## 2. Fundamentals of Probability Models

2.1 Events and Probability
2.2 Elements of Set Theory
2.3 Mathematics of Probability
2.4 Concluding Summary

## Ex. 2.2

Designing a left turn
Probability of 5 or more cars waiting
No. of cars No. of observation Relative frequency
$\left.\begin{array}{rrc}0 & 4 & 4 / 60 \\ 1 & 16 & 16 / 60 \\ 2 & 20 & 20 / 60 \\ 3 & 14 & 14 / 60 \\ 4 & 3 & 3 / 60 \\ 5 & 2 & 2 / 60 \\ 6 & 1 & 1 / 60 \\ 7 & 0 & 0 \\ 8 & 0 & 0\end{array}\right\}=3 / 60$

### 2.2.1 Important Definitions

Sample Space: The set of all possibilities in a probabilistic problem. Discrete sample space or Continuous sample space Finite sample space or infinite sample space
Sample Point: Each of the individual possibilities.
Event: A subset of the sample space.

Impossible event " $\Phi$ " is the event with no sample point
Certain event " S " is the event containing all the sample points in a same sample space. The sample space itself.

Complementary event " $\bar{E}$ " contains all the sample points in S that are not in E for an event E in a sample space S .

## Combination of Events

## Venn diagram



Union " $\cup$ " The occurrence of E1 or E2 or both ("or" is used in an inclusive sense, "and/or")
Intersection " $\cap$ " The joint occurrence of E1 and E2 $\mathbf{E}_{1} \cap \mathbf{E}_{2}=\mathbf{E}_{1} \mathbf{E}_{2}$
Mutually exclusive event Disjoint of E1 and E2.

$$
\mathrm{E} 1 \mathrm{E} 2=\Phi
$$

Collectively exhaustive event
Union of all the events constitute the sample space

### 2.2.2 Mathematical Operations of Sets

Equality of sets
Two sets are equal if and only if both sets contain exactly the same sample points

$$
\begin{array}{ll}
A \bigcup \phi=A, & A \cap \phi=\phi \\
A \bigcup A=A, & A \cap A=A \\
A \bigcup S=S, & A \bigcap S=A \tag{2.1c}
\end{array}
$$

Complementary sets

$$
\begin{align*}
& E \bigcup \bar{E}=S  \tag{2.2a}\\
& E \bigcap \bar{E}=\phi \\
& (\overline{\bar{E}})=E \tag{2.2b}
\end{align*}
$$



Commutative rule

$$
\begin{aligned}
& A \bigcup B=B \cup A \\
& A B=B A
\end{aligned}
$$

Associative rule

$$
\begin{aligned}
& (A \cup B) \cup C=A \cup(B \cup C) \\
& (A B) C=A(B C)
\end{aligned}
$$

Distributive rule

$$
\begin{aligned}
& (A \cup B) C=A C \cup B C \\
& (A B) \cup C=(A \cup C)(B \cup C)
\end{aligned}
$$



## De Morgan's rule

$$
\begin{align*}
& \overline{E_{1} \bigcup E_{2}}=\overline{E_{1}} \cap \overline{E_{2}}=\overline{E_{1}} \overline{E_{2}} \\
& \overline{E_{1} \cup E_{2} \cup \cdots \bigcup E_{n}}=\overline{E_{1}} \overline{E_{2}} \cdots \overline{E_{n}}  \tag{2.3a}\\
& \overline{E_{1} \bigcup \overline{E_{2}} \bigcup \cdots \bigcup \overline{E_{n}}}=E_{1} E_{2} \cdots E_{n} \\
& \overline{E_{1} E_{2} \cdots E_{n}}=\overline{E_{1}} \bigcup \overline{E_{2}} \bigcup \cdots \cup \overline{E_{n}} \tag{2.3b}
\end{align*}
$$

## Duality relation:

"The complement of unions and intersections is equal to the intersections and unions of the respective complements."

