

# 7 Probability Review

PVK 2019: MAD

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# Schedule

## 1. Theory

1. Setup
2. Random variable
3. Marginalization, Conditional Probability, and Bayes Rule
4. Expectation Value and Variance
5. Sampling

## 2. Exercises

## 3. Homework

# Theory

# Setup

- Random variables
- Probability assignment
- Probability density
- Event Probability
- Joint Probability
- Conditional Probability
- Expectation value of  $x$  over  $p(x)$
- Variance of  $x$  over  $p(x)$

$x, y, \dots$

$p(x), x \in \{x_1, \dots\}$

$p(x), x \in [x^-, x^+]$

$P(E)$

$p(x, y, \dots)$

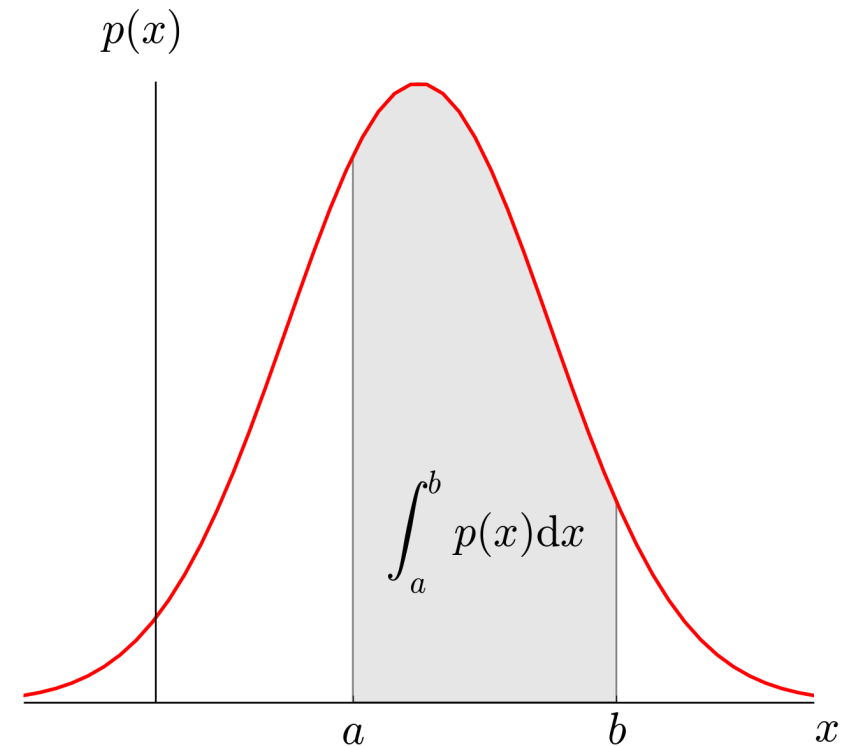
$p(x | y)$

$\mathbb{E}_{p(x)}[x]$

$Var(x) = \sigma^2$

# Random Variable

- Is not like a classical variable:
  - Classical variable: Defined by value
  - Random variable: Defined by distribution over values
- Discrete Random Variables
  - $p: \Omega \rightarrow \mathbb{R}$ , where eg.  $\Omega = \{0, 1, 2, \dots\}$
  - $\sum_{\Omega} p = 1, p \geq 0$
- Continuous Random Variables
  - $p: \Omega \rightarrow \mathbb{R}$ , where eg.  $\Omega = [0, 1]$
  - $\int_{\Omega} p = 1, p \geq 0$



# Marginalization, Conditional Probability, and Bayes Rule

- Marginalization:

$$p(x) = \int_y p(x, y) dy$$

- Conditional Probability

$$p(x | y) = \frac{p(x, y)}{p(y)}$$

- Bayes Rule:

$$p(x | y) = \frac{p(y | x)p(x)}{p(y)}$$

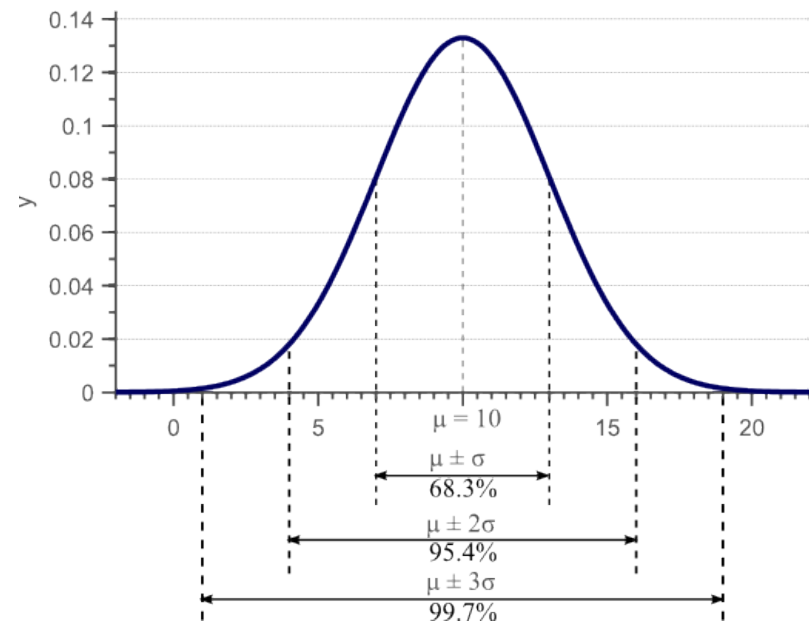
# Expectation Value and Variance

- Expectation value:

$$\mathbb{E}[f(x)] = \int_{-\infty}^{\infty} f(x)p(x)dx$$

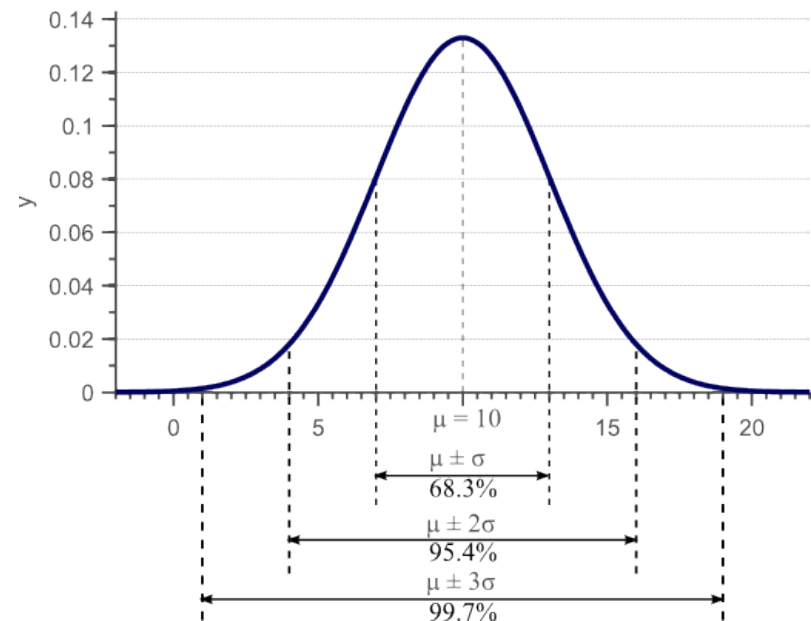
- Variance:

$$\begin{aligned} \text{Var}(f(x)) &= \sigma^2 \\ &= \mathbb{E}[f^2(x)] - \mathbb{E}^2[f(x)] \end{aligned}$$



# Sampling

- Generate samples from a random distribution
- Three approaches (covered):
  - Inverse CDF
  - Rejection Sampling
  - Markov Chain Monte Carlo





# Sampling cont. *Inverse CDF sampling*

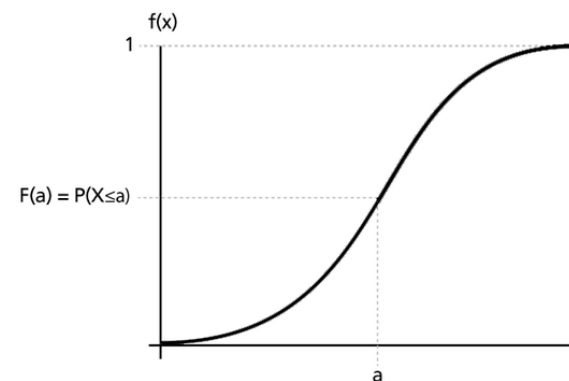
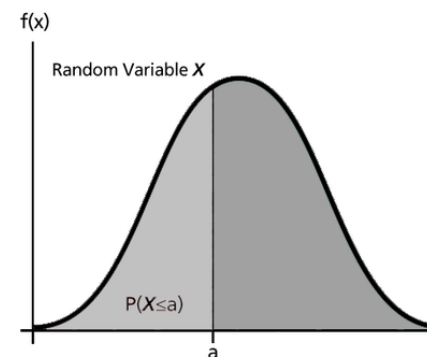
- Cumulative Distribution Function:

