
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 prerequisites
- ☐ We extensively rely on examples to drive home the usefulness of the material
- ☐ We rely on the use of tabulated data of probability distributions

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

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

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

$$\dots +$$

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

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

THE BINOMIAL DISTRIBUTION

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

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

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

*the binomial
distribution*

THE BINOMIAL DISTRIBUTION

□ We can show that:

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

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

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

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 cannot 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

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

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

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

EXAMPLE: SOFT PRETZELS

$$P\left\{\mathcal{S} = 1 \mid \sum_{i=1}^{20} X_i = 5\right\} = \frac{P\left\{\mathcal{S} = 1, \sum_{i=1}^{20} X_i = 5\right\}}{P\left\{\sum_{i=1}^{20} X_i = 5\right\}} =$$

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

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

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,

$$\begin{aligned} & P\left\{\sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S}=1\right\} P\{\tilde{S}=1\} \\ & \hline P\left\{\sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S}=1\right\} P\{\tilde{S}=1\} + P\left\{\sum_{i=1}^{20} \tilde{X}_i = 5 \mid \tilde{S}=0\right\} P\{\tilde{S}=0\} \\ & = \frac{(0.178)(0.5)}{(0.178)(0.5) + (0.032)(0.5)} \\ & = 0.848 \end{aligned}$$

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 , $P \{ \text{an event at } z \}$ 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

$$\left. \begin{aligned} P\{\tilde{X} = k\} &= \frac{e^{-m} m^k}{k!} \\ E\{\tilde{X}\} &= m \quad \text{var}\{\tilde{X}\} = m \end{aligned} \right\} \begin{array}{l} m \text{ is the Poisson} \\ \text{distribution parameter} \end{array}$$

- 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
 - from past experience, a flawed unit is manufactured every 197 units and so, *on average*, there are $\frac{1,024}{197} \approx 5.2$ flawed units in the 1,024 products that are are 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 occurrence of a flawed unit is infrequent
 - the probability of a flawed unit is rather small
 - each flawed unit is *independent* of every other flawed unit

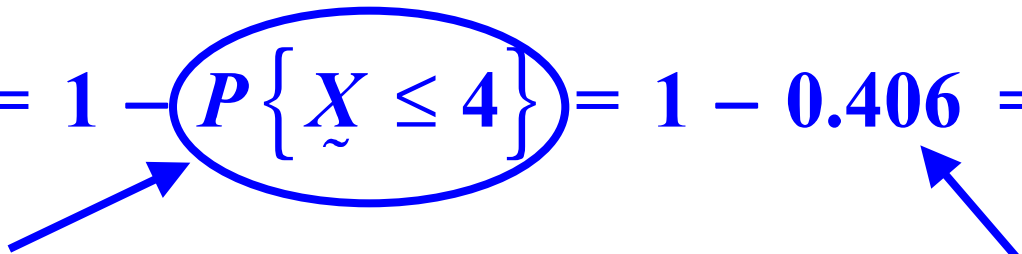
EXAMPLE: POISSON DISTRIBUTION

- 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 the probability of 4 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>	0.7
<i>“bad”</i>	<i>10 p/h are sold</i>	0.2
<i>“dismal”</i>	<i>6 p/h are sold</i>	0.1

