

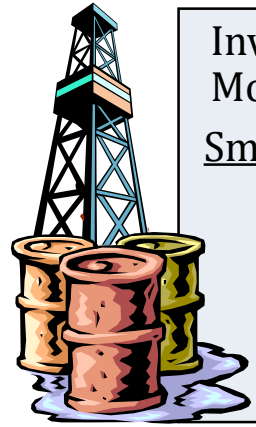
# **Session 9: Probability**

Module BUSN9690


**Business Statistics with Python**

# Example 1: Mick's Investments

Mick has invested in two stocks, Smith Oil and David Mining. He estimated that the possible outcomes of these investments three months from now were as follows.



Investment Gain or Loss in 3 Months (in £1,000)	
<u>Smith Oil</u>	<u>David Mining</u>
10	8
5	-2
0	
-20	



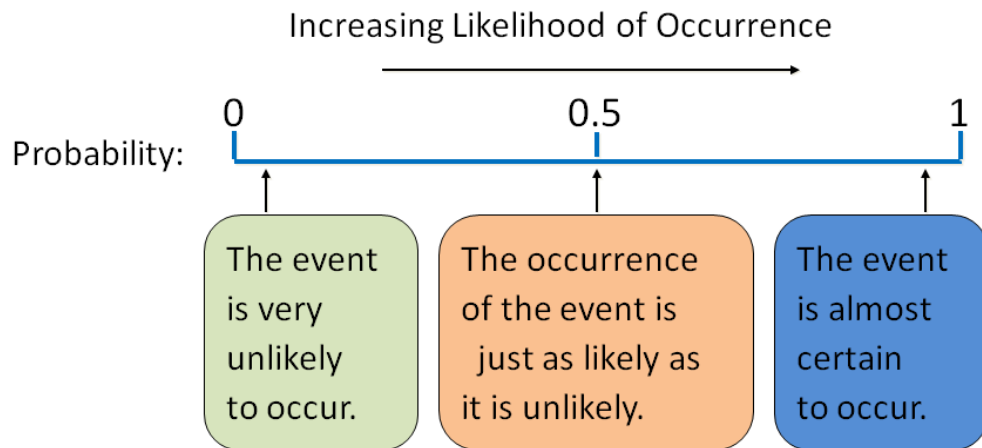
## Questions

- What is the probability that Smith Oil is profitable?
- What is the probability that Smith Oil **or** David Mining is profitable?
- What is the probability that Smith Oil **and** David Mining are profitable?
- What is the probability that David Mining is profitable, **given that** Smith Oil is profitable?

# Probability (Classical definition)

- The ratio of the number of favourable outcomes to the total number of outcomes

$$P(A) = \frac{\text{Number of outcomes leading to event } A}{\text{Total number of possible outcomes}}$$



$$0 \leq \text{Probability} \leq 1$$

# Sample space & event sample

## Experiment

## Sample Space

(possible outcomes)

Toss 2 Coins, Note Faces

HH, HT, TH, TT

Play a Football Game

Win, Lose, Tie

Inspect a Part, Note Quality

Defective, OK

- Experiment: Toss 2 Coins. Note Faces.

