

Uncertainty in transportation

Probabilistic Concepts

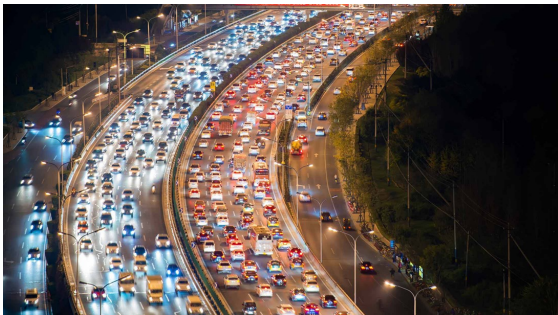
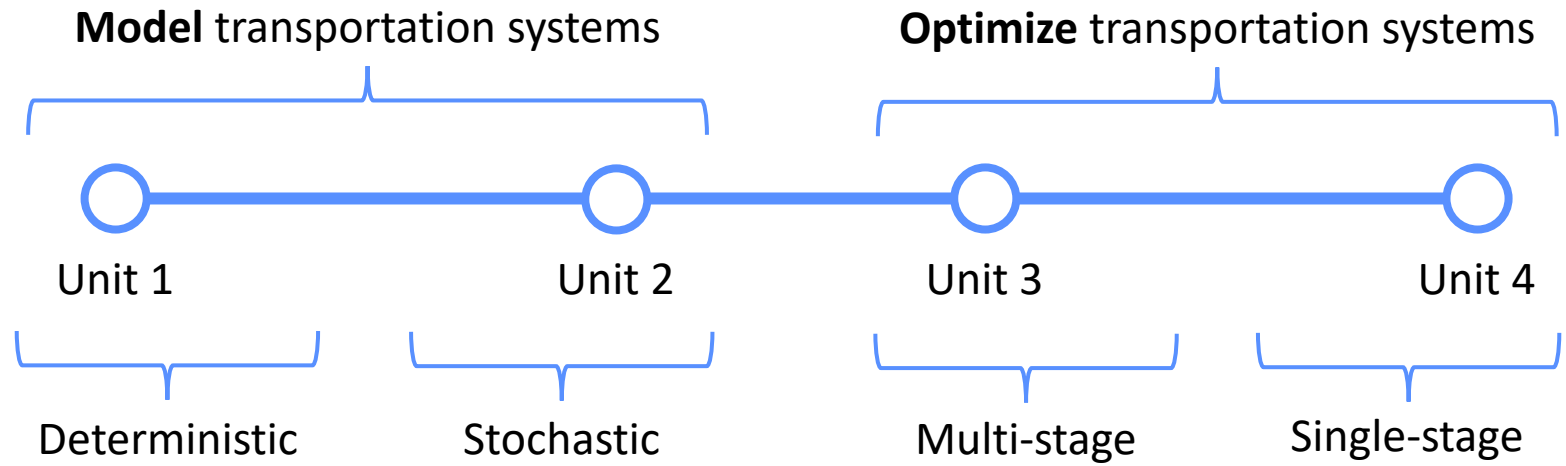
Cathy Wu

1.041/1.200 Transportation: Foundations and Methods

Readings

1. Larson, Richard C. and Amedeo R. Odoni. **Urban Operations Research**. Prentice-Hall (1981). Chapter 2: Probability. [URL](#).

