
ECE 307 – Techniques for Engineering Decisions

12. Probability Distributions

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SCOPE OF LECTURE

- ❑ We review basic probability distributions
- ❑ The entire lecture is simply a review of known probability material given the *course prerequisites*
- ❑ We extensively rely on examples to drive home the usefulness of the material
- ❑ We rely on the use of *probability distribution tables*, as found in the appendices of the *Clemen* book

OUTLINE OF DISTRIBUTION REVIEWED

Discrete distributions

- binomial

- Poisson

Continuous distributions

- exponential

- normal

THE BINOMIAL DISTRIBUTION

- **Binomial distributions are used to describe events with only two possible outcomes**
- **Basic requirements are**
 - ***dichotomous outcomes*: uncertain events occur in a sequence with each event having one of two possible outcomes such as:**

THE BINOMIAL DISTRIBUTION

➤ **success/failure**

➤ **on/off**

➤ **correct/incorrect**

➤ **true/false**

- *constant probability*: each event has the same probability of success
- *independence*: the outcome of each event is independent of the outcomes of any other event

BINOMIAL DISTRIBUTION EXAMPLE

- We consider a group of n identical machines with each machine having one of two states:

$$P \{ \textit{machine is on} \} = p$$

$$P \{ \textit{machine is off} \} = q = 1 - p$$

- For concreteness, let us set $n = 8$ and define for $i = 1, 2, \dots, 8$, the *r.v.s* :

BINOMIAL DISTRIBUTION EXAMPLE

$$X_{\sim i} = \begin{cases} 1 & \text{machine } i \text{ is on with prob. } p \\ 0 & \text{machine } i \text{ is off with prob. } q = 1 - p \end{cases}$$

- The probability that 3 or more machines are on is determined by the evaluation of the probability

$$P \left\{ \sum_{i=1}^8 X_{\sim i} \geq 3 \right\} = P \{ 3 \text{ or more machines are on} \}$$

BINOMIAL DISTRIBUTION EXAMPLE

$$= P \{ 3 \text{ machines are on} \} +$$

$$P \{ 4 \text{ machines are on} \} +$$

$$\dots +$$

$$P \{ 8 \text{ machines are on} \}$$

$$P \left\{ \sum_{i=1}^8 X_{\tilde{i}} \geq 3 \right\} = \sum_{r=3}^8 \frac{8!}{(8-r)!r!} p^r (1-p)^{8-r}$$

THE BINOMIAL DISTRIBUTION

- In general, for a r.v. \underline{R} with *dichotomous outcomes* of success and failure, the probability of r successes in n trials is

$$P \left\{ \underline{R} = r \text{ in } n \text{ trials with probability of success } p \right\}$$

$$= \frac{n!}{\binom{n-r}{r}!r!} p^r (1-p)^{n-r}$$

*the binomial
distribution*

THE BINOMIAL DISTRIBUTION

□ We can show that:

$$E\{\underset{\sim}{R}\} = n p$$

$$\text{var}\{\underset{\sim}{R}\} = n p(1 - p)$$

$$P\left\{\sum_{i=1}^n \underset{\sim}{X}_i \geq k\right\} = \sum_{r=k}^n \frac{n!}{(n-r)! r!} p^r (1-p)^{n-r}$$

EXAMPLE: SOFT PRETZELS

- A pretzel entrepreneur can sell each pretzel at \$ 0.50 with a market potential of 100,000 pretzels within a year; as there exists a competing product, he is not be the only seller
- Basic model is binomial:
 - new pretzel is a hit \Leftrightarrow captures 30 % of market in one year
(*success*)
 - new pretzel is a flop \Leftrightarrow captures 10 % of market in one year
(*failure*)

EXAMPLE: SOFT PRETZELS

- The probability of these two outcomes is equal
- Market tests are conducted with 20 pretzels taste tested against the competition; the result indicates that 5 out of 20 testers prefer the new pretzel
- We evaluate the conditional probability

$$P\left\{ \textit{new pretzel is a hit} \mid \textit{5 out of 20 people prefer new pretzel} \right\}$$

EXAMPLE: SOFT PRETZELS

- We define the success *r.v.*

$$\tilde{S} = \begin{cases} 1 & \text{new pretzel is a hit (success)} \\ 0 & \text{otherwise (failure)} \end{cases}$$

with

$$P\{\tilde{S} = 1\} = P\{\tilde{S} = 0\} = 0.5$$

and

$$\tilde{X}_i = \begin{cases} 1 & \text{person } i \text{ prefers new pretzel} \\ 0 & \text{otherwise} \end{cases}$$

- We evaluate

$$P\left\{\text{new pretzel is a hit} \mid 5 \text{ out of } 20 \text{ people prefer new pretzel}\right\}$$

EXAMPLE: SOFT PRETZELS

$$P\left\{\tilde{S} = 1 \mid \sum_{i=1}^{20} \tilde{X}_i = 5\right\} = \frac{P\left\{\tilde{S} = 1, \sum_{i=1}^{20} \tilde{X}_i = 5\right\}}{P\left\{\sum_{i=1}^{20} \tilde{X}_i = 5\right\}} =$$
$$\frac{P\left\{\sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S} = 1\right\} P\{\tilde{S} = 1\}}{P\left\{\sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S} = 1\right\} P\{\tilde{S} = 1\} + P\left\{\sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S} = 0\right\} P\{\tilde{S} = 0\}}$$