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 by defining the *r.v.*

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

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

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

- We compute the values of the Poisson distributed

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

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

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

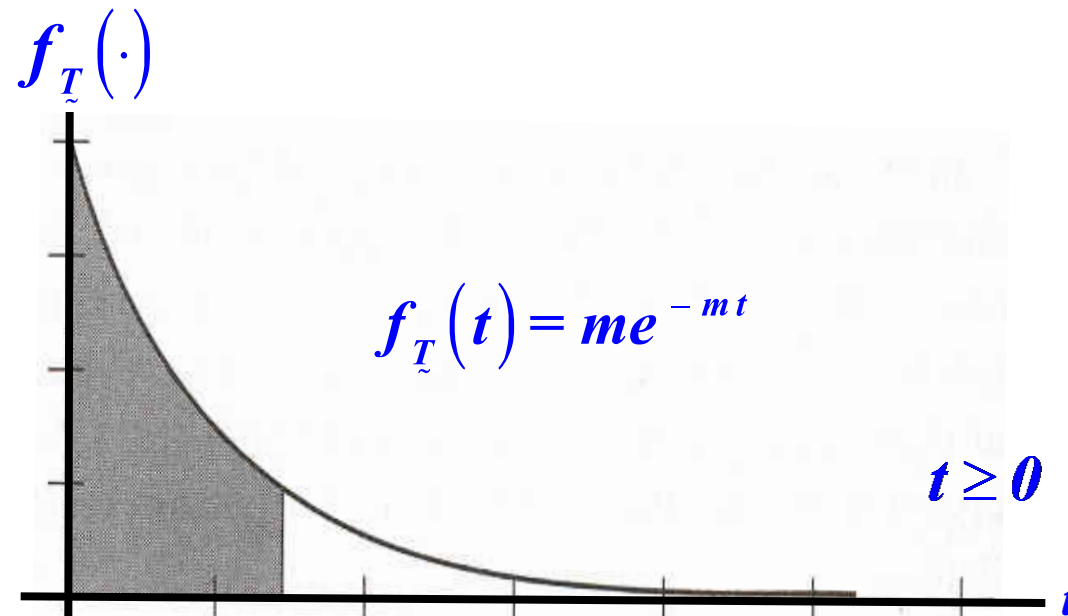
□ Similarly,

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

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

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
- ❑ Consider the Poisson distributed $r.v.$ \tilde{X} with \tilde{X} representing the number of events in a given quantity of measure, *e.g.*, period of time
- ❑ We define \tilde{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

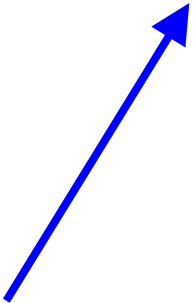
“good” location $\Leftrightarrow m = 20$ pretzels / hour

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

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

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

EXAMPLE: SOFT PRETZELS

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


□ We evaluate

$$P\{T_{\sim} > 3.5m\} =$$

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*

$$\left\{ \begin{array}{l} P\{m = 20\} = P\{\tilde{L} = \text{"good"} \mid \tilde{X} = 7\} = 0.733 \\ P\{m = 10\} = P\{\tilde{L} = \text{"bad"} \mid \tilde{X} = 7\} = 0.242 \\ P\{m = 6\} = P\{\tilde{L} = \text{"dismal"} \mid \tilde{X} = 7\} = 0.025 \end{array} \right.$$

EXAMPLE: SOFT PRETZELS

and so

$$P\{\tilde{T} > 3.5 \text{ minutes}\} = 0.3809$$

□ Therefore,

$$P\{\tilde{T} \leq 3.5 \text{ minutes}\} = 1 - 0.3809 = 0.6191$$

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 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. A blue arrow points from the label 'mean' to the term $(y-\mu)$ in the exponent. Another blue arrow points from the label 'variance' to the term σ^2 in the denominator of the exponent. A third blue arrow points from the label 'standard deviation' to the term σ 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

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

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

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

with

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

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 within the interval $[3.995, 4.005] \text{ mm}$; else, the company incurs \$10.40 in repair costs
- ❑ The machine is set to produce parts with the width of 4mm , but in reality, the width is a normally distributed *r.v.* W with

