



252R 기초 데이터 분석 및 실습

# 5. Mathematics (3)

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Welcome To Industrial Engineering



# Contents

252R 기초 데이터 분석 및 실습

- Intro
- Probability
- Statistic
- Distribution
- Optimization
- Term Project Description

## Intro

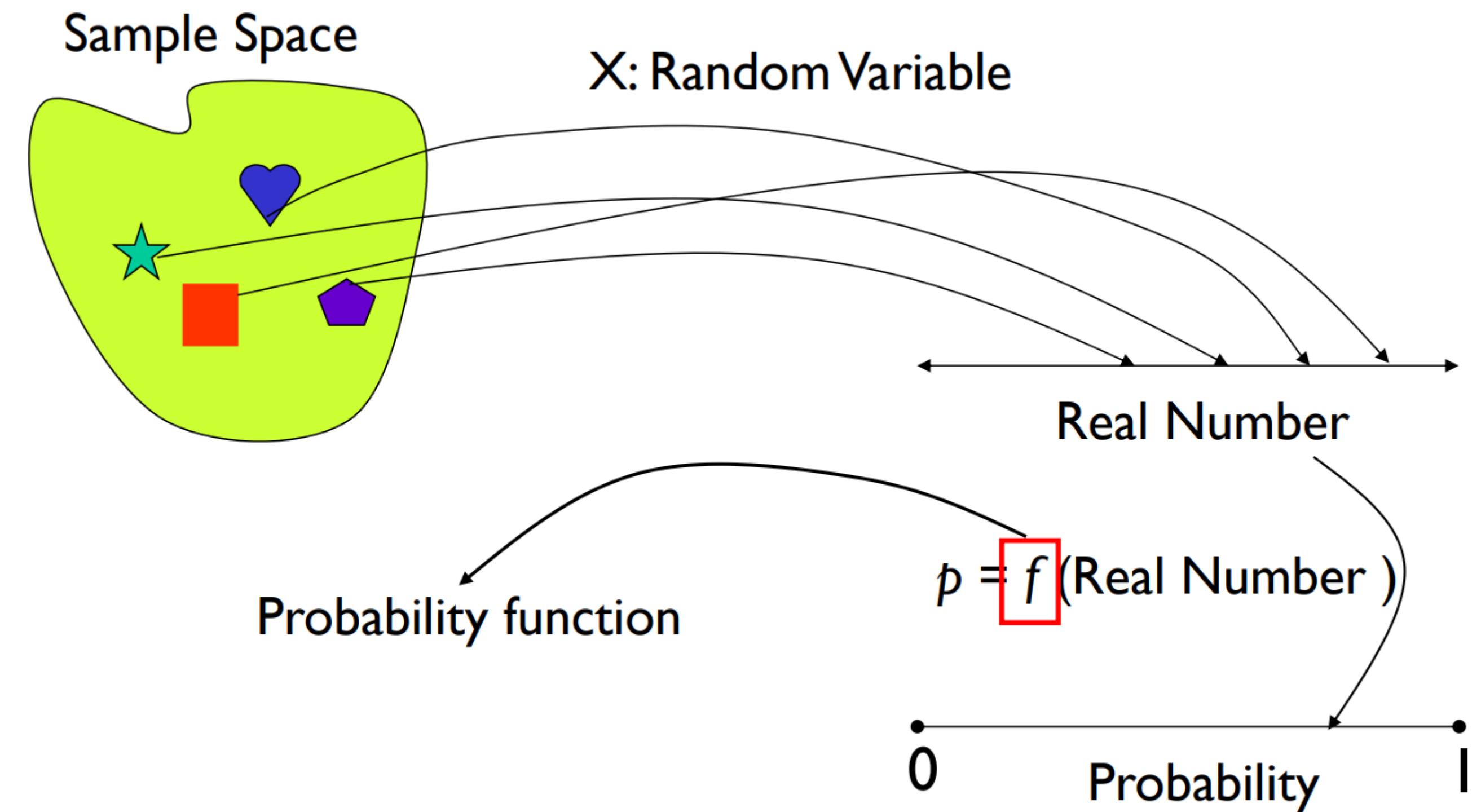
### Mathematics (3)

- 확률공간, 사건, 확률변수의 개념을 이해하고 각 관계를 설명할 수 있다.
- 이산/연속 확률변수를 구분하여 각 확률변수의 분포를 PMF/PDF로 나타낼 수 있다.
- 조건부확률을 계산하고 독립성의 정의와 직관을 이해한다.
- 기대값, 분산, 공분산의 정의를 알고, 이를 분포와 연관지어 이해한다.
- 주요 확률 분포의 모수와 특성을 이해하고, 각 분포가 어떤 데이터와 상황을 모델링하는지 설명할 수 있다.