EXAMPLE: SOFT PRETZELS

$P \left\{ \sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S} = 1 \right\}$
0.178 from the binomial table

is the binomial probability
that 5 out of 20 people prefer
the new pretzel with $p = 0.3$

$P \left\{ \sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S} = 0 \right\}$
0.032 from the binomial table

is the binomial probability
that 5 out of 20 people prefer
the new pretzel with $p = 0.1$

EXAMPLE: SOFT PRETZELS

□ Therefore,

$$P\left\{\sum_{i=1}^{20} X_{\tilde{i}} = 5 \mid \mathcal{S} = 1\right\} P\{\mathcal{S} = 1\}$$

$$P\left\{\sum_{i=1}^{20} X_{\tilde{i}} = 5 \mid \mathcal{S} = 1\right\} P\{\mathcal{S} = 1\} + P\left\{\sum_{i=1}^{20} X_{\tilde{i}} = 5 \mid \mathcal{S} = 0\right\} P\{\mathcal{S} = 0\}$$

$$= \frac{(0.178)(0.5)}{(0.178)(0.5) + (0.032)(0.5)}$$

$$= 0.848$$

THE POISSON DISTRIBUTION

- ❑ The **binomial distribution** is appropriate for the representation of successes in repeated trials
- ❑ The **Poisson distribution** is appropriate for the representation of specific events over time, space, or some other problem-specific dimension, e.g., the number of customers who are served by a butcher in a meat market, or the number of chips judged unacceptable in a production run

REQUIREMENTS FOR A POISSON DISTRIBUTION

- Events can happen at any of a large number of values within the range of measurement (time, space, etc.) and possibly along a continuum
- At a specific point z , P {an event at z } is very small and therefore events do not happen *too frequently*

REQUIREMENTS FOR A POISSON DISTRIBUTION

- Each event is independent of any other event and

$$P \{ \textit{event at any point} \}$$

is constant and *independent* of all other events

- In fact, the average number of events over a unit

of measure is constant

THE POISSON DISTRIBUTED *r.v.*

- \tilde{X} is the *r.v.* representing the number of events in a unit of measure

$$P\{\tilde{X} = k\} = \frac{e^{-m} m^k}{k!}$$

$$E\{\tilde{X}\} = m \quad \text{var}\{\tilde{X}\} = m$$

*m is the Poisson
distribution parameter*

- Interpretation: the Poisson distribution parameter is the mean or the variance of the distribution

EXAMPLE: POISSON DISTRIBUTION

- Consider an assembly line for manufacturing a particular product
 - 1,024 units are produced
 - based on past experience, a flawed unit is produced every 197 units and so, *on average*, there are $\frac{1,024}{197} \gg 5.2$ flawed units in the 1,024 units of the product produced

EXAMPLE: POISSON DISTRIBUTION

- Note that the **Poisson conditions are satisfied**
 - the sample has 1,024 units
 - there are only a few flawed units in the 1,024 sample, i.e., the event of the a flawed unit is infrequent
 - the probability of a flawed unit is rather small
 - each flawed unit is *independent* of every other flawed unit

EXAMPLE: POISSON DISTRIBUTION

- Poisson distribution is appropriate representation

with $m = 5.2$ and so,

$$P\{X = k\} = \frac{e^{-5.2} (5.2)^k}{k!}$$

- If we want to determine $P\{4 \text{ or more flawed units}\}$,

we compute

EXAMPLE: POISSON DISTRIBUTION

$$P\{\tilde{X} > 4\} = 1 - P\{\tilde{X} \leq 4\} = 1 - 0.406 = 0.594$$

lookup Poisson table for $k = 4, m = 5.2$

□ The Poisson table states that for $k = 12, m = 5.2$

$$P\{\tilde{X} \leq 12\} = 0.997$$

and therefore

$$P\{\tilde{X} > 12\} = 1 - P\{\tilde{X} \leq 12\} = 0.003$$

EXAMPLE: SOFT PRETZELS

- ❑ The pretzel enterprise is going well: several retail outlets and a street vendor are selling the pretzels
- ❑ A vendor in a new location can sell, on average, 20 pretzels per hour; the vendor in an existing location sells 8 pretzels per hour

EXAMPLE: SOFT PRETZELS

- A decision is made to try to set up a second street vendor at a different, new location
- New location is classified along three distinct categories with the given probabilities

<i>category</i>	<i>characterization</i>	<i>probability</i>
<i>“good”</i>	<i>20 p/h are sold</i>	<i>0.7</i>
<i>“bad”</i>	<i>10 p/h are sold</i>	<i>0.2</i>
<i>“dismal”</i>	<i>6 p/h are sold</i>	<i>0.1</i>

EXAMPLE: SOFT PRETZELS

- After the first week, a long enough period to make a mark, a 30-minute test is run and 7 pretzels are sold during the 30-minute test period
- We analyze the situation and define the *r.v.*

$$\tilde{L} = \begin{cases} \text{"good"} & 10 & p. \text{ sold during test period} \\ \text{"bad"} & 5 & p. \text{ sold during test period} \\ \text{"dismal"} & 3 & p. \text{ sold during test period} \end{cases}$$

and assume Poisson distribution applies

EXAMPLE: SOFT PRETZELS

- We determine the conditional probabilities of the new location conditioned on the 30-minute test outcomes and evaluate

$$P\{\underline{L} = \text{"good"} \mid \underline{X} = 7\}, P\{\underline{L} = \text{"bad"} \mid \underline{X} = 7\} \text{ and}$$
$$P\{\underline{L} = \text{"dismal"} \mid \underline{X} = 7\}$$