EXAMPLE: QUALITY CONTROL

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

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

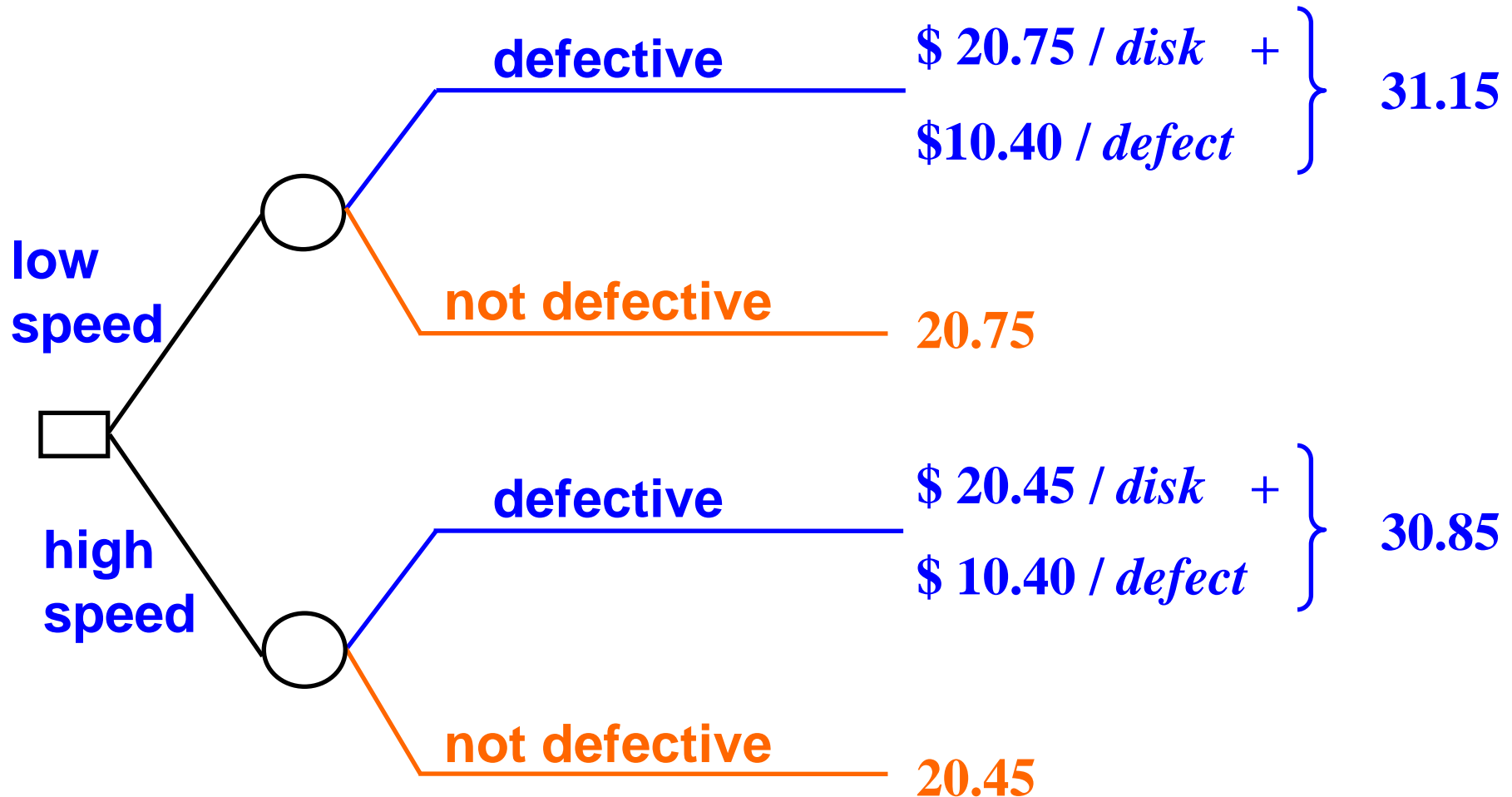
20.75 *slow speed*

20.45 *high speed*

EXAMPLE: QUALITY CONTROL

- ❑ We may interpret the cost data to imply that more disks can be produced at lesser 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 evaluate the probability of each outcome