# Probability Theory Basics

## Mathematics (3)

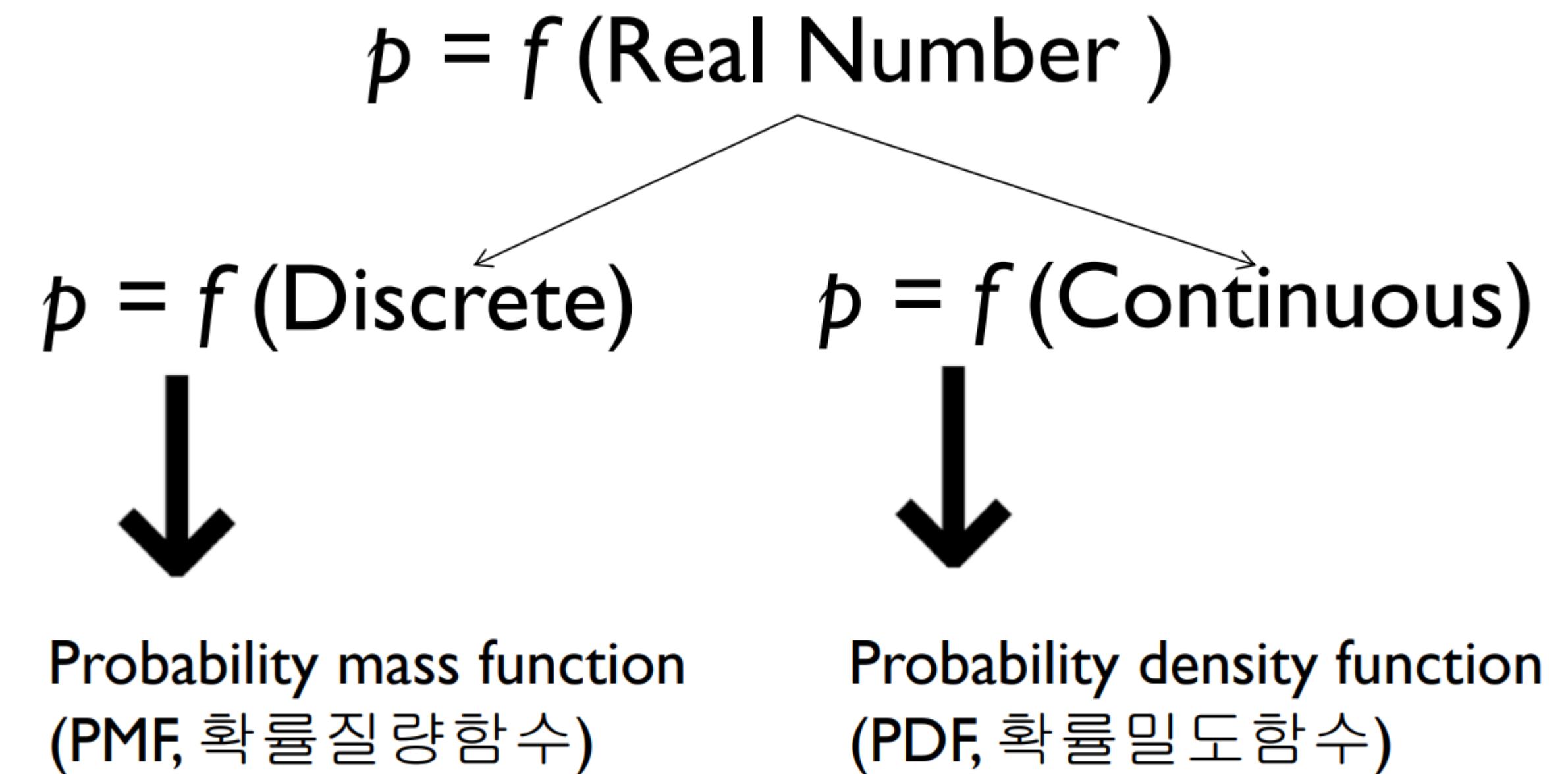
- Sample Space
- Event
- Random Variable
- Probability Function



