

# Notes for Operations Research & More

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**Part I**

**Preliminary Topics**





## Chapter 1

# Review of Linear Algebra



## Chapter 2

# Convex Sets



## Chapter 3

# Convex Functions and Generalizations



**Part II**

**Linear Programming**





# Chapter 4

## Formulation

### 4.1 Typical Problems

### 4.2 Formulation Skills

#### 4.2.1 Absolute Value

**Description:** Consider the following model statement:

$$\begin{aligned} \min \quad & \sum_{j \in J} c_j |x_j|, \quad c_j > 0 \\ \text{s.t.} \quad & \sum_{j \in J} a_{ij} x_j \gtrless b_i, \quad \forall i \in I \\ & x_j \text{ unrestricted}, \quad \forall j \in J \end{aligned}$$

**Modeling:**

$$\begin{aligned} \min \quad & \sum_{j \in J} c_j (x_j^+ + x_j^-), \quad c_j > 0 \\ \text{s.t.} \quad & \sum_{j \in J} a_{ij} (x_j^+ - x_j^-) \gtrless b_i, \quad \forall i \in I \\ & x_j^+, x_j^- \geq 0, \quad \forall j \in J \end{aligned}$$

#### 4.2.2 A Minimax Objective

**Description:** Consider the following model statement:

$$\begin{aligned} \min \quad & \max_{k \in K} \sum_{j \in J} c_{kj} x_j \\ \text{s.t.} \quad & \sum_{j \in J} a_{ij} x_j \gtrless b_i, \quad \forall i \in I \\ & x_j \geq 0, \quad \forall j \in J \end{aligned}$$

**Modeling:**

$$\begin{aligned} \min \quad & z \\ \text{s.t.} \quad & \sum_{j \in J} a_{ij} x_j \gtrless b_i, \quad \forall i \in I \\ & \sum_{j \in J} c_{kj} x_j \leq z, \quad \forall k \in K \\ & x_j \geq 0, \quad \forall j \in J \end{aligned}$$

### 4.2.3 A fractional Objective

**Description:** Consider the following model statement:

$$\begin{aligned}
 \min \quad & \frac{\sum_{j \in J} c_j x_j + \alpha}{\sum_{j \in J} d_j x_j + \beta} \\
 \text{s.t.} \quad & \sum_{j \in J} a_{ij} x_j \geq b_i, \quad \forall i \in I \\
 & x_j \geq 0, \quad \forall j \in J
 \end{aligned}$$

**Modeling:**

$$\begin{aligned}
 \min \quad & \sum_{j \in J} c_j x_j t + \alpha t \\
 \text{s.t.} \quad & \sum_{j \in J} a_{ij} x_j \geq b_i, \quad \forall i \in I \\
 & \sum_{j \in J} d_j x_j t + \beta t = 1 \\
 & t > 0 \\
 & x_j \geq 0, \quad \forall j \in J \\
 & (t = \frac{1}{\sum_{j \in J} d_j x_j + \beta})
 \end{aligned}$$

### 4.2.4 A range Constraint

**Description:** Consider the following model statement:

$$\begin{aligned}
 \min \quad & \sum_{j \in J} c_j x_j \\
 \text{s.t.} \quad & d_i \leq \sum_{j \in J} a_{ij} x_j \leq e_i, \quad \forall i \in I \\
 & x_j \geq 0, \quad \forall j \in J
 \end{aligned}$$

**Modeling:**

$$\begin{aligned}
 \min \quad & \sum_{j \in J} c_j x_j, \quad c_j > 0 \\
 \text{s.t.} \quad & u_i + \sum_{j \in J} a_{ij} x_j = e_i, \quad \forall i \in I \\
 & x_j \geq 0, \quad \forall j \in J \\
 & 0 \leq u_i \leq e_i - d_i, \quad \forall i \in I
 \end{aligned}$$