1.041/1.200 overview



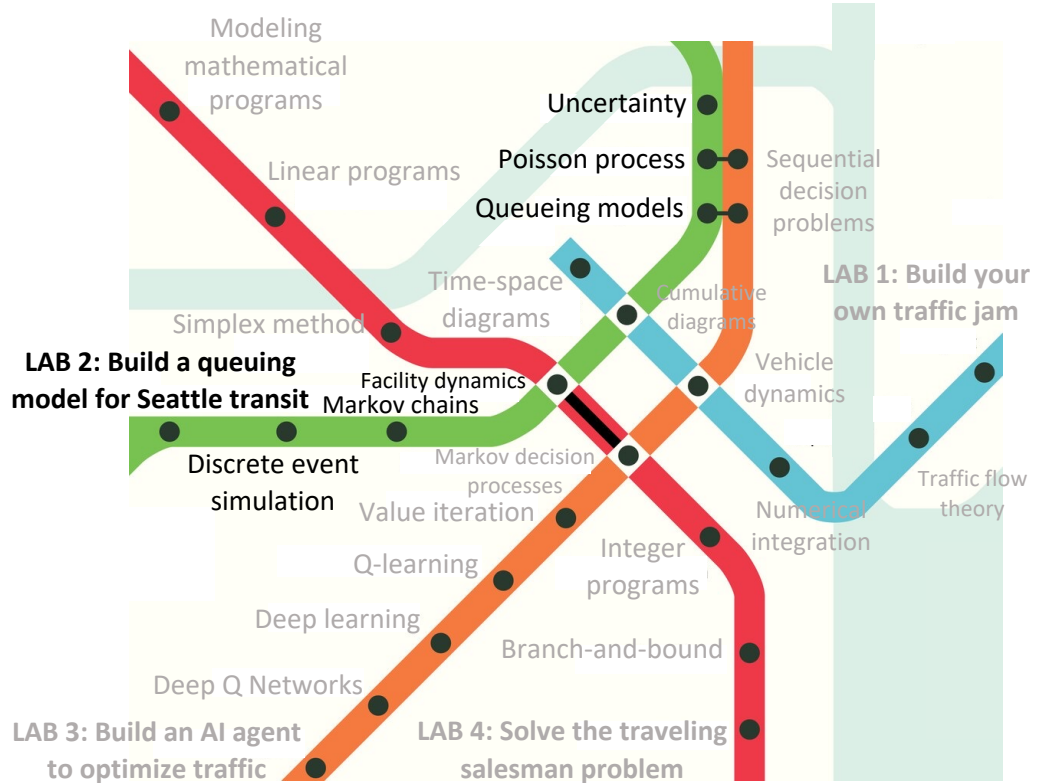
Unit 2: Queuing systems



Unit 2

Modeling

Stochastic



Outline

1. Uncertainty in transportation
2. Expected waiting time of a bus passenger problem

Outline

1. **Uncertainty in transportation**
2. Expected waiting time of a bus passenger problem

Uncertainty

- Decision-making typically takes place under uncertainty
- Today: deterministic → probabilistic concepts.
- What is the role of probability for modeling in transportation?

Examples: Sources of uncertainty in transportation

- Road traffic
 - Weather, road condition
 - Local events
 - Time of day (demands)
 - Day of week (demands)
 - Types of drivers on the road
 - Exact number of vehicles, driving behavior
- Bus stop
 - Exact arrivals of passengers
 - Road traffic
- Uber/Lyft ride
 - Availability of drivers
 - Other users making requests
 - Road traffic
 - Exact traffic signal timing
- Airline departure
 - Engine malfunction, safety checks
 - Boarding time
 - Run-way availability

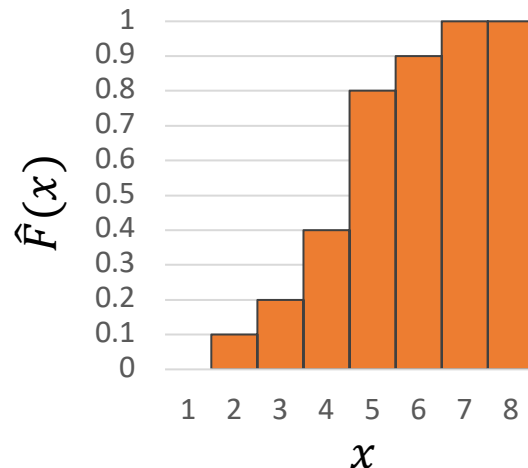
Empirical (cumulative) distribution function