# Probability Theory Basics

## Mathematics (3)

### ▪ Discrete or Continuous

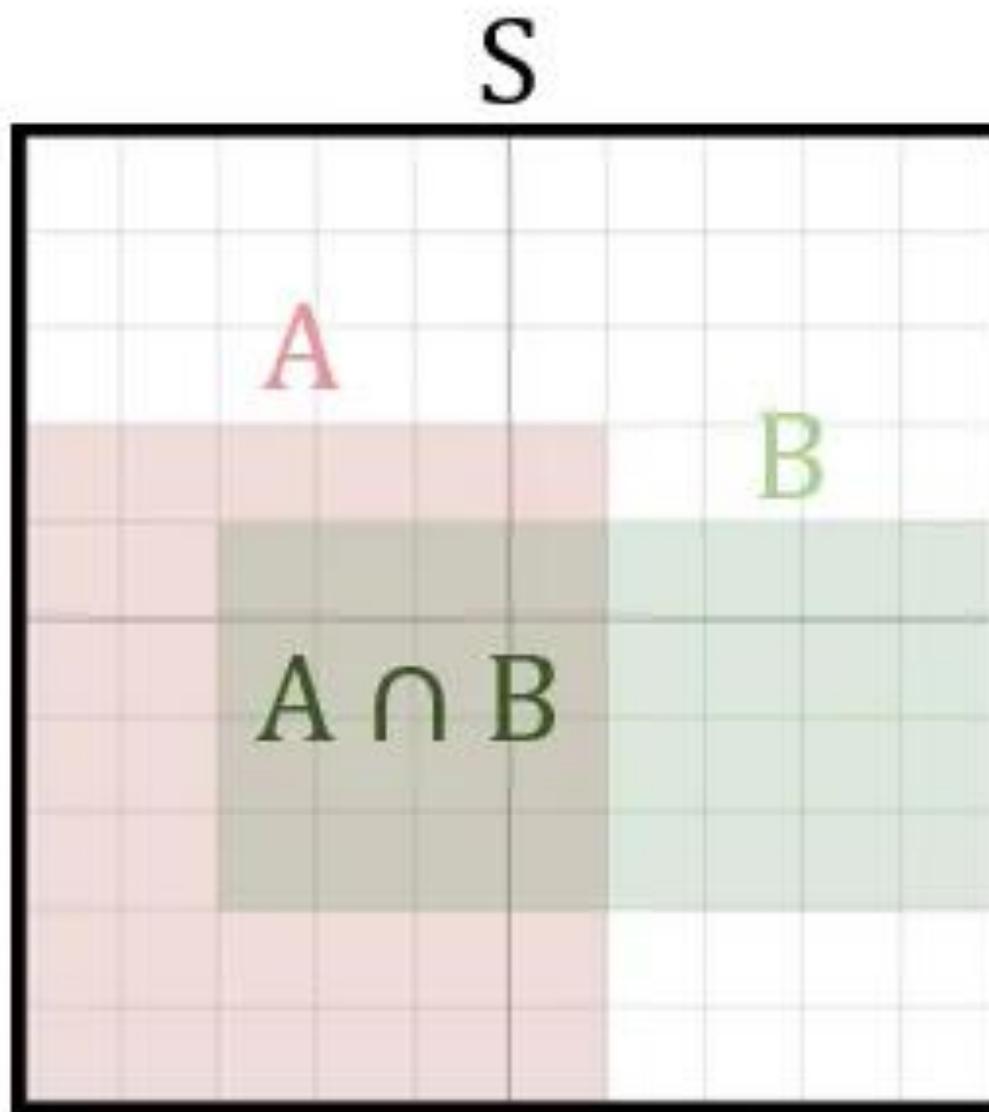


# Probability Theory Basics

## Mathematics (3)

### ▪ Conditional Probability

$$\begin{aligned} P(A|B) &= \frac{P(A \cap B)}{P(B)} = \frac{\frac{n(A \cap B)}{n(S)}}{\frac{n(B)}{n(S)}} \\ &= \frac{n(A \cap B)}{n(B)} \end{aligned}$$



# Summary Statistics

## Mathematics (3)

### ▪ Expectation, Variance, and Covariance(Correlation)

- Expectation of a discrete random variable,  $X$

$$E[X] = \sum_i x_i f_X(x_i)$$

- Expectation of a continuous random variable,  $X$

$$E[X] = \int x f_X(x) dx$$

# Summary Statistics

Mathematics (3)

## ▪ Expectation, Variance, and Covariance(Correlation)

$$\begin{aligned} V(X) &= E[\{X - E(X)\}^2] \\ &= E[X^2 - 2XE(X) + \{E(X)\}^2] \\ &= E(X^2) - 2\{E(X)\}^2 + \{E(X)\}^2 \\ &= E(X^2) - \{E(X)\}^2 \end{aligned}$$

$$SD[X] = \sqrt{V[X]}$$



$$Corr(X, Y) = \frac{Cov(X, Y)}{\sqrt{V(X) \cdot V(Y)}}$$

$$Cov(X, Y) = E[\{X - E(X)\}\{Y - E(Y)\}]$$

$$-1 \leq Corr(X, Y) \leq 1$$

# Key Probability Rules

## Mathematics (3)

- Sum Rule
- Product Rule
- Bayes' Theorem

$$p(\mathbf{x}) = \begin{cases} \sum_{\mathbf{y} \in \mathcal{Y}} p(\mathbf{x}, \mathbf{y}) & \text{if } \mathbf{y} \text{ is discrete} \\ \int_{\mathcal{Y}} p(\mathbf{x}, \mathbf{y}) d\mathbf{y} & \text{if } \mathbf{y} \text{ is continuous} \end{cases},$$