$$
\begin{aligned}
& \overline{A \bigcup B C}=\bar{A} \cap \overline{B C}=\bar{A}(\bar{B} \bigcup \bar{C})=\bar{A} \bar{B} \bigcup \bar{A} \bar{C} \\
& \overline{(A \bigcup B) C}=\overline{(A \bigcup B) \bigcup \bar{C}}=(\bar{A} \bar{B}) \bigcup \bar{C}
\end{aligned}
$$


$\overline{E_{1} \cup E_{2}}=\overline{E_{1}} \overline{E_{2}}$

## 2. Basic Probability Concepts

2.1 Events and Probability
2.2 Elements of Set Theory
2.3 Mathematics of Probability
2.4 Concluding Summary

### 2.3.1 Addition Rule

## Axioms of Probability

1) Event $E$ in a sample space $S$

$$
\begin{equation*}
\mathrm{P}(\mathrm{E}) \geq \theta \tag{2.4}
\end{equation*}
$$

2) For the certain event $S$

$$
\begin{equation*}
P(S)=1 \tag{2.5}
\end{equation*}
$$

3) Events $E_{1}$ and $E_{2}$ are mutually exclusive

$$
\begin{equation*}
P\left(E_{1} \cup E_{2}\right)=P\left(E_{1}\right)+P\left(E_{2}\right) \tag{2.6}
\end{equation*}
$$

If $\mathrm{E}_{1}$ and $\mathrm{E}_{2}$ are not mutually exclusive

$$
\begin{equation*}
P\left(E_{1} \cup E_{2}\right)=P\left(E_{1}\right)+P\left(E_{2}\right)-P\left(E_{1} E_{2}\right) \tag{2.8}
\end{equation*}
$$

4) $P(\overline{\mathrm{E}})=1-\mathrm{P}(\mathrm{E})$

## Ex. 2.13

A: Definitely completed
B: Questionable
C: Definitely incomplete

1) Sample Space?
2) Probability of exactly one job being completed ?

What is $\mathrm{P}\left(\mathrm{E}_{1} \cup \mathrm{E}_{2}\right)$ ?
$\mathrm{E}_{1}$ : Job 1 definitely completed

$$
\mathrm{E}_{1} \supset\{\mathrm{AA}, \mathrm{AB}, \mathrm{AC}\}
$$

$\mathrm{E}_{2}$ : Job 2 definitely completed

$$
\mathrm{E}_{2} \supset\{\mathrm{AA}, \mathrm{BA}, \mathrm{CA}\}
$$

### 2.3.2 Conditional Probability

## Conditional Probability $\mathrm{P}\left(\mathrm{E}_{1} \mid \mathrm{E}_{2}\right)$ :

Probability of an event assuming another event has occurred

$$
\begin{align*}
& P\left(E_{1} \mid E_{2}\right)=\frac{P\left(E_{1} E_{2}\right)}{P\left(E_{2}\right)}  \tag{2.11}\\
& \mathrm{P}\left(\overline{\mathrm{E}_{1}} \mid \mathrm{E}_{2}\right)=1-\mathrm{P}\left(\mathrm{E}_{1} \mid \mathrm{E}_{2}\right) \tag{2.12}
\end{align*}
$$

reconstituted sample space

## Ex. 2.18

Straight ahead $=\mathrm{E}_{1}$, Turn right $=\mathrm{E}_{2}$, Turn left $=\mathrm{E}_{3}$
$\mathrm{E}_{1}, \mathrm{E}_{2}$ and $\mathrm{E}_{3}$ are mutually exclusive and collectively exhausted.
(a) $\mathrm{P}\left(\mathrm{E}_{1}\right)=2.0 \mathrm{P}\left(\mathrm{E}_{2}\right), \quad \mathrm{P}\left(\mathrm{E}_{2}\right)=2.0 \mathrm{P}\left(\mathrm{E}_{3}\right)$

$$
\begin{aligned}
\mathrm{P}\left(\mathrm{E}_{1}\right)+\left(\mathrm{E}_{2}\right)+\mathrm{P}\left(\mathrm{E}_{3}\right)=1.0 \Rightarrow & 4.0 \mathrm{P}\left(\mathrm{E}_{3}\right)+2.0 \mathrm{P}\left(\mathrm{E}_{3}\right)+\mathrm{P}\left(\mathrm{E}_{3}\right)=1.0 \\
& \mathrm{P}\left(\mathrm{E}_{3}\right)=1 / 7=0.1429
\end{aligned}
$$