- Sample Space: HH, HT, TH, TT

## Events

## Outcomes in event

1 Head & 1 Tail

HT, TH

Heads on 1st Coin

HH, HT

At Least 1 Head

HH, HT, TH

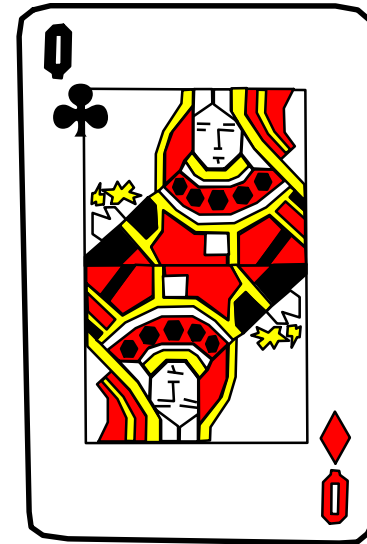
Heads on Both

HH

# Null event

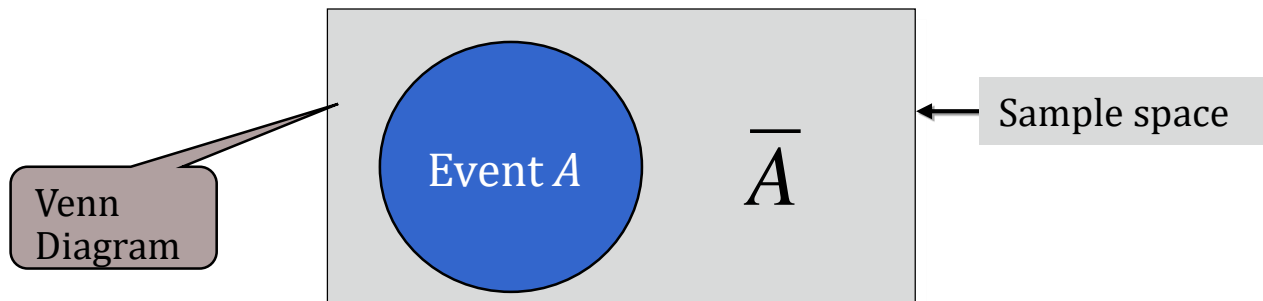
- **Null Event, denoted with  $\Phi$** 
  - Club & Diamond on Card Draw

Null Event



# Complement of an Event

The complement of event  $A$  is defined to be the event consisting of all sample points that are not in  $A$ . The complement of  $A$  is denoted  $\bar{A}$



Example: The probability that a MSc student can pass his exam is  $p$ . The probability that he will fail is  $1-p$

$$P(\text{success}) = 1 - P(\text{failure})$$

**Profitable and non-profitable???**

# Compound events

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## ● Union

- Outcomes in either Event A **or** Event B **or** Both
- 'OR' Statement
- $\cup$  Symbol (i.e.,  $A \cup B$ , A or B)

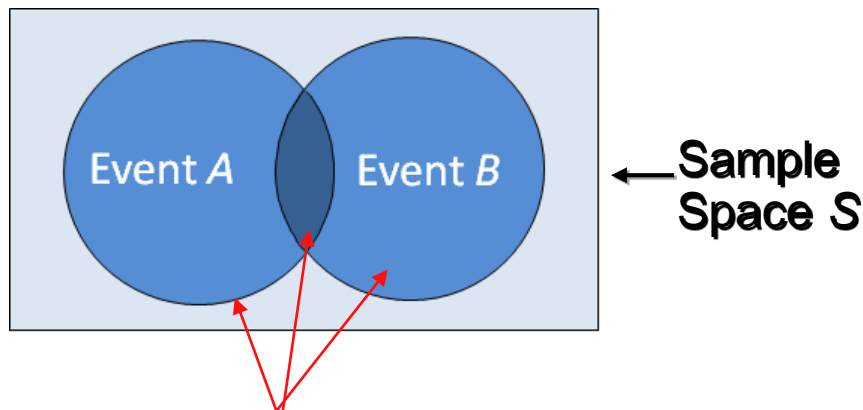
## ● Intersection

- Outcomes in Both Events A **and** B
- 'AND' Statement
- $\cap$  Symbol (i.e.,  $A \cap B$ , A & B)

# Union of Two Events

The union of events  $A$  and  $B$  is the set containing all sample points that are in  $A$  or  $B$  or both.

The union of events  $A$  and  $B$  is denoted by  $A \cup B$ ,  $A$  or  $B$



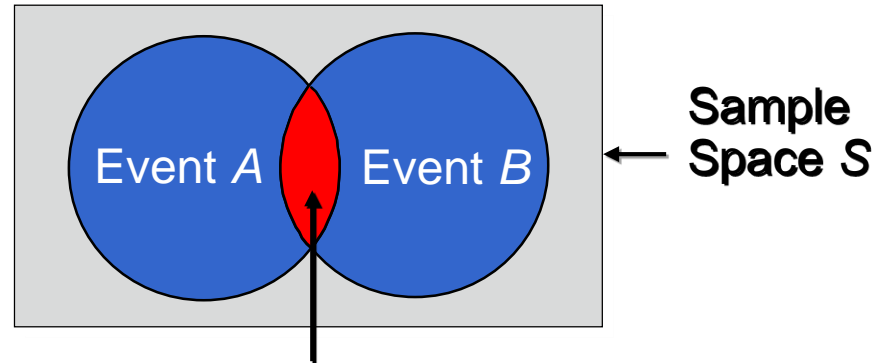
The entire blue area, including light blue and dark blue, is  $A \cup B$



# Intersection of Two Events

The intersection of events  $A$  and  $B$  is the set of all sample points that are in both  $A$  and  $B$ .