## Chapter 5

# Simplex Method

### 5.1 Basic Feasible Solutions and Extreme Points

### 5.2 Simplex Method

#### 5.2.1 Simplex Method Algorithm

#### 5.2.2 Simplex Method Tableau

#### 5.2.3 Simplex Method as a Search Algorithm

### 5.3 Revised Simplex Method

### 5.4 Simplex Method with Bounded Variables

### 5.5 Artificial Variable

#### 5.5.1 Two-Phase Method

#### 5.5.2 Big-M Method

#### 5.5.3 Single Artificial Variable Technique

### 5.6 Degeneracy and Cycling

#### 5.6.1 Degeneracy

#### 5.6.2 Cycling

#### 5.6.3 Cycling Prevention Rules

Lexicographic Rule

Bland's Rule

Successive Ratio Rule



## Chapter 6

# Duality Theory and Sensitivity Analysis



## Chapter 7

# Decomposition Principle





## Chapter 8

# Ellipsoid Algorithm



## Chapter 9

# Projective Algorithm



## Chapter 10

# Interior-Point Algorithm



## **Part III**

# **Geometry, Graph and Network Theory**





## Chapter 11

# Paths, Trees, and Cycles



## Chapter 12

# Shortest-Path Problem



## Chapter 13

# Minimum Spanning Tree Problem



## Chapter 14

# Maximum Flow Problem





## Chapter 15

# Minimum Cost Flow Problem



## Chapter 16

# Assignment and Matching Problem



## Chapter 17

# Graph Algorithms

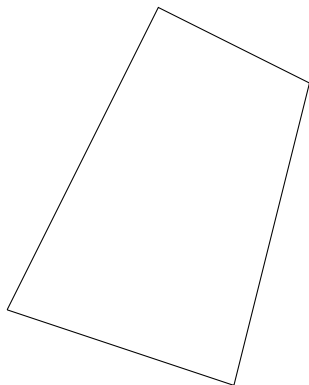


## Chapter 18

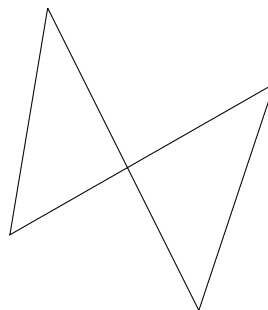
# Polygon Triangulation

### 18.1 Types of Polygons

**Def:** A **simple polygon** is a closed polygonal curve without self-intersection. Polygons are basic building blocks



Simple Polygon



Non-simple Polygon

in most geometric applications. It can model arbitrarily complex shapes, and apply simple algorithms and algebraic representation/manipulation.

### 18.2 Triangulation

**Def:** **Triangulation** is to partition polygon  $P$  into non-overlapping triangles using diagonals only. It reduces complex shapes to collection of simpler shapes. Every simple  $n$ -gon admits a triangulation which has  $n - 2$  triangles.

### 18.3 Art Gallery Theorem

**Scenario:** The floor plan of an art gallery modeled as a simple polygon with  $n$  vertices, there are guards which is stationed at fixed positions with 360 degree vision but cannot see through the walls. How many guards does the art gallery need for the security? (Fun fact: This problem was posted to Vasek Chvatal by Victor Klee in 1973)





## Part IV

# Integer and Combinatorial Programming



# Chapter 19

## Formulation

### 19.1 Typical Problems

### 19.2 Integer Programming Formulation Skills

#### 19.2.1 A Variable Taking Discontinuous Values

**Description:** In algebraic notation:

$$x = 0, \quad \text{or} \quad l \leq x \leq u$$

**Modeling:**

$$\begin{aligned} x &\leq uy \\ x &\geq ly \\ y &\in \{0, 1\} \end{aligned}$$

where

$$y = \begin{cases} 0, & \text{if } x = 0 \\ 1, & \text{if } l \leq x \leq u \end{cases}$$

#### 19.2.2 Fixed Costs

**Description:** In algebraic notation:

$$C(x) = \begin{cases} 0 & \text{for } x = 0 \\ k + cx & \text{for } x > 0 \end{cases}$$

**Modeling:**

$$\begin{aligned} C^*(x, y) &= ky + cx \\ x &\leq My \\ x &\geq 0 \\ y &\in \{0, 1\} \end{aligned}$$

where

$$y = \begin{cases} 0, & \text{if } x = 0 \\ 1, & \text{if } x \geq 0 \end{cases}$$

#### 19.2.3 Either-or Constraints

**Description:** In algebraic notation:

$$\sum_{j \in J} a_{1j}x_j \leq b_1 \quad \text{or} \quad \sum_{j \in J} a_{2j}x_j \leq b_2$$

**Modeling:**

$$\begin{aligned}
\sum_{j \in J} a_{1j}x_j &\leq b_1 + M_1y \\
\sum_{j \in J} a_{2j}x_j &\leq b_2 + M_1(1 - y) \\
y &\in \{0, 1\}
\end{aligned}$$

where

$$y = \begin{cases} 0, & \text{if } \sum_{j \in J} a_{1j}x_j \leq b_1 \\ 1, & \text{if } \sum_{j \in J} a_{2j}x_j \leq b_2 \end{cases}$$

Notice that the sign before  $M$  is determined by the inequality  $\geq$  or  $\leq$ , if it is “ $\geq$ ”, use “ $-$ ”, if it “ $\leq$ ”, use “ $+$ ”.