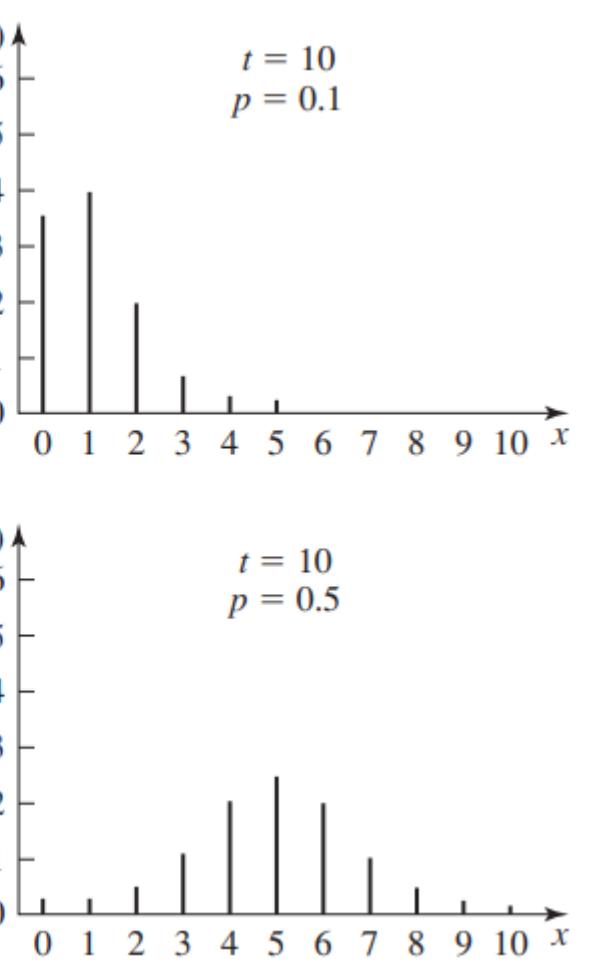
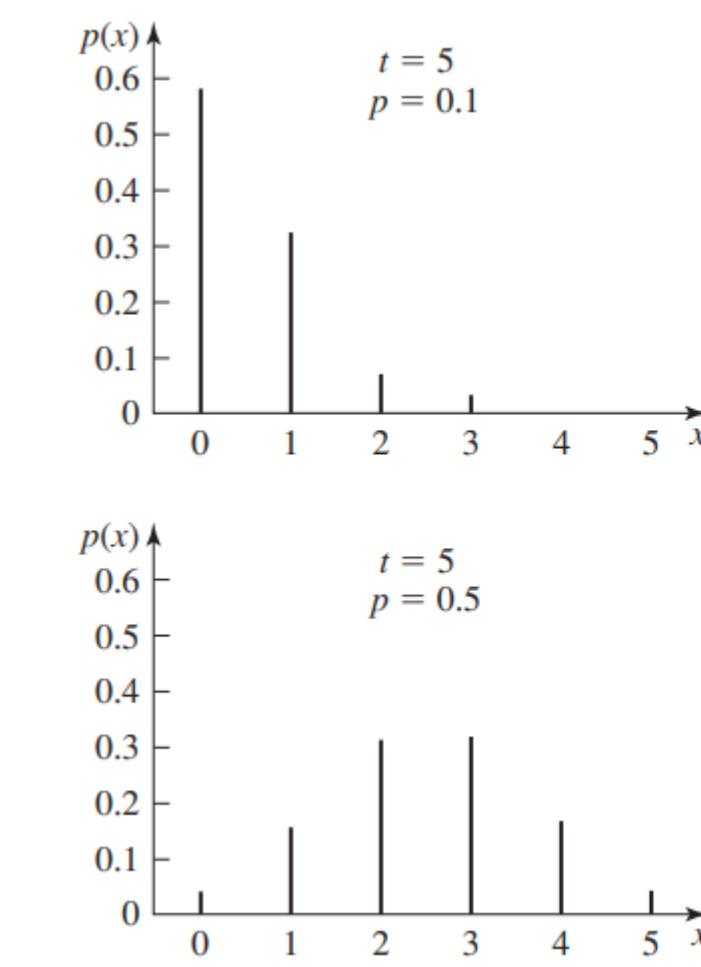
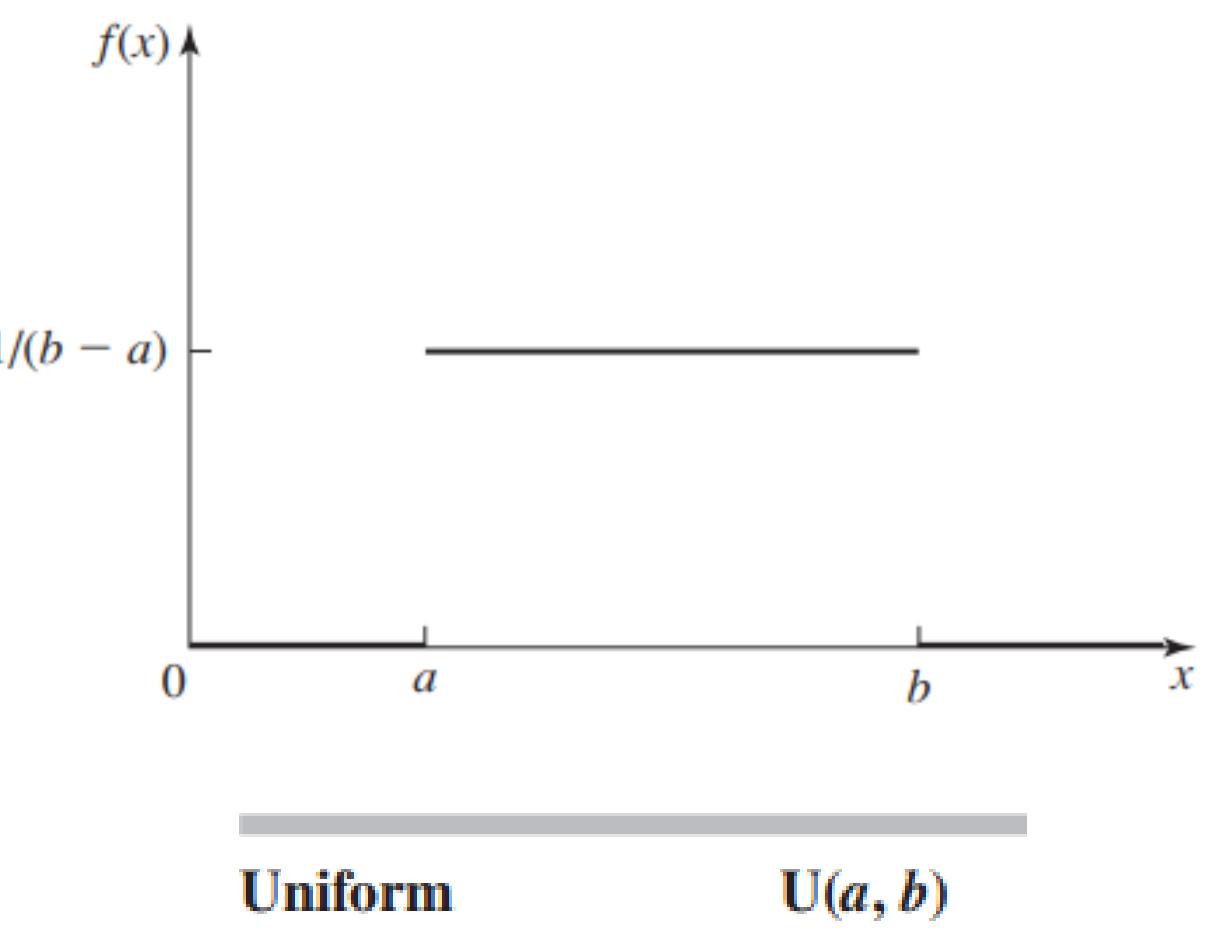