- We compute the values of the Poisson distributed

$$P\{\underline{X} = 7 \mid \underline{L} = \text{"good"}\} = \frac{e^{-10} (10)^7}{7!} = 0.09$$

EXAMPLE: SOFT PRETZELS

$$P\{X = 7 \mid L = \text{"bad"}\} = \frac{e^{-5} (5)^7}{7!} = 0.104$$

$$P\{X = 7 \mid L = \text{"dismal"}\} = \frac{e^{-3} (3)^7}{7!} = 0.022$$

□ Then $P\{L = \text{"good"} \mid X = 7\} =$

$$P\{X = 7 \mid L = \text{"good"}\} \cdot P\{L = \text{"good"}\}$$

$$\begin{aligned} & P\{X = 7 \mid L = \text{"good"}\} \cdot P\{L = \text{"good"}\} + P\{X = 7 \mid L = \text{"bad"}\} \\ & \cdot P\{L = \text{"bad"}\} + P\{X = 7 \mid L = \text{"dismal"}\} \cdot P\{L = \text{"dismal"}\} \end{aligned}$$

EXAMPLE: SOFT PRETZELS

$$P\{\underline{L} = \text{"good"} \mid \underline{X} = 7\} = \frac{(0.09)(0.7)}{(0.09)(0.7) + (0.104)(0.2) + (0.022)(0.1)}$$
$$= 0.733$$

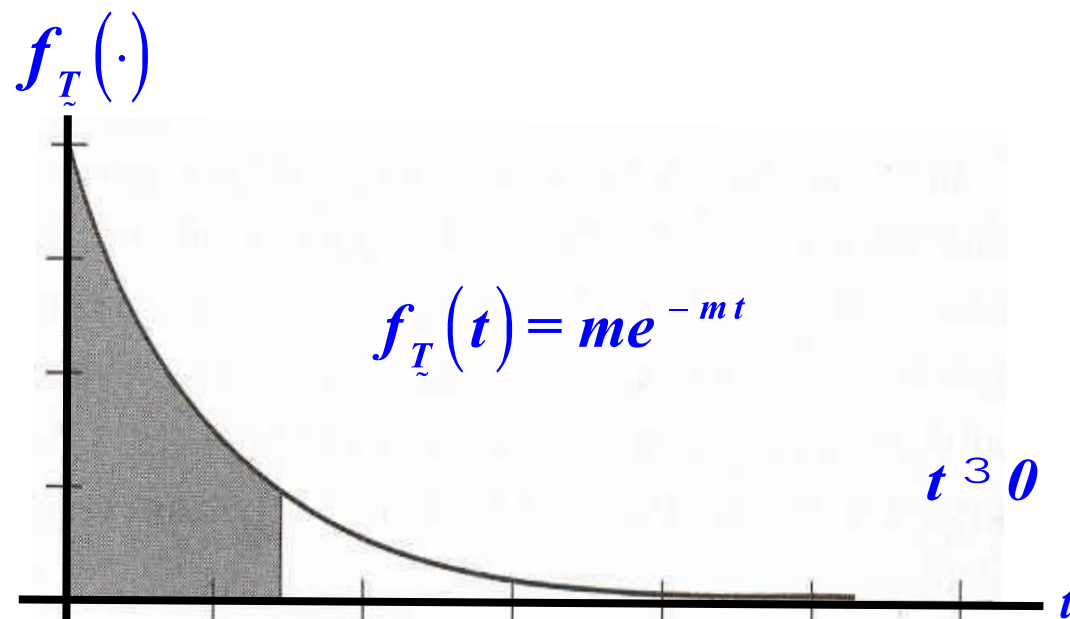
□ Similarly,

$$P\{\underline{L} = \text{"bad"} \mid \underline{X} = 7\} = 0.242$$

$$P\{\underline{L} = \text{"dismal"} \mid \underline{X} = 7\} = 1 - (0.733 + 0.242) = 0.025$$

EXPONENTIALLY DISTRIBUTED *r.v.*

- Unlike the discrete Poisson or the binomial distributed *r.v.s*, the exponentially distributed *r.v.* is **continuous**
- The density function has the form



EXPONENTIALLY DISTRIBUTED *r.v.*

- The exponentially distributed *r.v.* is related to the Poisson distribution in the following manner
- Consider the Poisson distributed *r.v.* \underline{X} , where \underline{X} represents the number of events in a specified quantity of measure, *e.g.*, period of time
- We define \underline{T} to be the *r.v.* for the uncertain quantity we measure, *e.g.*, the time between 2 sequential events or the distance between 2 accidents

EXPONENTIALLY DISTRIBUTED *r.v.*

□ Then, \tilde{T} has the exponential distribution with

$$F_{\tilde{T}}(t) = P\{\tilde{T} \leq t\} = 1 - e^{-mt},$$

$$E\{\tilde{T}\} = \frac{1}{m} \quad \text{and} \quad \text{var}\{\tilde{T}\} = \frac{1}{m^2}$$