$$cdf(x) = \int_{-\infty}^x p(\bar{x})d\bar{x}$$

- Inverse CDF sampling:

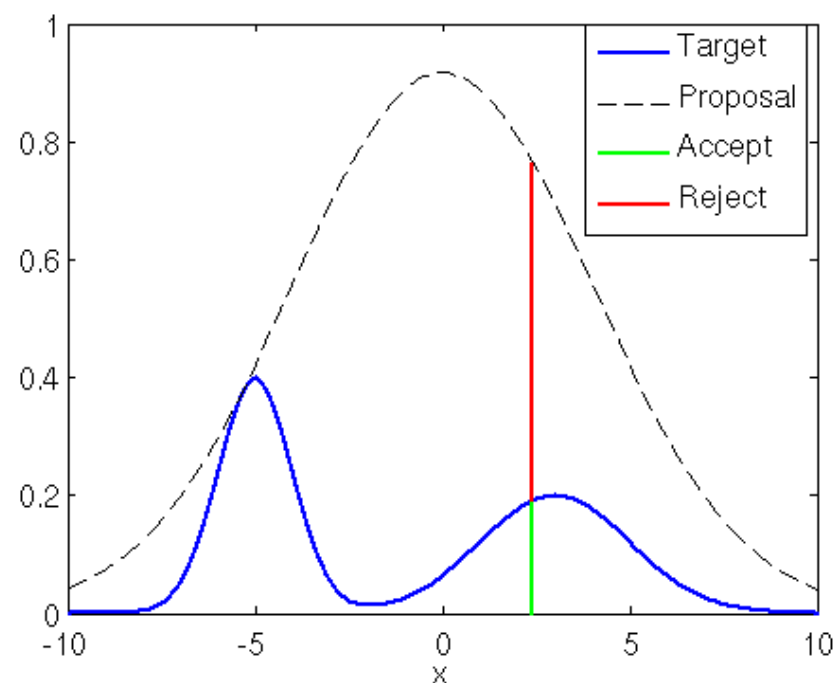
1. Sample  $u \sim \text{unif}(0,1)$
2. Generate sample  $x = cdf^{-1}(u)$

- Difficult: Have to compute integral & inverse!



