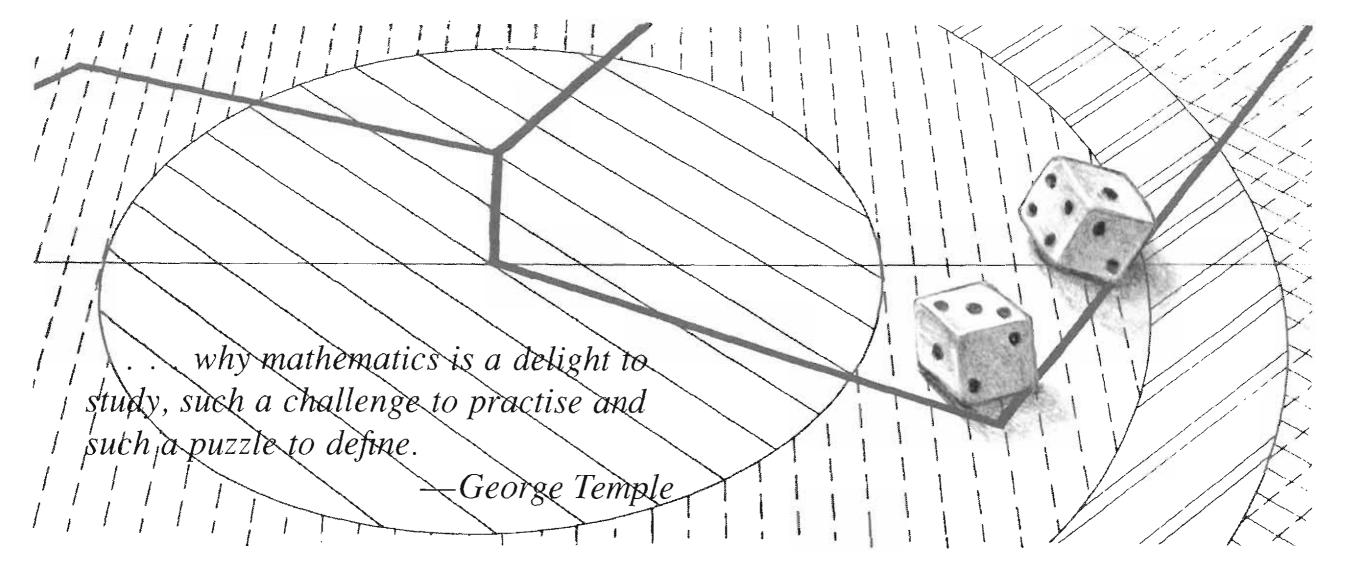
Enrico Fermi

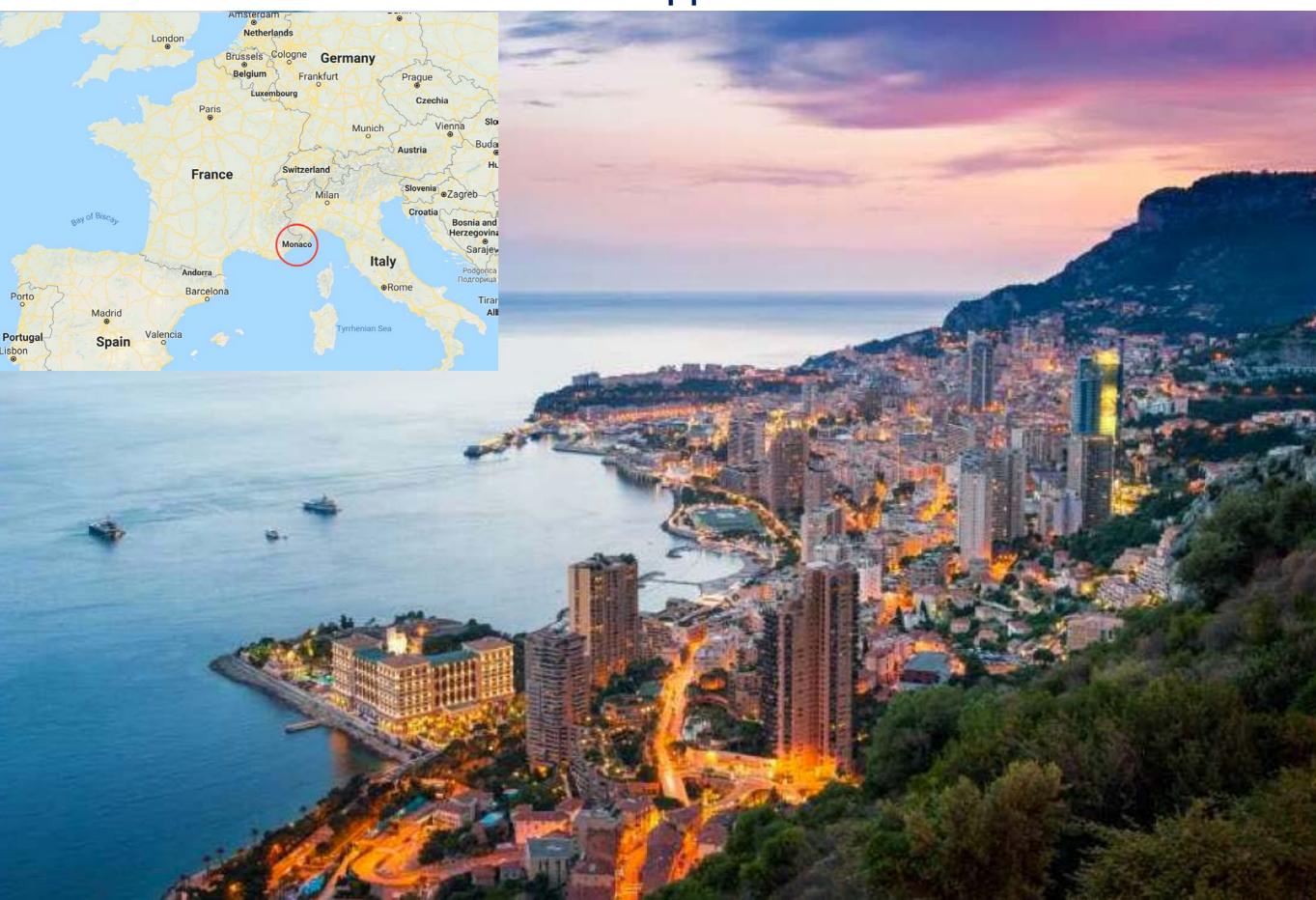
John von Neumann

Stan Ulam

# THE BEGINNING of the MONTE CARLO METHOD

by N. Metropolis





#### Example: estimating $\pi$ by Monte Carlo integration

MC approximation can be used for many applications, not just statistical ones. Suppose we want to estimate  $\pi$ . We know that the area of a circle with radius r is  $\pi r^2$ , but it is also equal to the following definite integral:

$$I = \int_{-r}^{r} \int_{-r}^{r} \mathbb{I}(x^2 + y^2 \le r^2) dx dy \tag{2.99}$$

Hence  $\pi = I/(r^2)$ . Let us approximate this by Monte Carlo integration. Let  $f(x,y) = \mathbb{I}(x^2 + y^2 \le r^2)$  be an indicator function that is 1 for points inside the circle, and 0 outside, and let p(x) and p(y) be uniform distributions on [-r, r], so p(x) = p(y) = 1/(2r). Then

$$I = (2r)(2r) \int \int f(x,y)p(x)p(y)dxdy$$

$$= 4r^{2} \int \int f(x,y)p(x)p(y)dxdy$$

$$\approx 4r^{2} \frac{1}{S} \sum_{i=1}^{S} f(x_{s},y_{s})$$
(2.100)
(2.101)
(2.102)

We find  $\hat{\pi} = 3.1416$  with standard error 0.09 (see Section 2.7.3 for a discussion of standard errors). We can plot the points that are accepted/ rejected as in Figure 2.19.