The intersection of events  $A$  and  $B$  is denoted by  $A \& B$ ,  $A \cap B$



The red area is  $A \cap B$

# Sets in Python

`business_analytics={'machine learning','big data','stats with python'}` #modules of business analytics

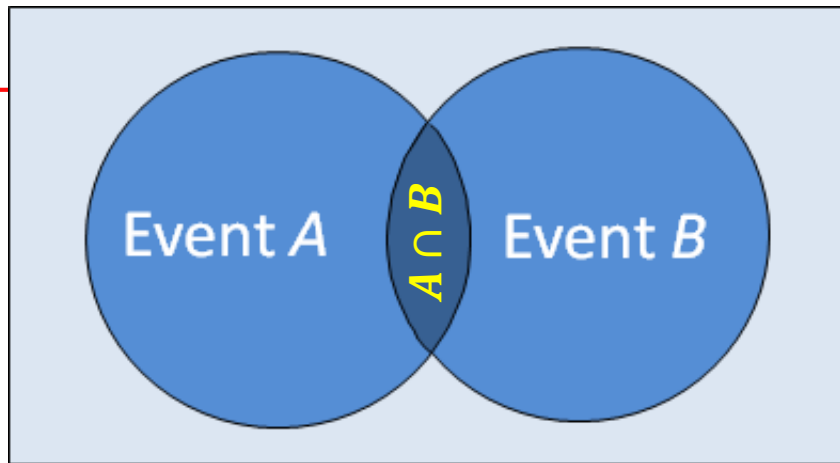
`supply_chain={'machine learning','supply chain'}` #modules of supply chain

`common_modules=business_analytics.intersection(supply_chain)` #modules shared by both

`all_modules=business_analytics.union(supply_chain)` # all modules

```
>>> business_analytics={'machine learning','big data','stats with python'}
>>> supply_chain={'machine learning','supply chain'}
>>> common_modules=business_analytics.intersection(supply_chain)
>>> all_modules=business_analytics.union(supply_chain)
>>> common_modules
{'machine learning'}
>>> all_modules
{'big data', 'supply chain', 'machine learning', 'stats with python'}
>>>
```

# Addition Law



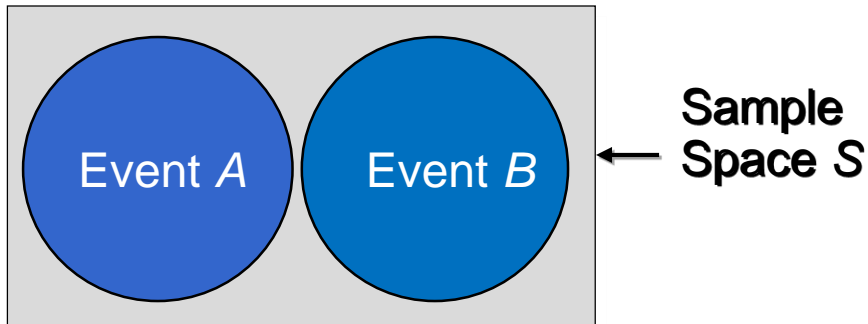
The addition law provides a way to compute the probability of event  $A$ , or  $B$ , or both  $A$  and  $B$  occurring.

The law is written as:

$$P(A \text{ or } B) = P(A) + P(B) - P(A \cap B)$$

# Mutually Exclusive Events

Two events are mutually exclusive if, when one event occurs, the other cannot occur.



If events  $A$  and  $B$  are mutually exclusive,  $P(A \cap B) = 0$ .

The addition law for mutually exclusive events is:


$$P(A \text{ or } B) = P(A) + P(B)$$

there's no need to include " $- P(A \cap B)$ "