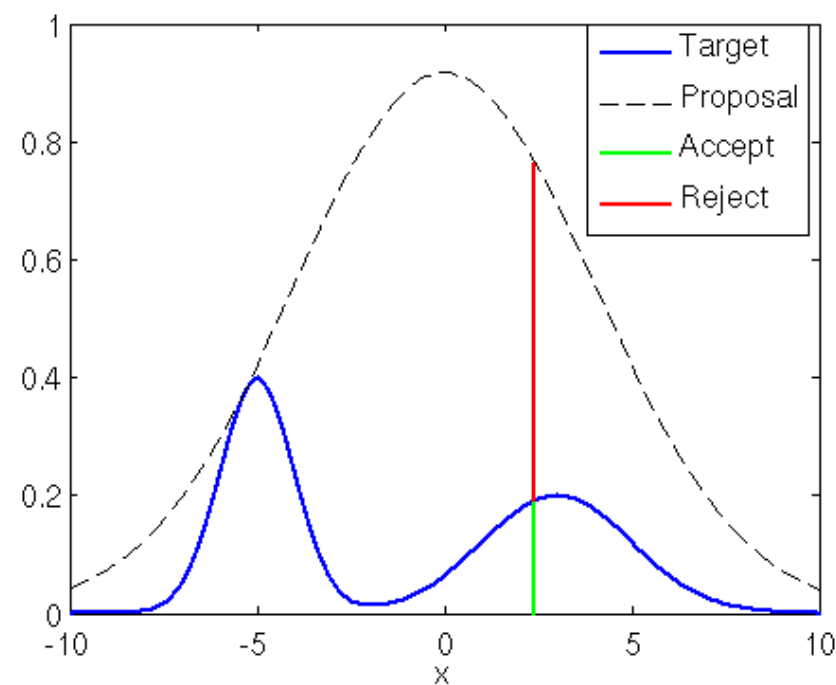
## Sampling cont. *Rejection Sampling*

- Define envelope “proposal”:  
$$p(x) \leq L \cdot q(x)$$
- Rejection Sampling:
  1. Draw  $y \sim q$  and  $u \sim \text{unif}(0,1)$
  2. If  $u \cdot L \cdot q(y) \leq p(y)$  keep the sample  $y$ , else discharge
- Note: We are not sampling from  $p$  directly, only evaluating!



## Sampling cont. *Rejection Sampling* Example

- Given:
  - We sample  $y = -5$  and  $u = 0.5$
- Task:
  - Do we keep the sample?

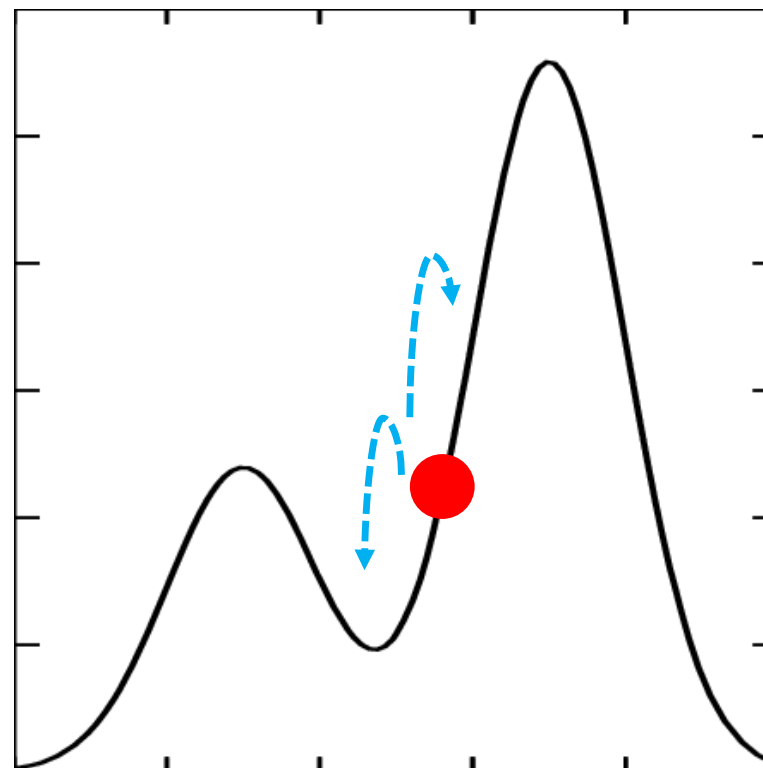


# Sampling cont. *MCMC*

- MCMC:

1. Sample  $y \sim \mathcal{N}(x, \sigma^2)$  and  $u \sim \text{unif}(0,1)$
2. If  $u \cdot p(x) \leq p(y)$  set  $x \leftarrow y$  else keep  $x \leftarrow x$

- Note: If  $p(y) \geq p(x)$  we always move to  $y$



## Sampling cont. *MCMC* Example

- Given:
  - Target distribution  $p(x) = 2x, x \in [0, 1]$
  - $x_1 = 0.5$
- Task (independent):
  - If we sample  $y = 0.6$  and  $u = 0.001$  do we move? How does the buffer look like?
  - If we sample  $y = 0.4$  and  $u = 0.9$  do we move? How does the buffer look like?