### 19.2.4 Conditional Constraints

**Description:** If constraint A is satisfied, then constraint B must also be satisfied

$$\text{If } \sum_{j \in J} a_{1j}x_j \leq b_1 \text{ then } \sum_{j \in J} a_{2j}x_j \leq b_2$$

The key part is to find the opposite of the first condition. We are using  $A \Rightarrow B \Leftrightarrow \neg B \Rightarrow \neg A$ . Therefore it is equivalent to

$$\sum_{j \in J} a_{1j}x_j > b_1 \text{ or } \sum_{j \in J} a_{2j}x_j \leq b_2$$

Furthermore, it is equivalent to

$$\sum_{j \in J} a_{1j}x_j \geq b_1 + \epsilon \text{ or } \sum_{j \in J} a_{2j}x_j \leq b_2$$

Where  $\epsilon$  is a very small positive number.

**Modeling:**

$$\begin{aligned}
\sum_{j \in J} a_{1j}x_j &\geq b_1 + \epsilon - M_2y \\
\sum_{j \in J} a_{2j}x_j &\leq b_2 + M_2(1 - y) \\
y &\in \{0, 1\}
\end{aligned}$$

### 19.2.5 Special Ordered Sets

**SOS1 Description** Out of a set of yes-no decisions, at most one decision variable can be yes.

$$\begin{aligned}
x_1 = 1, x_2 = x_3 = \dots = x_n = 0 \\
\text{or} \\
x_2 = 1, x_1 = x_3 = \dots = x_n = 0 \\
\text{or } \dots
\end{aligned}$$

**Modeling:**

$$\sum_i x_i = 1, \quad i \in N$$

**SOS2 Description 1** Out of a set of binary variables, at most two variables can be nonzero. In addition, the two variables must be adjacent to each other in a fixed order list.

**Modeling:** If  $x_1, x_2, \dots, x_n$  is a SOS2, then

$$\begin{aligned}
\sum_{i=1}^n x_i &\leq 2 \\
x_i + x_j &\leq 1, \forall i \in \{1, 2, \dots, n\}, j \in \{i+2, i+3, \dots, n\} \\
x_i &\in \{0, 1\}
\end{aligned}$$

**SOS2 Description 2** There is another type of definition, that is out of a set of nonnegative variables **not binary here**, at most two variables can be nonzero. In addition, the two variables must be adjacent to each other in a fixed order list. All variables summing to 1.

This definition of SOS2 is used in the following section *Piecewise Linear Formulations*

### 19.2.6 Piecewise Linear Formulations

**Description:** The objective function is a sequence of line segments, e.g.  $y = f(x)$ , consists  $k - 1$  linear segments going through  $k$  given points  $(x_1, y_1), (x_2, y_2), \dots, (x_k, y_k)$ .

Denote

$$d_i = \begin{cases} 1, & x \in (x_i, x_{i+1}) \\ 0, & \text{otherwise} \end{cases}$$

Then the objective function is

$$\sum_{i \in \{1, 2, \dots, k-1\}} y = d_i f_i(x)$$

**Modeling:** Given that objective function as a piecewise linear formulation, we can have these constraints

$$\begin{aligned} \sum_{i \in \{1, 2, \dots, k-1\}} d_i &= 1 \\ d_i &\in \{0, 1\}, i \in \{1, 2, \dots, k-1\} \\ x &= \sum_{i \in \{1, 2, \dots, k\}} w_i x_i \\ y &= \sum_{i \in \{1, 2, \dots, k\}} w_i y_i \\ w_1 &\leq d_1 \\ w_i &\leq d_{i-1} + d_i, i \in \{2, 3, \dots, k-1\} \\ w_k &\leq d_{k-1} \end{aligned}$$

In this case,  $w_i \in SOS2$  (second definition)

### 19.2.7 Conditional Binary Variables

**Description:** Choose at most  $n$  binary variable to be 1 out of  $x_1, x_2, \dots, x_m, m \geq n$ . If  $n = 1$  then it is SOS1.