LOW – SPEED PROBABILITY EVALUATION

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

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

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

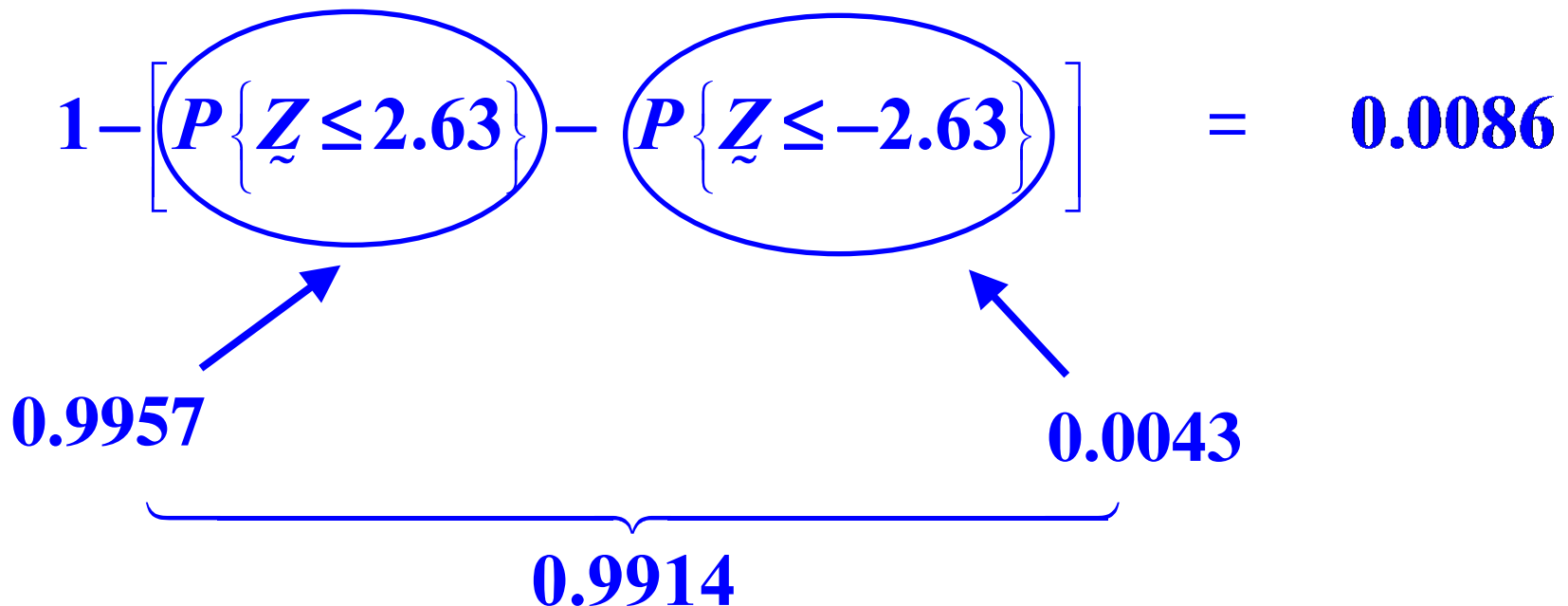
$$\tilde{Z} = \frac{\tilde{W} - 4}{0.0019}$$