□ The exponentially distributed *r.v.* is **completely**

specified by the parameter m

EXAMPLE: SOFT PRETZELS

- We know that it takes 3.5 *minutes* to bake a pretzel and we wish to determine the probability that the next customer will arrive after the pretzel baking is completed, i.e., $P\{\underline{T} > 3.5 \text{ minutes}\}$
- We also are given that the location types are classified as being

EXAMPLE: SOFT PRETZELS

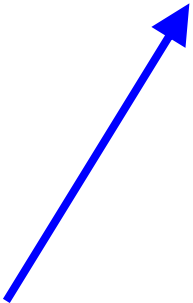
“good” location $\Leftrightarrow m = 20$ pretzels / hour

“bad” location $\Leftrightarrow m = 10$ pretzels / hour

“dismal” location $\Leftrightarrow m = 6$ pretzels / hour

- We compute the probability by conditioning on the location type to obtain

EXAMPLE: SOFT PRETZELS

$$\begin{aligned} P\{\tilde{T} > 3.5 \text{ minutes}\} &= P\{\tilde{T} > 3.5 \text{ minutes} \mid m=20\} \cdot P\{m=20\} + \\ &P\{\tilde{T} > 3.5 \text{ minutes} \mid m=10\} \cdot P\{m=10\} + \\ &P\{\tilde{T} > 3.5 \text{ minutes} \mid m=6\} \cdot P\{m=6\} \\ &\equiv 0.0583 \text{ hour} \end{aligned}$$


□ We evaluate

$$P\{\tilde{T} > 3.5m\} =$$

EXAMPLE: SOFT PRETZELS

$$e^{-0.0583(20)} P\{m = 20\} + e^{-0.0583(10)} P\{m = 10\} + e^{-0.0583(6)} P\{m = 6\}$$

*ex post
probabilities*

$$P\{m = 20\} = P\{\underline{L} = \text{"good"} \mid \underline{X} = 7\} = 0.733$$

$$P\{m = 10\} = P\{\underline{L} = \text{"bad"} \mid \underline{X} = 7\} = 0.242$$

$$P\{m = 6\} = P\{\underline{L} = \text{"dismal"} \mid \underline{X} = 7\} = 0.025$$

EXAMPLE: SOFT PRETZELS

and so

$$P\{T_{\sim} > 3.5 \text{ minutes}\} = 0.3809$$

□ Therefore,

$$P\{T_{\sim} \leq 3.5 \text{ minutes}\} = 1 - 0.3809 = 0.6191$$

and the interpretation is that the majority of the customers arrives before the pretzels are baked

THE NORMAL DISTRIBUTION

□ The *normal* or *Gaussian* distribution is, by far, the most important probability distribution since the *Law of Large Numbers* implies that the distribution of many uncertain variables is governed by the *normal* distribution, or more commonly known as the *bell curve*