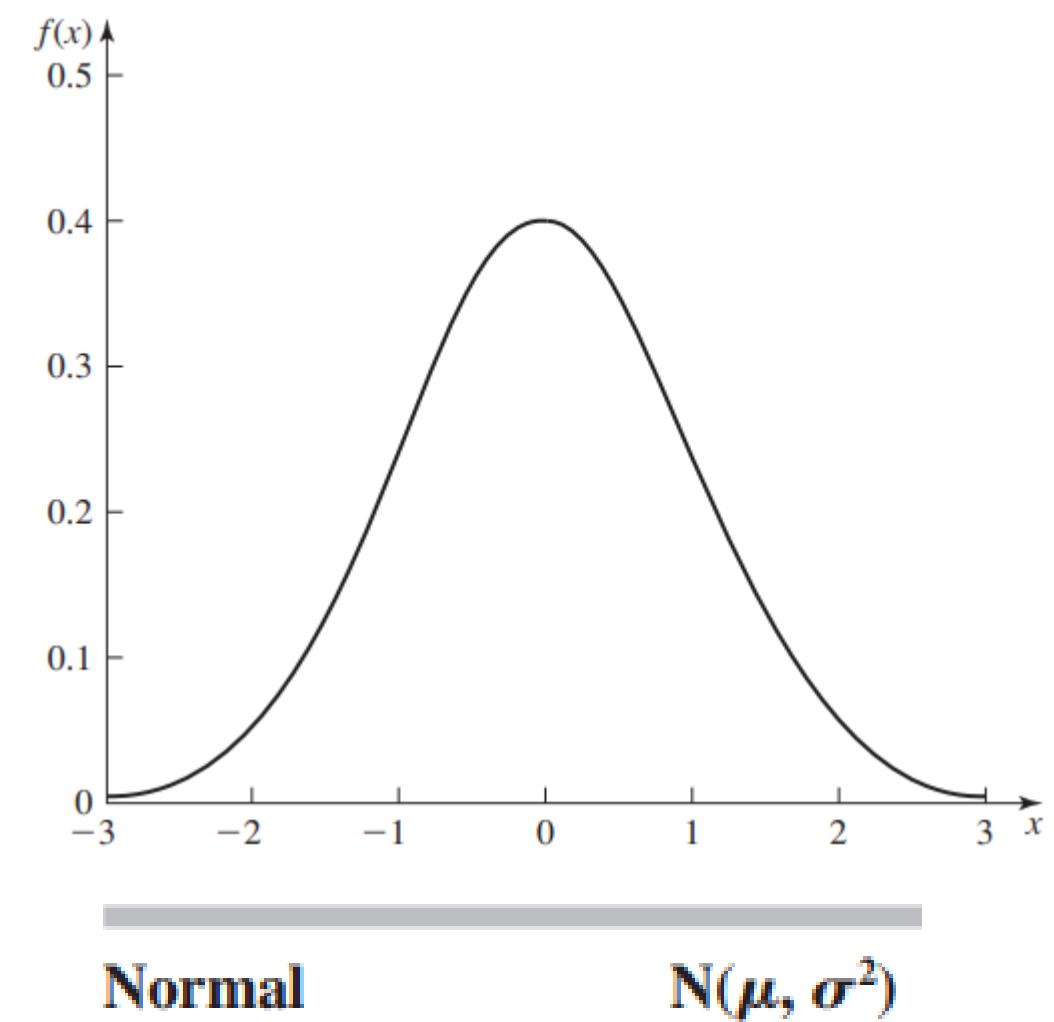
$$p(x_i) = \int p(x_1, \dots, x_D) dx_{\setminus i}$$