(b) $P\left(E_{2} \mid E_{2} \cup E_{3}\right)=P\left[E_{2} \cap\left(E_{2} \cup E_{3}\right)\right] / P\left(E_{2} \cup E_{3}\right)$

$$
\begin{aligned}
& =P\left(\mathrm{E}_{2}\right) / \mathrm{P}\left(\mathrm{E}_{2} \cup \mathrm{E}_{3}\right)=2 / 7 /(2 / 7+1 / 7) \\
& =2 / 3=0.6667
\end{aligned}
$$

(c) $P\left(\overline{E_{2}} \mid E_{2} \cup E_{3}\right)=1-P\left(E_{2} \mid E_{2} \cup E_{3}\right)=1-2 / 3=1 / 3$

### 2.3.3 Multiplication Rule

$$
\begin{equation*}
\mathrm{P}\left(\mathrm{E}_{1} \mathrm{E}_{2}\right)=\mathrm{P}\left(\mathrm{E}_{1} \mid \mathrm{E}_{2}\right) \mathrm{P}\left(\mathrm{E}_{2}\right), \mathrm{P}\left(\mathrm{E}_{1} \mathrm{E}_{2}\right)=\mathrm{P}\left(\mathrm{E}_{2} \mid \mathrm{E}_{1}\right) \mathrm{P}\left(\mathrm{E}_{1}\right) \tag{2.14}
\end{equation*}
$$

$\mathrm{E}_{1}$ and $\mathrm{E}_{2}$ are statistically independent.

$$
\begin{align*}
& \mathrm{P}\left(\mathrm{E}_{2} \mid \mathrm{E}_{1}\right)=\mathrm{P}\left(\mathrm{E}_{2}\right), \quad \mathrm{P}\left(\mathrm{E}_{1} \mid \mathrm{E}_{2}\right)=\mathrm{P}\left(\mathrm{E}_{1}\right)  \tag{2.13}\\
& \mathrm{P}\left(\mathrm{E}_{1} \mathrm{E}_{2}\right)=\mathrm{P}\left(\mathrm{E}_{1}\right) \mathrm{P}\left(\mathrm{E}_{2}\right) \tag{2.15}
\end{align*}
$$

Three Events

$$
\begin{align*}
& \mathrm{P}\left(\mathrm{E}_{1} \mathrm{E}_{2} \mathrm{E}_{3}\right)=\mathrm{P}\left(\mathrm{E}_{1} \mathrm{E}_{2} \mid \mathrm{E}_{3}\right) \mathrm{P}\left(\mathrm{E}_{3}\right) \\
& \mathrm{P}\left(\mathrm{E}_{1} \mathrm{E}_{2} \mathrm{E}_{3}\right)=\mathrm{P}\left(\mathrm{E}_{1} \mid \mathrm{E}_{2} \mathrm{E}_{3}\right) \mathrm{P}\left(\mathrm{E}_{2} \mid \mathrm{E}_{3}\right) \mathrm{P}\left(\mathrm{E}_{3}\right) \tag{2.14a}
\end{align*}
$$

$\mathrm{E}_{1}, \mathrm{E}_{2}$ and $\mathrm{E}_{3}$ are statistically independent.

$$
\begin{equation*}
\mathrm{P}\left(\mathrm{E}_{1} \mathrm{E}_{2} \mathrm{E}_{3}\right)=\mathrm{P}\left(\mathrm{E}_{1}\right) \mathrm{P}\left(\mathrm{E}_{2}\right) \mathrm{P}\left(\mathrm{E}_{3}\right) \tag{2.15a}
\end{equation*}
$$

## Mutually exclusive

If the occurrence of one event precludes the occurrence of another event
> Addition rule

Statistically independent
If the occurrence of one event does not affect the probability of another event
> Multiplication rule

Ex. Toss a coin

- heads or tails: Mutually exclusive
- first trial and second trial: Statistically independent