**Modeling:**

$$\sum_{k \in \{1, 2, \dots, m\}} x_k \leq n$$

**Description:** Choose exactly  $n$  binary variable to be 1 out of  $x_1, x_2, \dots, x_m, m \geq n$

**Modeling:**

$$\sum_{k \in \{1, 2, \dots, m\}} x_k = n$$

**Description:** Choose  $x_j$  only if  $x_k = 1$

**Modeling:**

$$x_j = x_k$$

**Description:** “and” condition, iff  $x_1, x_2, \dots, x_m = 1$  then  $y = 1$

**Modeling:**

$$\begin{aligned} y &\leq x_i, i \in \{1, 2, \dots, m\} \\ y &\geq \sum_{i \in \{1, 2, \dots, m\}} x_i - (m - 1) \end{aligned}$$

### 19.2.8 Elimination of Products of Variables

**Description:** For variables  $x_1$  and  $x_2$ ,

$$y = x_1 x_2$$

**Modeling:** If  $x_1, x_2$  are binary, it is the same as “and” condition of binary variables. If  $x_1$  is binary, while  $x_2$  is continuous and  $0 \leq x_2 \leq u$ , then

$$y \leq ux_1$$

$$y \leq x_2$$

$$y \geq x_2 - u(1 - x_1)$$

$$y \geq 0$$

If both  $x_1$  and  $x_2$  are continuous, it is non-linear, we can use Piecewise linear formulation to simulate.