□ We consider a normally distributed *r.v.* \underline{Y}

$$\underline{Y} \sim \mathcal{N}(\mu, \sigma)$$

THE NORMAL DISTRIBUTION

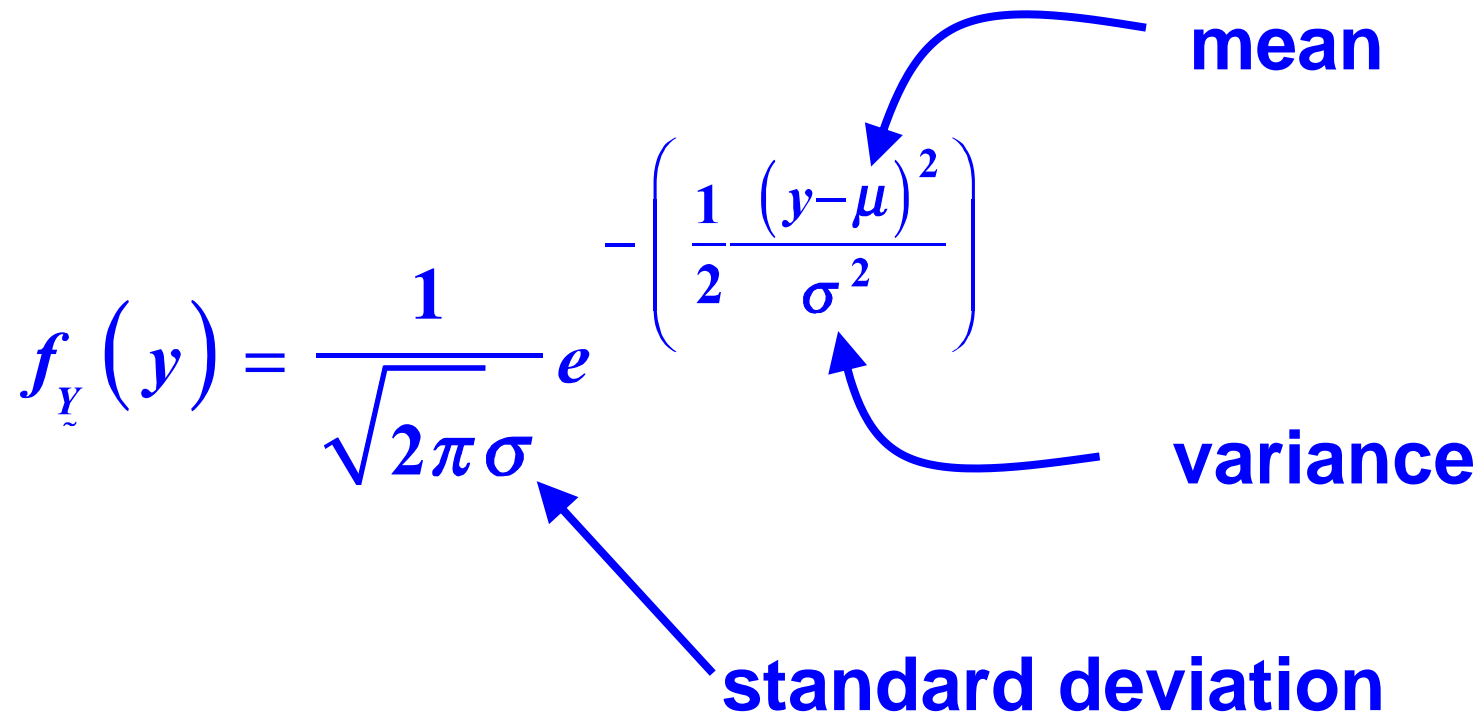
□ The density function is

$$f_{\tilde{Y}}(y) = \frac{1}{\sqrt{2\pi\sigma}} e^{-\left(\frac{1}{2} \frac{(y-\mu)^2}{\sigma^2}\right)}$$

mean

variance

standard deviation

The diagram shows the normal distribution density function formula. Three blue arrows point from text labels to parts of the formula: 'mean' points to the $(y-\mu)$ term in the exponent; 'variance' points to the σ^2 term in the denominator of the exponent; and 'standard deviation' points to the σ term in the denominator of the leading fraction.

with $E\{\tilde{Y}\} = \mu$ and $var\{\tilde{Y}\} = \sigma^2$

THE STANDARD NORMAL DISTRIBUTION

- Consider the *r.v.* \underline{Z} which has the standard normal distribution

$$\underline{Z} \sim \mathcal{N}(0,1)$$

1. The relationship between the *r.v.s* \underline{Y} and \underline{Z} is given by the affine expression:

$$\underline{Z} = \frac{\underline{Y} - \mu}{\sigma}$$

with

$$P\{\underline{Y} \leq a\} = P\{\underline{Z} \leq (a - \mu) / \sigma\}$$

THE STANDARD NORMAL DISTRIBUTION

□ Note that

$$E \left\{ \underset{\sim}{Z} \right\} = 0 \quad \text{and} \quad \text{var} \left\{ \underset{\sim}{Z} \right\} = 1$$

□ In general, any value of the normal distribution is obtained from the *standard normal distribution* with the affine transformation

$$\underset{\sim}{Z} = \frac{\underset{\sim}{Y} - \mu}{\sigma}$$

EXAMPLE: QUALITY CONTROL

- We consider a disk drive manufacturing process in which a particular machine produces a part used in the final assembly; the part must rigorously meet the width requirements to be within the interval $[3.995, 4.005]$ *mm* ; else, the company incurs \$10.40 in repair costs
- The machine is set to produce parts with the width of 4mm , but in actual conditions, the width is a normally distributed *r.v.* \tilde{W} with

EXAMPLE: QUALITY CONTROL

$$\tilde{W} \sim \mathcal{N}(4, \sigma)$$

and

$$\sigma = f(\text{speed of machine}) = \begin{cases} 0.0019 & \text{slow speed} \\ 0.0026 & \text{high speed} \end{cases}$$