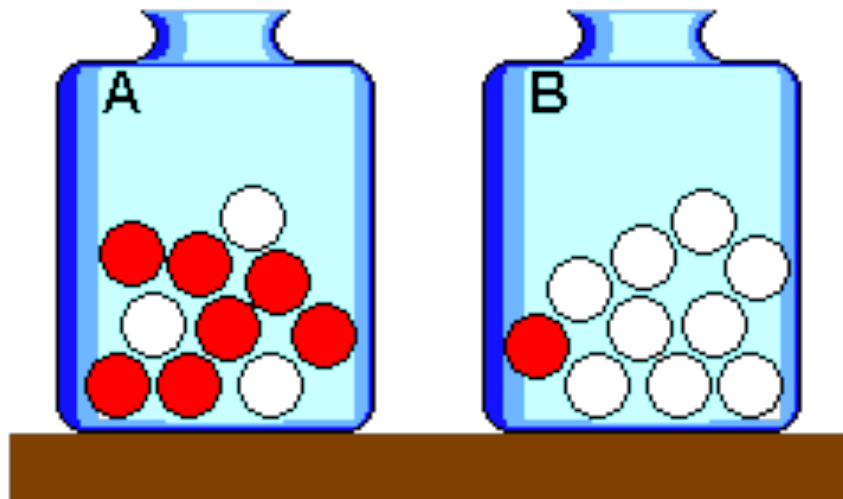
# Exercises

# Exercise 1

- Given:
  - PDF:  $p(x, y) = c, x \in [-1, 1], y \in [-1, 1]$
- Task:
  - Find  $c$  such that it's a valid PDF
  - Compute the probability that  $x$  is larger than 0,  $P(x > 0)$
- Tip:
  - $P(x > a) = \int_a^{\infty} p(x)dx$

## Exercise 2

- Given:
  - We have two jars ( $A, B$ ) with red ( $R$ ) and white ( $W$ ) marbles.
  - Drawing a marble from  $A$  or  $B$  is equally likely:  $P(A) = P(B) = \frac{1}{2}$
- Task:
  - Write out Bayes Rule with the appropriate random variables for  $P(A | R)$
  - Compute the normalization factor  $P(R)$
  - Compute  $P(A | R)$





# Exercise 3

- Task:
  - What could the expectation value and the variance roughly correspond to in something that you have seen in mechanics?

# Exercise 4

- Task:
  - Compute the expectation value of a dice
  - Compute the variance of a dice

# Exercise 5

- Task:
  - When and why do we use sampling?
  - Given an example of where you could use sampling?

# Homework

# HW 1

- Given:
  - $p(\text{„gold“} \mid \text{spam}) = 0.6$
  - $p(\text{„gold“} \mid \text{no spam}) = 0.01$
  - $p(\text{spam}) = 0.2$
- Task:
  - Find the probability that an email containing the word „gold“ is spam, ie.  $p(\text{spam} \mid \text{„gold“})$ .

# HW 2

- Given:

- We want to sample from a uniform distribution  $p(x) = \frac{1}{2}, x \in [-1, 1]$
- But can only generate samples from  $q(x) = 1/\sqrt{\pi} \cdot \exp -x^2$  (normal dist)

- Task:

- Determine an optimal  $L$  such that  $p(x) \leq L \cdot q(x)$ , use  $L_{optimal} = \left\| \frac{p(x)}{q(x)} \right\|_{\infty}$
- You draw  $y = 1.1$  and  $u = 0.9$  – is the sample accepted or rejected?

# HW 3

- Given:
  - The target distribution is  $p(x) = 2x, x \in [0, 1]$
  - You have access to the following values:
    - $x_1 = 0.5$
    - $s_1 = 0.1$  and  $s_2 = -0.2$  both drawn from  $s \sim \mathcal{N}(0, \sigma^2)$
    - $u_1 = 0.9$  and  $u_2 = 0.3$  both drawn from  $u \sim \text{unif}(0,1)$
- Task:
  - Perform MCMC for two steps
- Tip:
  - $y = \mu + s$  then  $y \sim \mathcal{N}(\mu, \sigma^2)$  if  $s \sim \mathcal{N}(0, \sigma^2)$

# HW 4

This problem was first stated in 1777 by Comte de Buffon

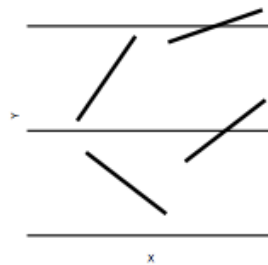


Figure 10: Buffon's needles

*Let a needle of length 1 be thrown at random onto a horizontal plane ruled with parallel straight lines spaced by a distance 1 from each other. What is the probability  $p$  that the needle will intersect one of these lines? (see Figure 10) Write a pseudo-code which estimate the probability  $p$ .*