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

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

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

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

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

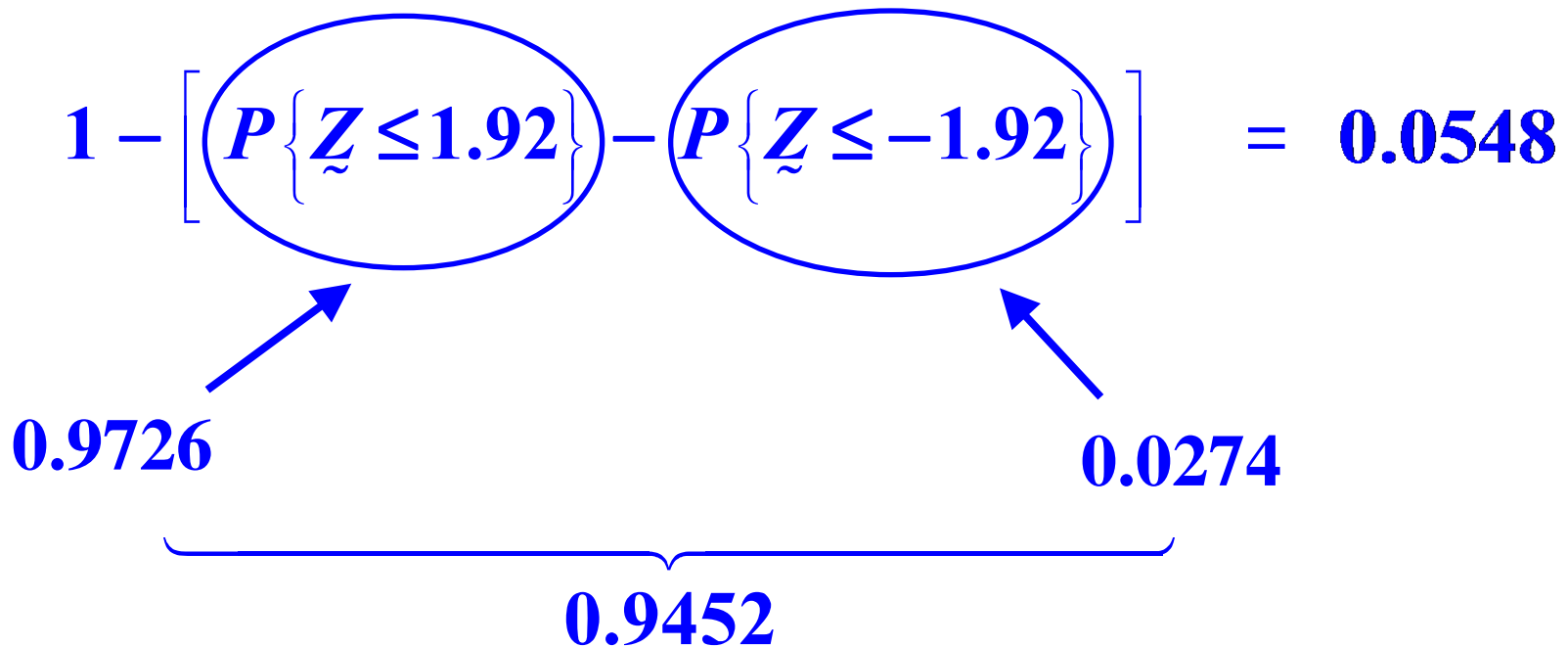
$$\tilde{Z} = \frac{\tilde{W} - 4}{0.0026}$$