## Chapter 20

# Branch and Bound





## Chapter 21

# Branch and Cut



## Chapter 22

# Packing and Matching



## Chapter 23

# Traveling Salesman Problem



## Chapter 24

# Knapsack Problem





## Part V

# Nonlinear Programming



## Chapter 25

# KKT Optimality Conditions



## Chapter 26

# Lagrangian Duality



## Chapter 27

# Unconstrained Optimization





## Chapter 28

# Penalty and Barrier Functions



## Part VI

# Algorithms and Computational Complexity



## Chapter 29

# Computational Complexity



## Chapter 30

# Sorting

30.1 Elementary Sorting Algorithms

30.2 Heap-sort

30.3 Quick-sort

30.4 Sorting in Linear Time

30.5 Medians and Order Statistics





## Chapter 31

# Data Structures

31.1 Elementary Data Structures

31.2 Hash Tables

31.3 Binary Search Trees

31.4 Red-Black Trees

31.5 B-Trees

31.6 Fibonacci Heaps

31.7 van Emde Boas Trees



## Chapter 32

# Design and Analysis Techniques

32.1 Dynamic Programming

32.2 Greedy Algorithms

32.3 Amortized Analysis

32.4 Multi-threaded Algorithms

32.5 Matrix Operations

32.6 Polynomials and the FFT

32.7 Number-Theoretic Algorithms

32.8 String Matching

32.9 Computational Geometry

32.10 NP-Completeness

32.11 Approximation Algorithms



## Part VII

# Heuristics and Meta-heuristics



Part VIII

Game Theory





## Chapter 33

# Games with Ordinal Payoffs

**33.1 Ordinal Games in Strategic Form**

**33.2 Perfect-information Games**

**33.3 General Dynamic Games**



## Chapter 34

# Games with Cardinal Payoffs

34.1 Expected Utility Theory

34.2 Strategic-form Games

34.3 Extensive-form Games



## Chapter 35

# Knowledge, Common Knowledge, Beliefs

35.1 Common Knowledge

35.2 Adding Beliefs to Knowledge

35.3 Rationality



## Chapter 36

# Refinements of Subgame-perfect Equilibrium

36.1 Weak Sequential Equilibrium

36.2 Sequential Equilibrium

36.3 Perfect Bayesian Equilibrium





## Chapter 37

# Incomplete Information

37.1 Static Games

37.2 Dynamic Games

37.3 The Type-Space Approach



**Part IX**

**Decision Analysis**



## Part X

# Probability, Stochastic Processes and Markov Chains



## Chapter 38

# Probability





# Chapter 39

## Random Variables

### 39.1 Relationship between Some Random Variables

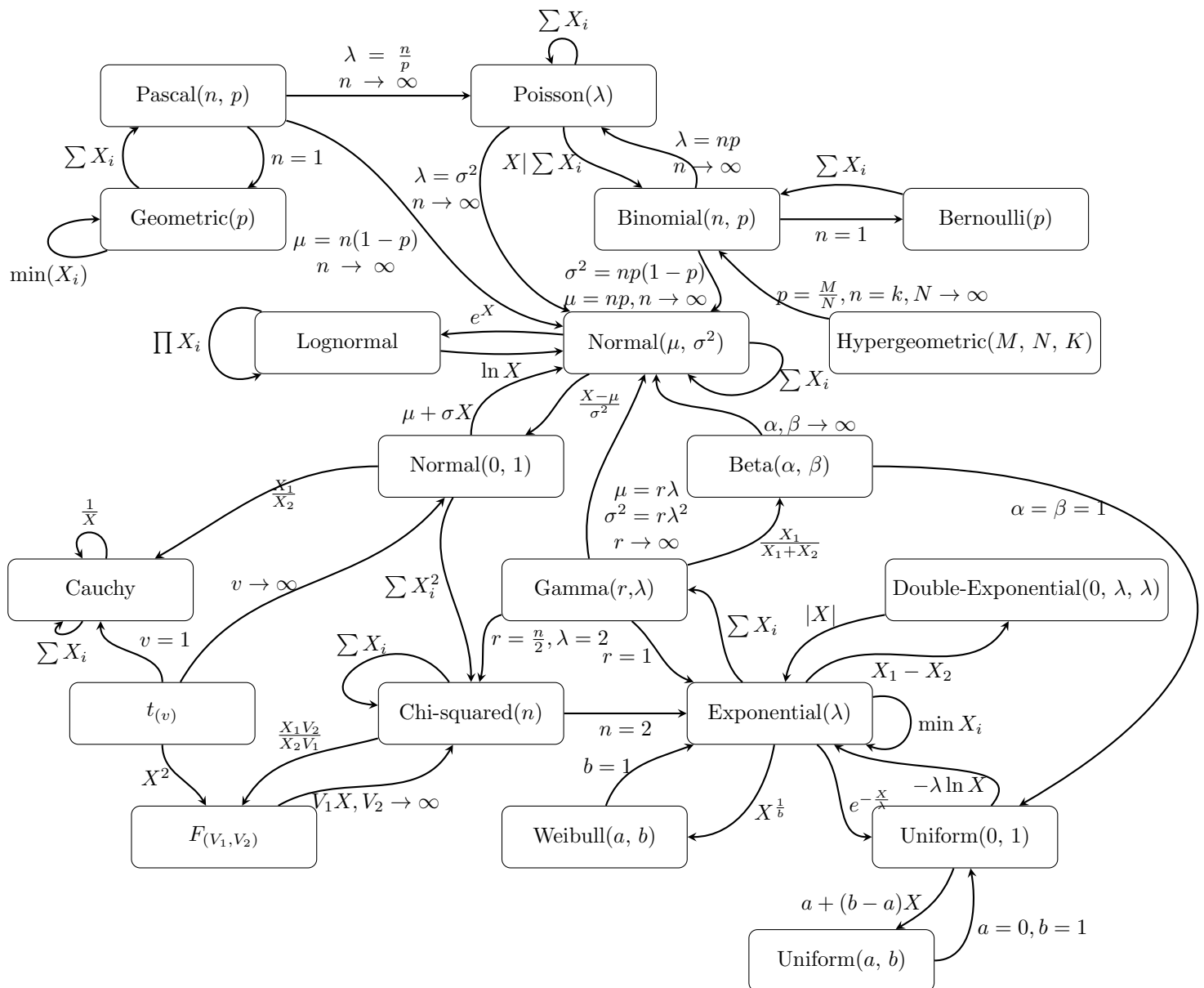


Figure 39.1: Relationship between Some Random Variables

## 39.2 Discrete Random Variables

Table 39.1: Discrete Random Variables

Distribution	PMF	CDF	Expectation	Variance	MGF
Discrete Uniform( $a, b$ )	$f(x) = \frac{1}{b-a+1}$ $x = a, a+1, \dots, b$	$F(x) = \frac{x-a+1}{b-a+1}$ $x = a, a+1, \dots, b$	$E[X] = \frac{b-a}{2}$	$D[X] = \frac{(b-a+1)^2-1}{12}$	$M(t) = \frac{e^{at}-e^{(b+1)t}}{(b-a+1)(1-e^t)}$ $t \in \mathbb{R}$
Bernoulli( $p$ )	$f(x) = p^x(1-p)^{1-x}$ $x \in \{0, 1\}$	$F(x) = \begin{cases} 0, & x < 0 \\ 1-p, & 0 \leq x \leq 1 \\ 1, & x > 1 \end{cases}$	$E[X] = p$	$D[X] = p(1-p)$	$M(t) = 1-p+pe^t$ $t \in \mathbb{R}$