# **Examples, multiplication rule, conditional probability**

# Example: Mick's Investments

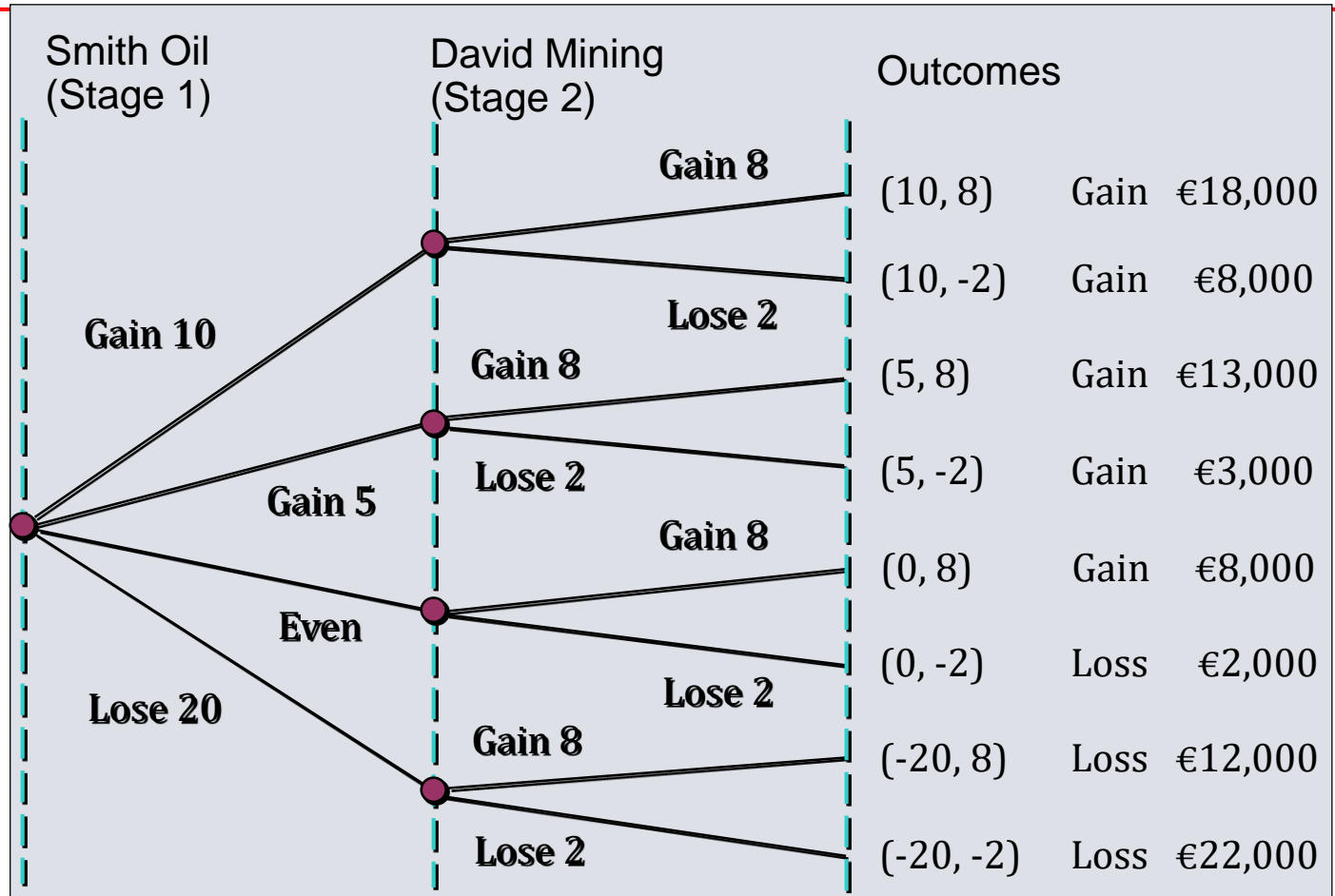
Mick has invested in two stocks, Smith Oil and David Mining. He estimated that the possible outcomes of these investments three months from now were as follows.



Investment Gain or Loss in 3 Months (in £1,000)	
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# A Counting Rule for Multiple-Step Experiments