Ex. 2.20 Failure of Foundation

$$
\begin{array}{ll}
\mathrm{B}: \text { Failure of Bearing Capacity } & \mathrm{P}(\mathrm{~B})=0.001 \\
\mathrm{~S}: \quad \text { by Excessive Settlement } & \mathrm{P}(\mathrm{~S})=0.008 \\
\mathrm{P}(\mathrm{~B} \mid \mathrm{S})=0.1 &
\end{array}
$$

a) Probability of failure of foundation

$$
\begin{aligned}
\mathrm{P}(\mathrm{~B} \cup \mathrm{~S}) & =\mathrm{P}(\mathrm{~B})+\mathrm{P}(\mathrm{~S})-\mathrm{P}(\mathrm{BS})=\mathrm{P}(\mathrm{~B})+\mathrm{P}(\mathrm{~S})-\mathrm{P}(\mathrm{~B} \mid \mathrm{S}) \mathrm{P}(\mathrm{~S}) \\
& =0.001+0.008-0.1 \times 0.008=0.0082
\end{aligned}
$$

b) Probability of excessive settlement but no failure in bearing capacity

$$
\begin{aligned}
\mathrm{P}(\mathrm{~S} \overline{\mathrm{~B}}) & =\mathrm{P}(\overline{\mathrm{~B}} \mid \mathrm{S}) \mathrm{P}(\mathrm{~S})=(1-\mathrm{P}(\mathrm{~B} \mid \mathrm{S})) \mathrm{P}(\mathrm{~S}) \\
& =(1-0.1) 0.008=0.0072
\end{aligned}
$$

$$
P(B \mid S) \leq 1 / 8 \quad \text { Why? }
$$

### 2.3.4 Theorem of Total Probability

$\mathrm{E}_{1}, \mathrm{E}_{2}, \ldots \mathrm{E}_{\mathrm{n}}$ : mutually exclusive and collectively exhaustive events

$$
\begin{align*}
& \quad E_{i} \cap E_{j}=\Phi, \quad E_{1} \cup E_{2} \cup, \ldots \cup E_{n}=S \\
& A=A S \\
&=A\left(E_{1} \cup E_{2} \cup \ldots \cup E_{n}\right) \\
&=A E_{1} \cup A E_{2} \cup, \ldots \cup A_{n} \\
& P(A)=P\left(A E_{1}\right)+P\left(A E_{2}\right)+\ldots \ldots+P\left(A E_{n}\right) \\
&=P\left(A \mid E_{1}\right) P\left(E_{1}\right)+P\left(A \mid E_{2}\right) P\left(E_{2}\right)+\ldots+P\left(A \mid E_{n}\right) P\left(E_{n}\right)  \tag{2.19}\\
&=\sum_{i}^{n} P\left(A \mid E_{i}\right) P\left(E_{i}\right) \\
& \hline E_{2} \\
& \hline
\end{align*}
$$

## Ex. 2.27

F: Failure S: Storm T: Tornado H: Hit

$$
\begin{aligned}
P(F)= & P(F \mid S T H) P(S T H)+P(F \mid S T \bar{H}) P(S T \bar{H}) \\
& +P(F \mid S \bar{T}) P(S \bar{T})+P(F \mid \bar{S} \bar{T}) P(\bar{S} \bar{T}) \quad P(\bar{S} T)=\phi \\
P(S T H)= & P(H \mid S T) P(T \mid S) P(S)=0.01875 \\
P(S T \bar{H})=P(\bar{H} \mid S T) P(T \mid S) P(S)=0.10625 &
\end{aligned}
$$

$$
P(F \mid S \bar{T})=0.05
$$

$$
P(S \bar{T})=P(\bar{T} \mid S) P(S)
$$

$$
=0.75 \times 0.5=0.375
$$

$$
P(F \mid \bar{S} \bar{T})=0.0
$$

$$
P(F)=1.0 \times 0.01875+0.1 \times 0.10625
$$

$$
+0.05 \times 0.375=\underline{0.04812}
$$

| $\overline{\mathrm{S}}: 0.5$ | 0.0 |  | $\begin{aligned} & \overline{\mathrm{H}}: \\ & 0.85 \end{aligned}$ |
| :---: | :---: | :---: | :---: |
| S:0.5 | 0.05 | 0.1 |  |
| 25 |  | 1.0 | H: |
|  | $\overline{\mathbf{T}}: 0.75$ | T:0.25 |  |