□ The respective costs in \$ of the disk drive are

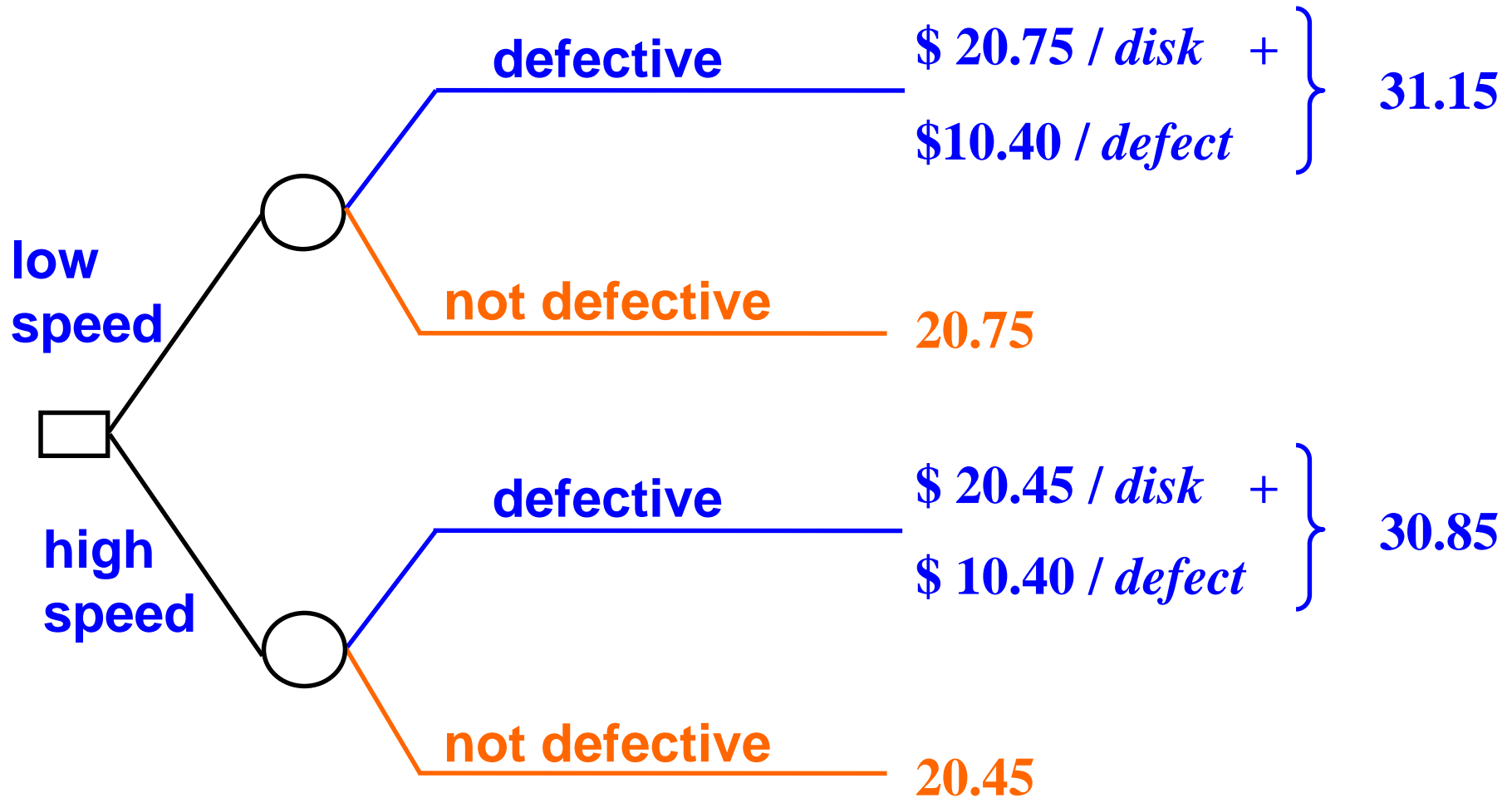
20.75 *slow speed*

20.45 *high speed*

EXAMPLE: QUALITY CONTROL

- ❑ We may interpret the cost data to imply that more disks can be produced at lower costs at the high speed
- ❑ The problem is to **select** the machine speed to obtain the more cost effective result
- ❑ A decision tree is useful in the analysis of the situation

EXAMPLE: QUALITY CONTROL



□ We compute the probability of each outcome


LOW – SPEED PROBABILITY EVALUATION

$$P\{\textit{defective disk is produced}\} =$$

$$P\{\tilde{W} < 3.995 \textit{ or } \tilde{W} > 4.005\} =$$

$$1 - P\{3.995 \leq \tilde{W} \leq 4.005\} =$$

$$\tilde{Z} = \frac{\tilde{W} - 4}{0.0019}$$


$$1 - P\left\{\frac{3.995 - 4}{0.0019} \leq \tilde{Z} \leq \frac{4.005 - 4}{0.0019}\right\} =$$

LOW – SPEED PROBABILITY EVALUATION

$$1 - P\{-2.63 \leq \tilde{Z} \leq 2.63\} =$$

$$1 - \left[P\{\tilde{Z} \leq 2.63\} - P\{\tilde{Z} \leq -2.63\} \right] = 0.0086$$

0.9957

0.0043

0.9914

HIGH – SPEED PROBABILITY EVALUATION

$$P\{\textit{defective disk is produced}\} =$$

$$P\{\tilde{W} < 3.995 \textit{ or } \tilde{W} > 4.005\} =$$

$$1 - P\{3.995 \leq \tilde{W} \leq 4.005\} =$$

$$\tilde{Z} = \frac{\tilde{W} - 4}{0.0026}$$

$$1 - P\left\{\frac{3.995 - 4}{0.0026} \leq \tilde{Z} \leq \frac{4.005 - 4}{0.0026}\right\} =$$

HIGH – SPEED PROBABILITY EVALUATION

$$1 - P\{-1.92 \leq \tilde{Z} \leq 1.92\} =$$

$$1 - \left[P\{\tilde{Z} \leq 1.92\} - P\{\tilde{Z} \leq -1.92\} \right] = 0.0548$$

Diagram illustrating the calculation of the probability $1 - P\{-1.92 \leq \tilde{Z} \leq 1.92\}$. The expression is shown as $1 - [P\{\tilde{Z} \leq 1.92\} - P\{\tilde{Z} \leq -1.92\}] = 0.0548$. The two probabilities are circled, and arrows point from the values 0.9726 and 0.0274 to them. A bracket below the terms $P\{\tilde{Z} \leq 1.92\}$ and $P\{\tilde{Z} \leq -1.92\}$ indicates their difference is 0.9452.