- From relative frequencies:
 - Example: k = time headway (minutes) of bus arrivals
- Compute cumulative frequencies:
- Yielding **empirical cumulative distribution function (cdf)**:

$$\hat{F}(x) \begin{cases} 0 & : -\infty < x < 2 \\ c_k & : k \leq x < k + 1, k = 2, \dots, 6 \\ 1 & : 7 \leq x < \infty \end{cases}$$

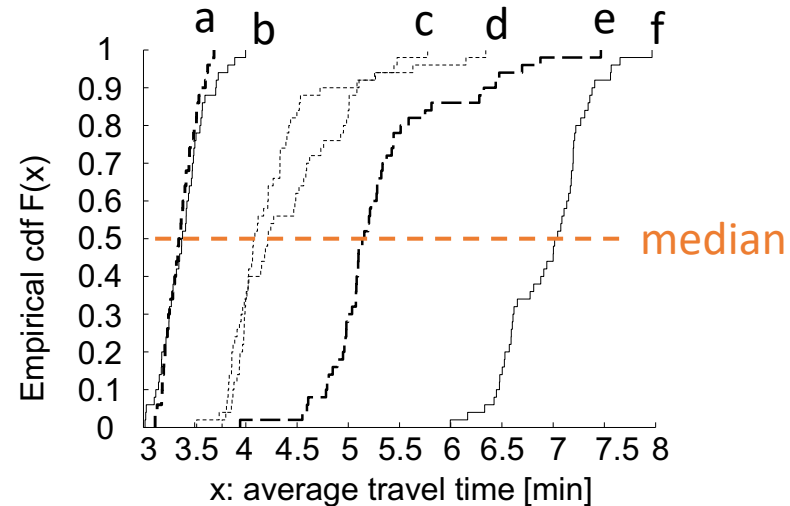
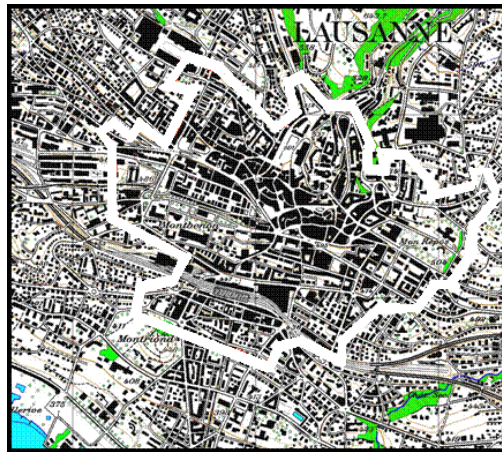
- $\hat{F}(x)$ is the proportion of observations smaller or equal to x .

k	2	3	4	5	6	7
$v_k = \frac{n_k}{n}$.1	.1	.2	.4	.1	.1
k	2	3	4	5	6	7
c_k	.1	.2	.4	.8	.9	1



Example: traffic signal plans

- Road network of Lausanne, Switzerland
- Empirical cdfs of the average travel times under **different signal plans**
- Goes beyond comparing averages
- Information on the variance (e.g. travel time reliability)



Outline

1. Uncertainty in transportation
2. **Expected waiting time of a bus passenger problem**

Expected waiting time

- Every morning, you arrive at a bus stop and wait patiently for the bus to arrive.
- A bus is scheduled to arrive every 20 minutes.
- Why does it feel like you wait on average more than 10 minutes?



Expected waiting time phenomenon

- Expected waiting time for a randomly arriving prospective rider
- Consider a prospective rider on this bus line who arrives at the stop at a random time, independently of the bus schedule (or without knowledge of the schedule).
- Suppose buses arrive at the stop, on average, every $E[X]$ time units.
 - X is called the bus headway.
 - Let σ_X be the bus headway variance.
- What then is the expected waiting time of this prospective rider until the next bus arrives?
 - Hint: The answer is not $E[X]/2$. Why not?