### 2.3.5 Bayes' Theorem

$\mathrm{E}_{1}, \mathrm{E}_{2}, \ldots \mathrm{En}$ : mutually exclusive and collectively exhaustive events

$$
\begin{aligned}
& \mathrm{P}\left(\mathrm{AE}_{\mathrm{i}}\right)=\mathrm{P}\left(\mathrm{~A} \mid \mathrm{E}_{\mathrm{i}}\right) \mathrm{P}\left(\mathrm{E}_{\mathrm{i}}\right) \\
& \mathrm{P}\left(\mathrm{E}_{\mathrm{i}} \mathrm{~A}\right)=\mathrm{P}\left(\mathrm{E}_{\mathrm{i}} \mid \mathrm{A}\right) \mathrm{P}(\mathrm{~A}) \\
& \mathrm{P}\left(\mathrm{E}_{\mathrm{i}} \mid \mathrm{A}\right) \mathrm{P}(\mathrm{~A})=\mathrm{P}\left(\mathrm{~A} \mid \mathrm{E}_{\mathrm{i}}\right) \mathrm{P}\left(\mathrm{E}_{\mathrm{i}}\right)
\end{aligned}
$$



$$
\begin{align*}
P\left(E_{i} \mid A\right) & =\frac{P\left(A \mid E_{i}\right) P\left(E_{i}\right)}{P(A)}  \tag{2.20}\\
& =\frac{P\left(A \mid E_{i}\right) P\left(E_{i}\right)}{\sum_{j} P\left(A \mid E_{j}\right) P\left(E_{j}\right)}
\end{align*}
$$

Reverse or Inverse
Probability

## Ex. 2.30

G: Good-Quality

$$
\mathrm{P}(\mathrm{G})=0.8 \quad \mathrm{P}(\overline{\mathrm{G}})=0.2
$$

T: Sample pass the test

$$
\mathrm{P}(\mathrm{~T} \mid \mathrm{G})=0.9 \quad \mathrm{P}(\mathrm{~T} \mid \overline{\mathrm{G}})=0.1
$$

Sample passed the test. Knowing the fact,

$$
\begin{aligned}
P\left(G \mid T_{1}\right) & =\frac{P\left(G T_{1}\right)}{P\left(T_{1}\right)}=\frac{P\left(T_{1} \mid G\right) P(G)}{P\left(T_{1} \mid G\right) P(G)+P\left(T_{1} \mid \bar{G}\right) P(\bar{G})} \\
& =\frac{0.9 \times 0.8}{0.9 \times 0.8+0.1 \times 0.2}=0.973
\end{aligned}
$$



$$
\begin{aligned}
P\left(G \mid T_{2}\right)= & \frac{P\left(T_{2} \mid G\right) P(G)}{P\left(T_{2} \mid G\right) P(G)+P\left(T_{2} \mid \bar{G}\right) P(\bar{G})} & P\left(G \mid T_{1} \bar{T}_{2}\right) & =\frac{P\left(T_{1} \bar{T}_{2} \mid G\right) P(G)}{P\left(T_{1} \bar{T}_{2} \mid G\right) P(G)+P\left(T_{1} \bar{T}_{2} \mid \bar{G}\right) P(\bar{G})} \\
& =\frac{0.9 \times 0.973}{0.9 \times 0.973+0.1 \times 0.027}=0.997 & & =\frac{0.9 \times 0.1 \times 0.8}{0.9 \times 0.1 \times 0.8+0.1 \times 0.9 \times 0.2}=0.80
\end{aligned}
$$

$\mathrm{P}(\mathrm{G})$ : Prior Probability, $\mathrm{P}(\mathrm{G} \mid \mathrm{T})$ : Posterior Probability
Bayes' theorem is useful to for revising or updating the calculated probability as more data and information becomes available. [Bayesian Updating]