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

$$\underbrace{p(\mathbf{x} | \mathbf{y})}_{\text{posterior}} = \frac{\underbrace{p(\mathbf{y} | \mathbf{x})}_{\text{likelihood}} \underbrace{p(\mathbf{x})}_{\text{prior}}}{\underbrace{p(\mathbf{y})}_{\text{evidence}}}$$

# Examples of Distributions

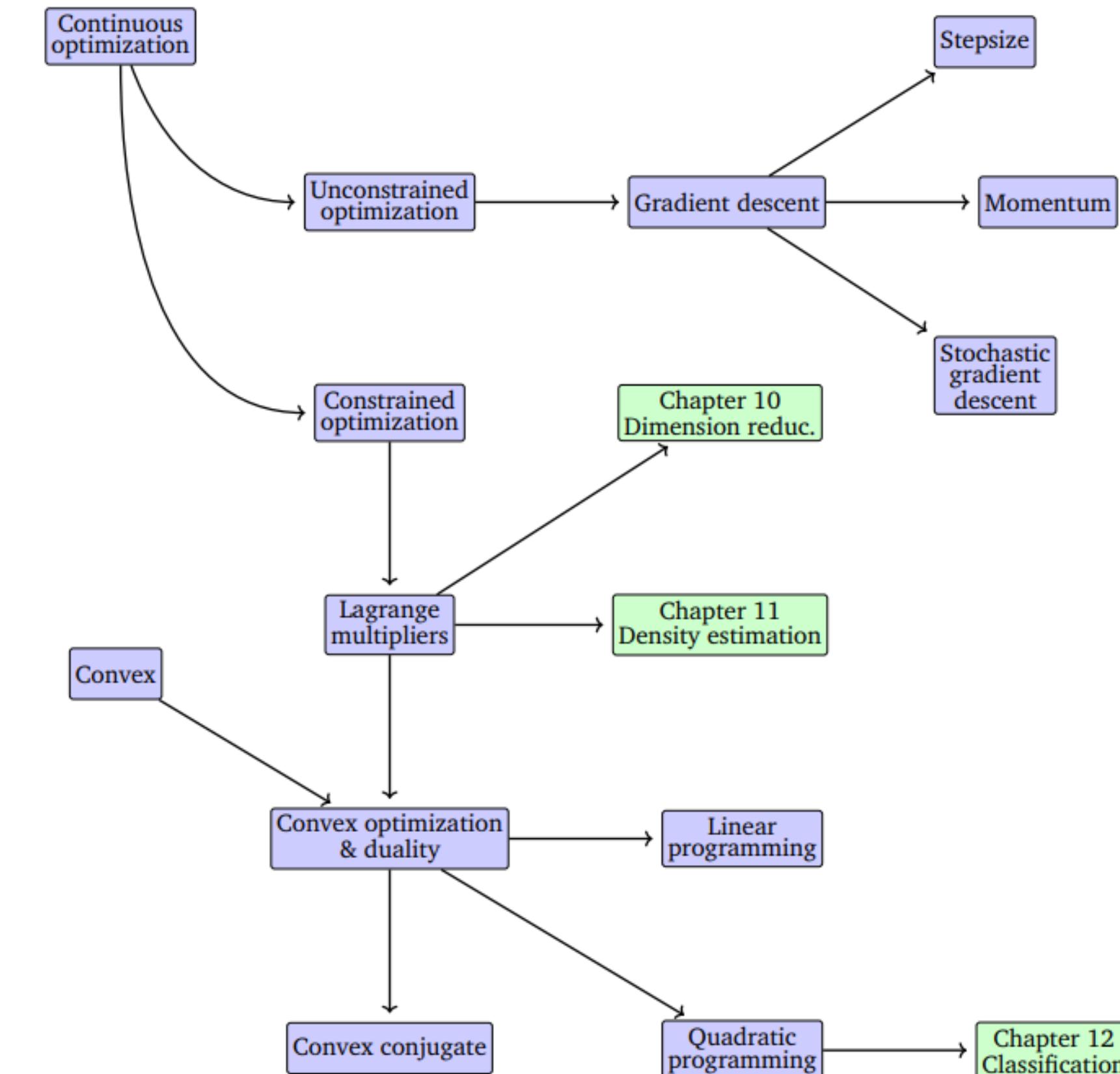
## Mathematics (3)



# Optimization

## Mathematics (3)

- Continuous?
- Constrained?
- Convex?



# Continuous Optimization

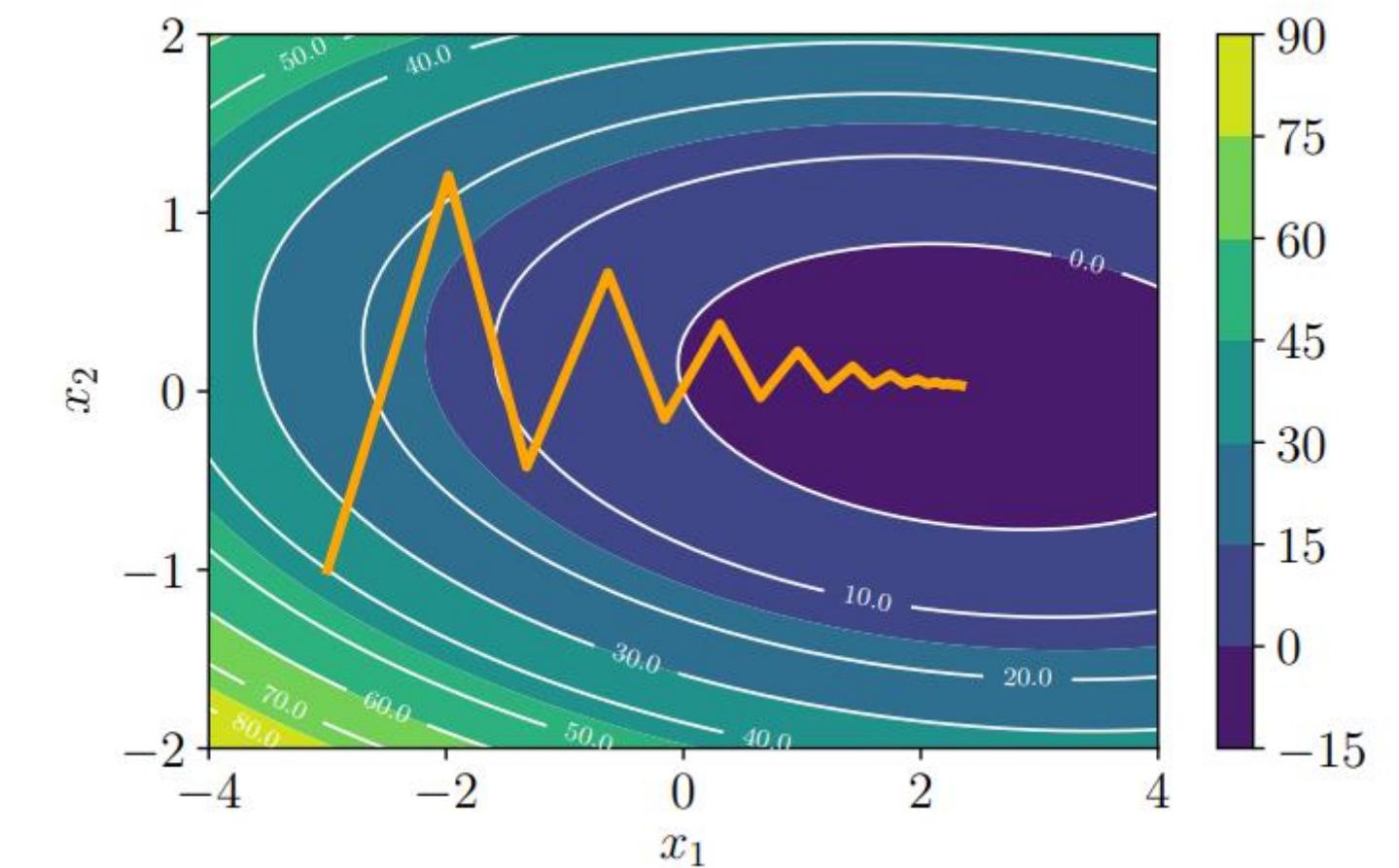
## Mathematics (3)