Expected waiting time phenomenon

- Expected waiting time for a randomly arriving prospective rider
- Consider a prospective rider on this bus line who arrives at the stop at a random time.
- Suppose buses arrive at the stop, on average, every $E[X]$ time units.
 - X is called the bus headway.
 - Let σ_x be the bus headway variance.
- What is the expected waiting time of this prospective rider until the next bus arrives?

Expected waiting time phenomenon

- Expected waiting time for a randomly arriving prospective rider
- Consider a prospective rider on this bus line who arrives at the stop at a random time.
- Suppose buses arrive at the stop, on average, every $E[X]$ time units.
 - X is called the bus headway.
 - Let σ_x be the bus headway variance.
- What is the expected waiting time of this prospective rider until the next bus arrives?

Inspection paradox

- This is an example of a problem involving “random incidence”, “inspection paradox”.
 - See Chap. 2 in R.C. Larson A. Odoni, Urban Operations Research (2007)
- The **inspection paradox** arises whenever the probability of observing a quantity is related to the quantity being observed.

References

1. Larson, Richard C. and Amedeo R. Odoni. **Urban Operations Research**. Prentice-Hall (1981). Chapter 2: Probability.
2. Lectures Modeling and Simulation in Logistics (2007) by Profs. Thomas Liebling and Carolina Osorio
3. Lecture Mesoscopic Simulation, Prof. Michel Bierlaire
4. Slides adapted from Carolina Osorio.

Appendix: Probability Review

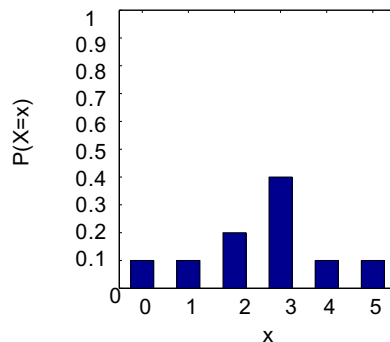
Discrete random variable

- Random variable: outcome of random experiment
- Discrete random variable (r.v.) X, Y , etc.
- Realizations: x, y , etc.
- Defined over a set of discrete values: $\{0, \dots, K\}$
- Is the outcome of a random experiment according to the probability distribution, called the **probability mass function** (pmf):

$$P(X = k) = f_k, \quad k = 0, \dots, K$$

Such that

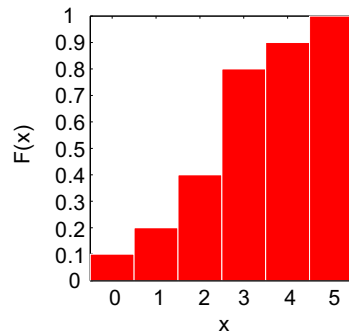
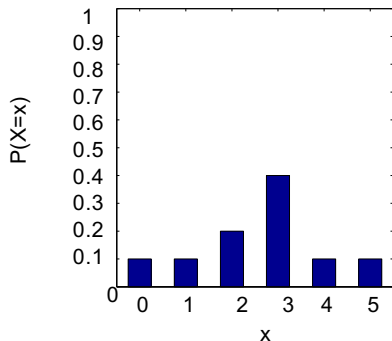
$$\begin{cases} f_k \geq 0, k = 0, \dots, K \\ \sum_{k=0}^K f_k = 1 \end{cases}$$



Cumulative distribution function (cdf): $F(x)$

$$F(x) = P(X \leq x) = \sum_{k \leq x} f_0 + \dots + f_{\lfloor x \rfloor}$$

$F(x)$ is the probability that X is smaller or equal to x .