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

HIGH – SPEED PROBABILITY EVALUATION

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

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


0.9726 0.0274
0.9452

MEAN VALUE EVALUATION

- We next evaluate the mean cost per disk

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

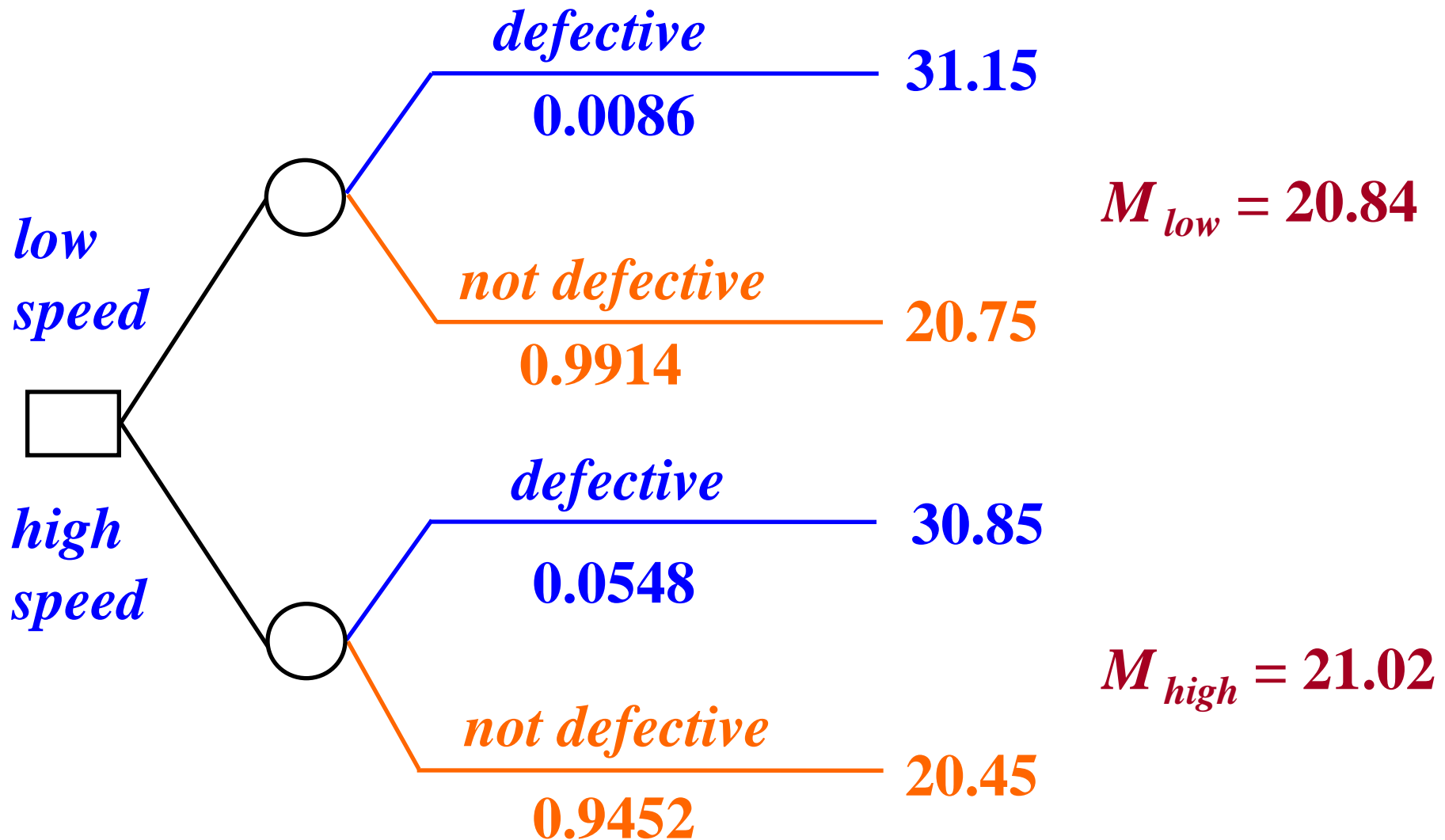
$$= 20.84$$

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

$$= 21.02$$

- 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 is that unused tickets are refunded only if a reservation is cancelled 24 *h* before 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 SELLING 18 RESERVATIONS

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

- If 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,

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

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

CASE STUDY: OVERBOOKING

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

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

- We next consider the profit *r.v.* π_{\sim} , where,

$$\pi_{\sim} = \underline{R} - \underline{C} = \underline{R} - (C_1 + C_2)$$

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

CASE STUDY: OVERBOOKING

□ For reservations = 16

$$E\{\tilde{R}\} = (16)(225) = 3,600$$

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

$$= 900 + 100 \underbrace{E\left\{ \sum_{i=1}^{16} \tilde{P}_i \right\}}_{(16)(.96) = 15.36}$$

$$= 900 + 1,536$$

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

$$E\{\pi_{\tilde{}}|16 \text{ res}\} = E\{R_{\tilde{}}\} - E\{C_{\tilde{}}\}$$

$$= E\{R_{\tilde{}}\} - E\{C_1 + C_2\}$$

$$= 3,600 - 2,436$$

$$= 1,164$$

CASE STUDY: OVERBOOKING

□ For reservations = 17

$$E\{R_{\sim}\} = (17)(225) = 3,825$$

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

$$= 900 + 782.70 + 799.34$$

$$= 2,482.04$$

CASE STUDY: OVERBOOKING

also,

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

$$= 325(0.4996)$$

$$= 162.37$$

and so

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

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

CASE STUDY: OVERBOOKING

□ For reservations = 18

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

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

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

$$= 2,496.11$$

CASE STUDY: OVERBOOKING

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

and

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

$$= 1,125.24$$

$$< 1,180.59$$

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