MEAN VALUE EVALUATION

- We next evaluate the mean cost per disk

$$E\{cost / disk | low\ speed\} = (0.9914)(20.75) + (0.0086)(31.15)$$

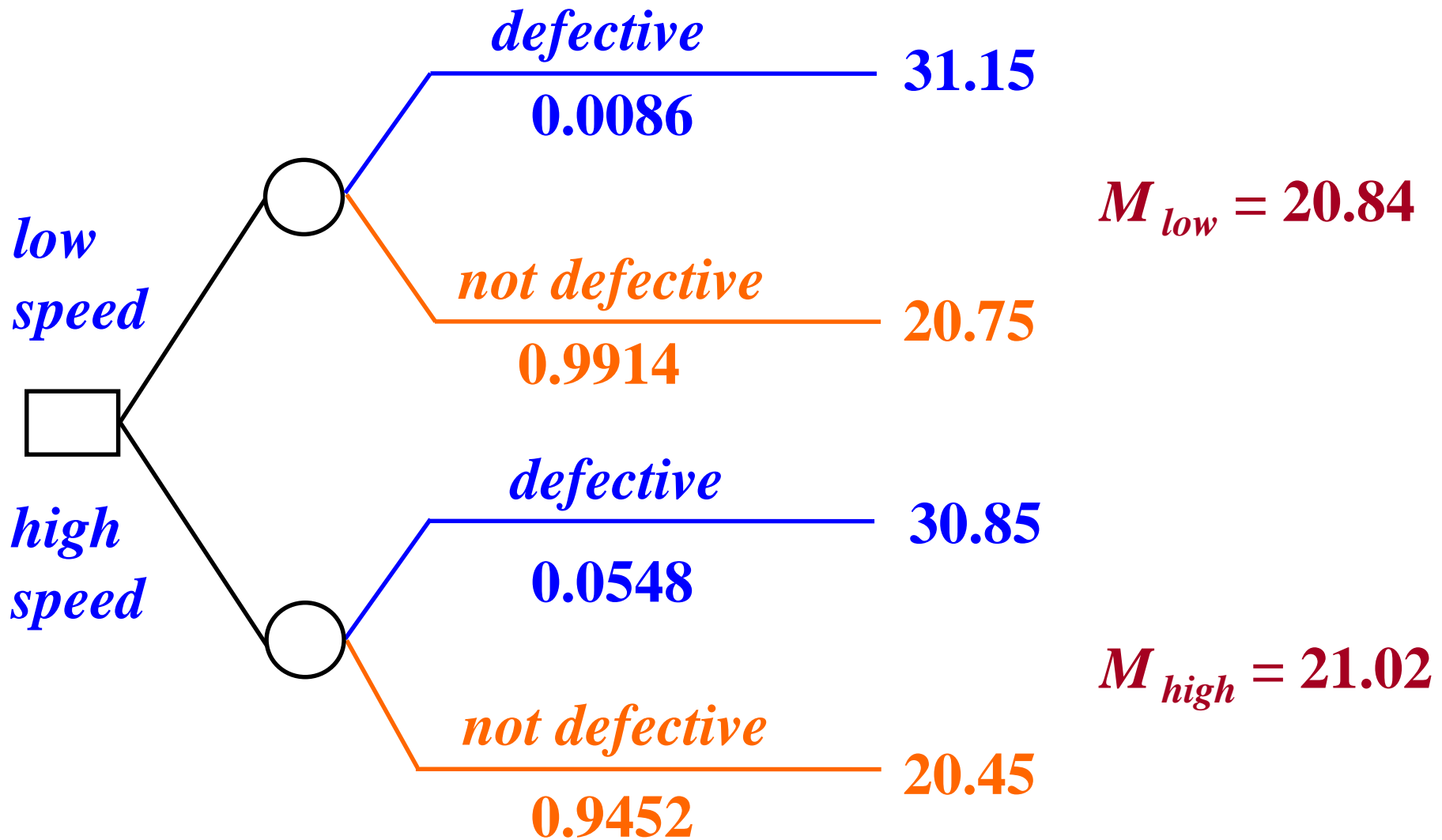
$$= 20.84$$

$$E\{cost / disk | high\ speed\} = (0.9452)(20.45) + (0.0548)(30.85)$$

$$= 21.02$$

- We summarize the information in the decision tree

EXAMPLE: QUALITY CONTROL



CASE STUDY: OVERBOOKING

- ❑ *M Airlines* has a commuter plane capable of flying 16 passengers
- ❑ The plane is used on a route for which *M Airlines* charges \$ 225
- ❑ The airliner's cost structure is based on

the fixed costs for each flight	\$ 900
the variable costs/passenger	\$ 100
the “no-show” rate	4 %

CASE STUDY – OVERBOOKING

- ❑ The refund policy specifies that unused tickets are refunded only if a reservation is cancelled 24 *h* ahead of the scheduled departure
- ❑ The overbooking policy pays \$ 100 as an incentive to each bumped passenger and refunds the ticket
- ❑ The decision required is to determine how many reservations should the airliner sell on this plane

SAMPLE CALCULATION FOR THE CASE 18 RESERVATIONS ARE SOLD

total revenues : $\tilde{R} = 225 \cdot 18 = 4,050$

passenger fixed and variable costs :

$$C_1 = 900 + 100 \cdot \min\{\text{number of "shows", } 16\} \$$$

bumping costs :

$$C_2 = (225 + 100) \cdot \max\{0, \text{number of "shows"} - 16\} \$$$

refunds to customers

$$\text{total costs : } \tilde{C} = C_1 + C_2$$

CASE STUDY: OVERBOOKING

□ We evaluate

$$P \left\{ \text{no. of "shows"} > 16 \mid \text{reservations sold} = 18 \right\}$$

□ We assume that each reservation is a r.v. $P_{\sim i}$:

$$P_{\sim i} = \begin{cases} 1 & \text{passenger } i \text{ is a "show" with prob. } 0.96 \\ 0 & \text{passenger } i \text{ is a "no show" with prob. } 0.04 \end{cases}$$