Example. For $x = 2$, $F(x) = f_0 + f_1 + f_2$

$F(x)$ is such that:

- **Bounded:** $0 \leq f(x) \leq 1$
- **Monotone increasing:** If $x \leq y$, then $F(x) \leq F(y)$

Expected value and variance

- Expectation:

- $E[X] = \sum_{k=0}^K k f_k$

- Variance:

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

- Squared deviation of a random variable from its mean

- For a discrete distribution with potential outcomes $k = 0, \dots, K$:

- $$\begin{aligned} \text{Var}[X] &= \sum_{k=0}^K (k - E[X])^2 f_k \\ &= \sum_{k=0}^K k^2 f_k - E[X]^2 \end{aligned}$$

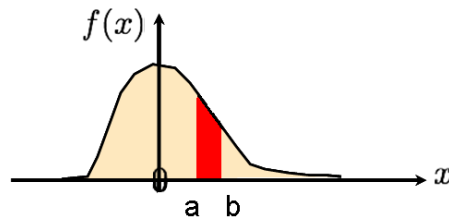
- Standard deviation:

- $\sigma = \sqrt{\text{Var}[X]}$

Continuous random variable

- Continuous random variable (r.v.) X, Y , etc.
- Realizations: x, y , etc.
- Probability density function: $f_X(x)$
- Defined by a continuous interval $[a, b]$, could be infinite

- E.g. $\int_{-\infty}^{\infty} f_X(x) dx = 1$



- Nonnegativity: $f_X(x) \geq 0, \forall x$
- Cumulative distribution function (cdf): $F_X(x)$

$$F(x) = P(X \leq x) = \int_{-\infty}^x f(u) du$$

Continuous random variable

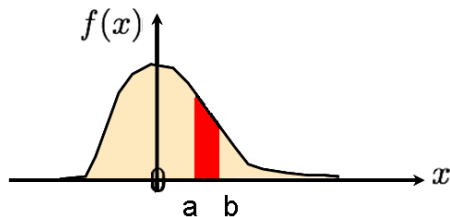
- Expectation:

$$E[X] = \int_{-\infty}^{\infty} xf(x)dx$$

- Variance:

$$\sigma_X^2 = Var(x) = \int_{-\infty}^{\infty} (xf(x) - E[X]^2)dx = E[X^2] - (E[X])^2$$

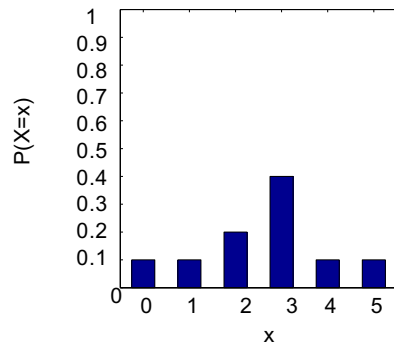
- $P(a \leq X \leq b) = \int_a^b f(x)dx$ is equal to the area:



Discrete vs. continuous distributions

Probability mass function:

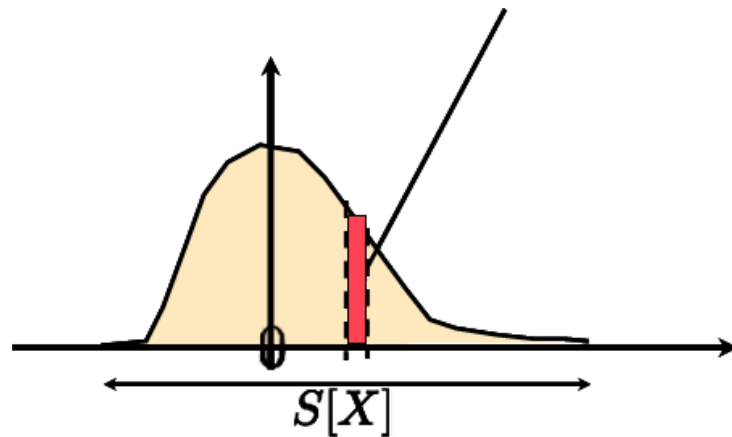
$$P(X = k) = f_k, \quad k = 0, \dots, K$$



Example: uniform distribution

Probability density function:

$$P(x \leq X \leq x + dx) = f(x)dx$$



Support: locations where $f(x)$ is nonzero