Binomial( $n, p$ )	$f(x) = \binom{n}{x} p^x(1-p)^{n-x}$ $x = 0, 1, \dots, n$	$F(x) = \sum_{k=0}^x \binom{n}{k} p^k(1-p)^{n-k}$ $x = 0, 1, \dots, n$	$E[X] = np$	$D[X] = np(1-p)$	$M(t) = (1-p+pe^t)^n$ $t \in \mathbb{R}$
Poisson( $\mu$ )	$f(x) = \frac{\mu^x e^{-\mu}}{x!}$ $x = 0, 1, \dots, n, \dots$	$f(x) = \frac{\Gamma(x+1, \mu)}{\Gamma(x+1)}$ $x = 0, 1, \dots, n, \dots$	$E[X] = \mu$	$D[X] = \mu$	$M(t) = e^{\mu(e^t-1)}$ $t \in \mathbb{R}$
Geometric( $p$ )	$f(x) = p(1-p)^x$ $x = 0, 1, \dots, n, \dots$	$F(x) = 1 - (1-p)^{x+1}$ $x = 0, 1, \dots, n, \dots$	$E[X] = \frac{1-p}{p}$	$D[X] = \frac{1-p}{p^2}$	$M(t) = \frac{p}{1-(1-p)e^t}$ $t < -\ln(1-p)$
Pascal( $n, p$ )	$f(x) = \binom{n-1+x}{x} p^n(1-p)^x$ $x = 0, 1, 2, \dots, n, \dots$	$F(x) = 1 - I_p(k+1, n)$ $x = 0, 1, 2, \dots, n, \dots$	$E[X] = \frac{n(1-p)}{p}$	$D[X] = \frac{n(1-p)}{p^2}$	$M(t) = (\frac{p}{1-(1-p)e^t})^n$ $t < -\ln(1-p)$

## 39.3 Continuous Random Variables

Table 39.2: Continuous Random Variables

Distribution	PDF	CDF	Expectation	Variance	MGF
Uniform( $a, b$ )	$f(x) = \frac{1}{b-a}$ $x = [a, b]$	$F(x) = \frac{x-a}{b-a}$ $x = [a, b]$	$E[X] = \frac{b-a}{2}$	$D[X] = \frac{(b-a)^2}{12}$	$M(t) = \begin{cases} 1, & t=0 \\ \frac{e^{bt}-e^{at}}{t(b-a)}, & t \neq 0 \end{cases}$
Normal( $\mu, \sigma$ )	$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$ $x \in \mathbb{R}$	$F(x) = \int_{-\infty}^x \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$ $x \in \mathbb{R}$	$E[X] = \mu$	$D[X] = \sigma^2$	$e^{\frac{t(\mu\sigma^2+2\mu)}{2}}$ $t \in \mathbb{R}$
Exponential( $\lambda$ )	$f(x) = \lambda e^{-\lambda x}$ $x > 0$	$F(x) = 1 - e^{-\lambda x}$ $x > 0$	$E[X] = \frac{1}{\lambda}$	$D[X] = \frac{1}{\lambda^2}$	$\frac{1}{1-\frac{t}{\lambda}}$ $t < \lambda$
Erlang( $n, \lambda$ )	$f(x) = \frac{\lambda^n x^{n-1} e^{-\lambda x}}{(n-1)!}$ $x > 0$	$F(x) = 1 - \sum_{i=0}^{n-1} \frac{\lambda^i x^i e^{-\lambda x}}{i!}$ $x > 0$	$E[X] = \frac{n}{\lambda}$	$D[X] = \frac{n}{\lambda^2}$	$\frac{1}{(1-\frac{t}{\lambda})^n}$ $t < \lambda$

## Chapter 40

# Limit Theorems



## Chapter 41

# The Bernoulli and Poisson Process



## Chapter 42

# Discrete-Time Markov Chains





## Chapter 43

# Continuous-Time Markov Chains



Part XI

Queuing Theory



## Chapter 44

# Queuing Model



## Chapter 45

# Birth-and-Death Queuing Models





## Chapter 46

# Multidimensional Birth-and-Death Queuing Models



## Chapter 47

# Phase-Type Queue



## Chapter 48

# Bulk Queue



## Chapter 49

# Imbedded-Markov-Chain Queuing Models





## Chapter 50

# Queuing Network



Part XII

**Inventory Theory**



**Part XIII**

**Reliability Theory**



## Chapter 51

# Maintenance Optimization





**Part XIV**

**Bayesian Statistic**



Part XV

Classical Statistic



**Part XVI**

**Simulation**