CASE STUDY: OVERBOOKING

- Given reservations sold = 18 , then we need to evaluate

$$P \left\{ \sum_{i=1}^{18} P_{\sim i} > 16 \mid 18 \text{ reservations} \right\}$$

- We first evaluate

$$P \left\{ \sum_{i=1}^{17} P_{\sim i} > 16 \mid 17 \text{ res} \right\} = P \left\{ \sum_{i=1}^{17} P_{\sim i} \geq 17 \mid 17 \text{ res} \right\} = P \left\{ \sum_{i=1}^{17} P_{\sim i} = 17 \mid 17 \text{ res} \right\}$$

binomial *r.v.* with $p = 0.96$

CASE STUDY: OVERBOOKING

$$= (.96)^{17} (.04)^0 \longleftarrow 0.4996$$

□ Then,

$$P \left\{ \sum_{i=1}^{18} P_{\sim i} > 16 \mid 18 \text{ res} \right\} = P \left\{ \sum_{i=1}^{18} P_{\sim i} \geq 17 \mid 18 \text{ res} \right\} =$$

$$P \left\{ \sum_{i=1}^{18} P_{\sim i} = 17 \mid 18 \text{ res} \right\} + P \left\{ \sum_{i=1}^{18} P_{\sim i} = 18 \mid 18 \text{ res} \right\} = 0.8359$$
$$\underbrace{\hspace{10em}}_{18 (.4996) (.04)} \quad + \quad \underbrace{\hspace{10em}}_{(.4996) (.96)}$$

CASE STUDY: OVERBOOKING

- Given reservations sold = 19, then we compute and show that

$$P \left\{ \sum_{i=1}^{19} P_{\sim i} > 16 \mid 19 \text{ res} \right\} = .9616$$

- We next consider the profits *r.v.* π , where,

$$\pi = R - C = R - (C_1 + C_2)$$

and evaluate $E\{\pi\}$ for different values of reservations sold

CASE STUDY: OVERBOOKING

□ For reservations = 16

$$E\{R_{\sim}\} = (16)(225) = 3,600$$

$$E\{C_1\} = 900 + 100 \sum_{n=0}^{16} nP \left\{ \sum_{i=1}^{16} P_{\sim i} = n \right\}$$

$$= 900 + 100 E \left\{ \sum_{i=1}^{16} P_{\sim i} \right\}$$

$(16)(.96) = 15.36$

$$= 900 + 1,536$$

binomial
distribution



CASE STUDY: OVERBOOKING

$$= 2,436;$$

also,

$$E\{C_2\} = (225 + 100) \max\left\{0, \sum_{i=1}^{16} P_{\tilde{i}} - 16\right\} = 0$$

and so

$$\begin{aligned} E\{\pi_{\tilde{}}|16 \text{ res}\} &= E\{R_{\tilde{}}\} - E\{C_{\tilde{}}\} \\ &= E\{R_{\tilde{}}\} - E\{C_1 + C_2\} \\ &= 3,600 - 2,436 \\ &= 1,164 \end{aligned}$$

CASE STUDY: OVERBOOKING

□ For reservations = 17

$$E\{\tilde{R}\} = (17)(225) = 3,825$$

$$E\{C_1\} = 900 + 100 \sum_{n=0}^{16} n P \left\{ \sum_{i=1}^{17} \tilde{P}_i = n \right\} + 100.16 \cdot P \left\{ \sum_{i=1}^{17} \tilde{P}_i = 17 \right\}$$

$$= 900 + 782.70 + 799.34$$

$$= 2,482.04$$

CASE STUDY: OVERBOOKING

also,

$$E\{C_2\} = 325P \left\{ \sum_{i=1}^{17} P_{\sim i} = 17 \right\}$$

$$= 325(0.4996)$$

$$= 162.37$$

and so

$$E\{\pi_{\sim} | 17 \text{ res}\} = 3,825 - 2,482.04 - 162.37$$

$$= 1,180.59 > 1,164$$

CASE STUDY: OVERBOOKING

□ For reservations = 18

$$E\{R\} = (18)(225) = 4,050$$

$$E\{C_1\} = 900 + 100 \sum_{n=0}^{16} nP \left\{ \sum_{i=1}^{18} P_{\sim i} = n \right\} + 1,600 \cdot P \left\{ \sum_{i=1}^{18} P_{\sim i} > 16 \right\}$$

$$= 900 + 253.22 + 1,342.89$$

$$= 2,496.11$$

CASE STUDY: OVERBOOKING

$$E\{C_2\} = 325 P \left\{ \underbrace{\sum_{i=1}^{18} P_{\sim i} = 17}_{.3597} \right\} + 650 P \left\{ \underbrace{\sum_{i=1}^{18} P_{\sim i} = 18}_{.4796} \right\} = 428.65$$

and

$$E\{\pi_{\sim} | 18 \text{ res}\} = 4,050 - 2,496.11 - 428.65$$

$$= 1,125.24$$

$$< 1,180.59$$

CASE STUDY: OVERBOOKING

□ We can show that for reservations = 19

$$E\left\{\pi_{\sim} \mid 19 \text{ res}\right\} < 1180.59$$

□ We conclude that the profits are maximized for

reservations = 17 and so any overbooking over

that number results in lower profits