### ■ Can we differentiate?

- No... In general, we can't get analytical solution.

→ Heuristics!

→ There are many algorithms for approximation.

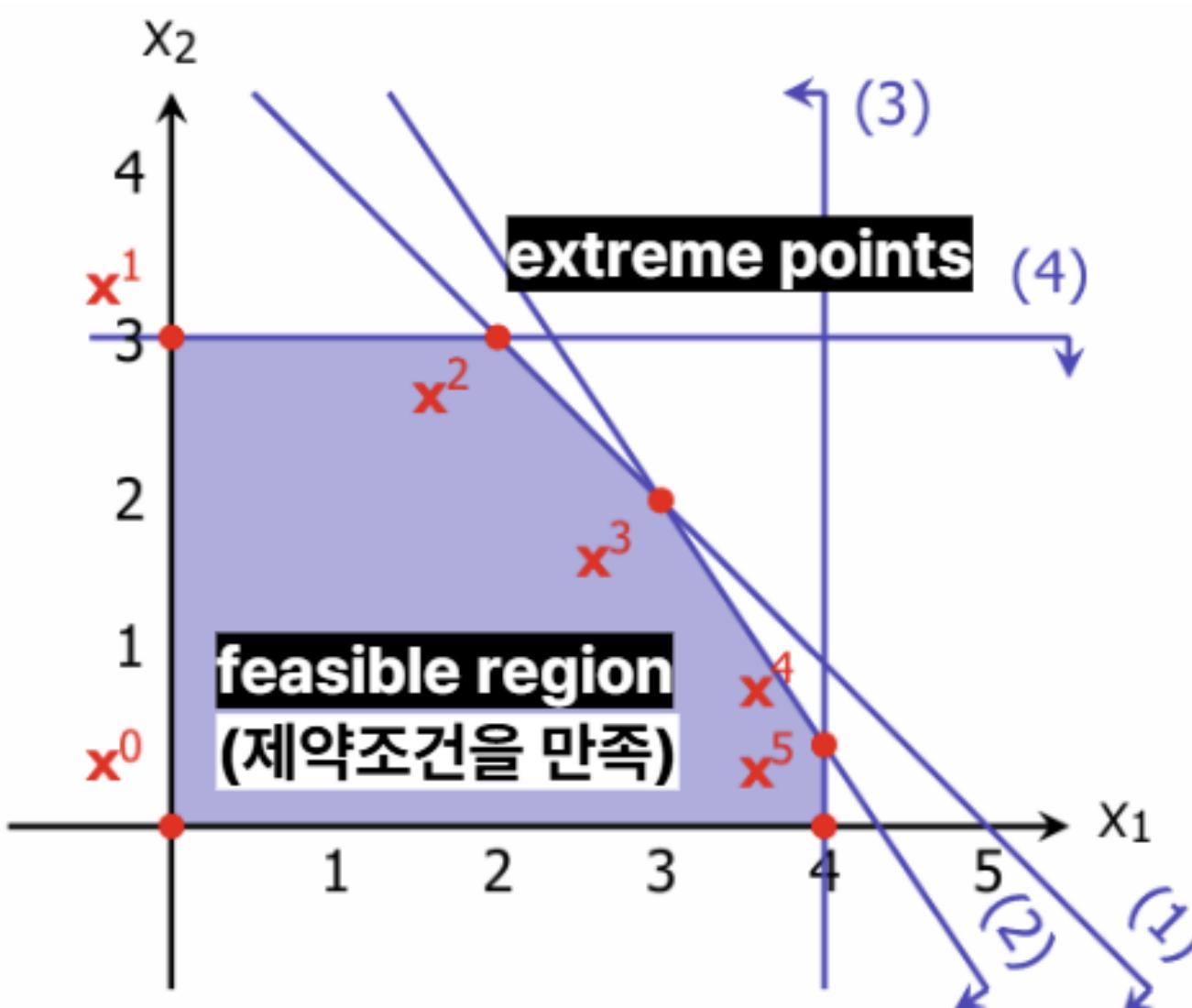


# Convex Optimization - Linear Programming

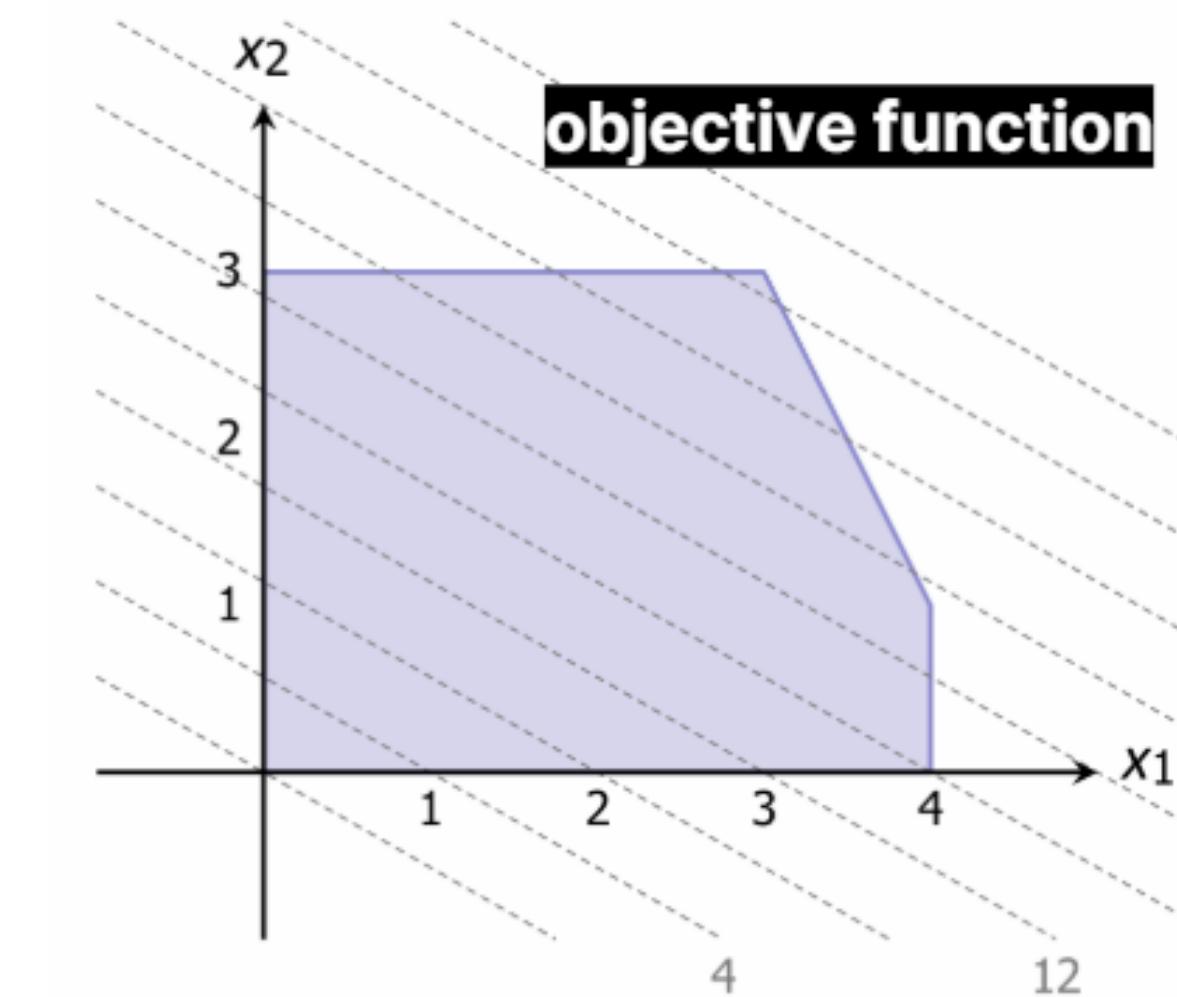
## Mathematics (3)

$$\begin{aligned}x_1 + x_2 &\leq 5 & (1) \\3x_1 + 2x_2 &\leq 13 & (2) \\x_1 &\leq 4 & (3) \\x_2 &\leq 3 & (4) \\x_1 &\geq 0 & (5) \\x_2 &\geq 0 & (6)\end{aligned}$$

**constraints**



→ feasible set에서  
목적함수를 최대화/최소화하는 해가 최적해(optimal)!



# Summary and Bridging

## Mathematics (3)

### ■ 선형대수

- 데이터와 모델을 설명하는 언어
- 행렬은 모든 데이터 과학의 근간

### ■ 벡터 미적분

- 다변량 함수를 다루는 방법
- 그래디언트와 헤시안

### ■ 확률과 분포

- 생성 모델
- 베이지안 추론

### ■ 최적화

- 학습 알고리즘

# Project Description

# QnA

## References

252R 기초 데이터 분석 및 실습

- Mathematics for Machine Learning Book
- A First Course in Probability
- Introduction to Operation Research



감사합니다