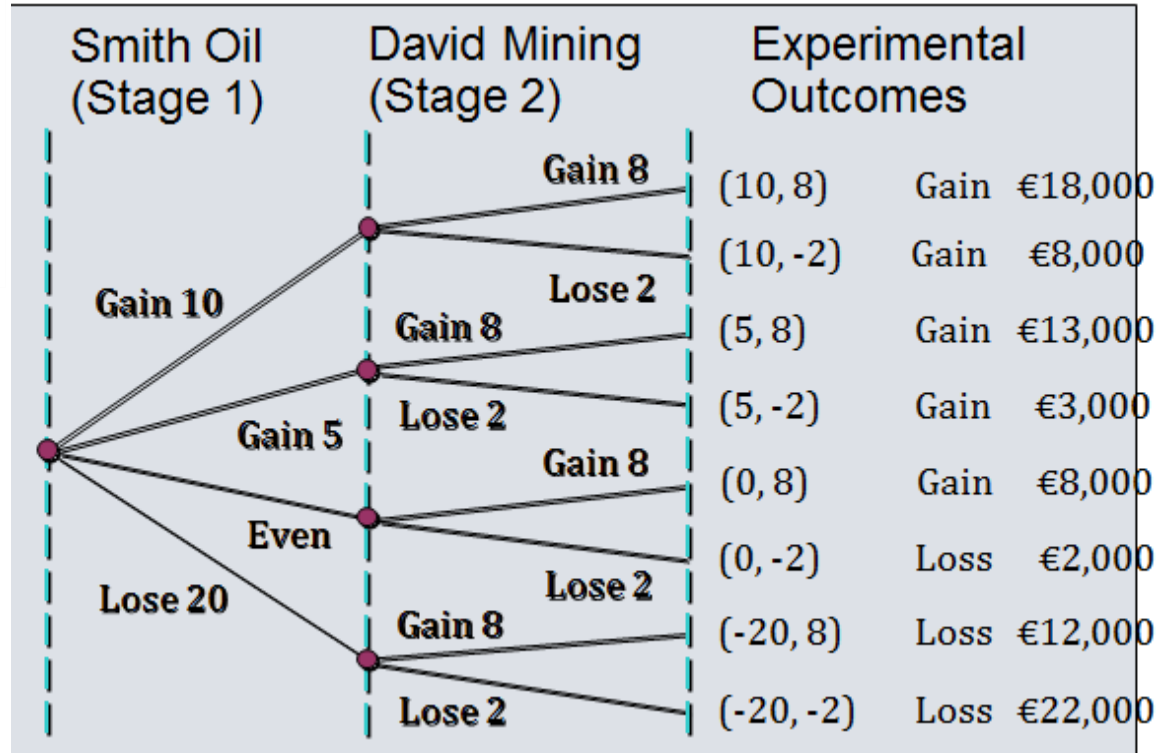


# A Counting Rule for Multiple-Step Experiments

Mick's investments can be viewed as a two-step experiment. It involves two stocks, each with a set of experimental outcomes.

- Smith Oil:  $n_1 = 4$
- David Mining:  $n_2 = 2$
- Total Number of Experimental Outcomes:  
 $n_1 \times n_2 = 4 \times 2 = 8$

Tree diagram





# An example

Applying the subjective method, an analyst made the following probability assignments.

<u>Exper.</u>	<u>Outcome</u>	<u>Net Gain or Loss</u>	<u>Probability</u>
	(10, 8)	£18,000 Gain	.20
	(10, -2)	£8,000 Gain	.08
	(5, 8)	£13,000 Gain	.16
	(5, -2)	£3,000 Gain	.26
	(0, 8)	£8,000 Gain	.10
	(0, -2)	£2,000 Loss	.12
	(-20, 8)	£12,000 Loss	.02
	(-20, -2)	£22,000 Loss	.06

# Events and their Probabilities

<u>Exper. Outcome</u>	<u>Net Gain or Loss</u>	<u>Probability</u>
(10, 8)	€18,000 Gain	.20
(10, -2)	€8,000 Gain	.08
(5, 8)	€13,000 Gain	.16
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(0, 8)	€8,000 Gain	.10
(0, -2)	€2,000 Loss	.12
(-20, 8)	€12,000 Loss	.02
(-20, -2)	€22,000 Loss	.06

Event  $S$  = Smith Oil Profitable

$$S = \{(10, 8), (10, -2), (5, 8), (5, -2)\}$$

$$\begin{aligned} P(S) &= P(10, 8) + P(10, -2) + P(5, 8) + P(5, -2) \\ &= .20 + .08 + .16 + .26 \\ &= \textcircled{.70} \end{aligned}$$

Event  $D$  = David Mining Profitable

$$D = \{(10, 8), (5, 8), (0, 8), (-20, 8)\}$$

$$\begin{aligned} P(D) &= P(10, 8) + P(5, 8) + P(0, 8) + P(-20, 8) \\ &= .20 + .16 + .10 + .02 \\ &= \textcircled{.48} \end{aligned}$$

# Probability of the Union of Two Events

Event  $S$  = Smith Oil Profitable

Event  $D$  = David Mining Profitable

$S$  or  $D$  = Smith Oil Profitable  
or David Mining Profitable

$S$  or  $D$  =  $\{(10, 8), (10, -2), (5, 8), (5, -2), (0, 8), (-20, 8)\}$

$$\begin{aligned} P(S \text{ or } D) &= P(10, 8) + P(10, -2) + P(5, 8) + P(5, -2) \\ &\quad + P(0, 8) + P(-20, 8) \\ &= .20 + .08 + .16 + .26 + .10 + .02 \\ &= \textcircled{0.82} \end{aligned}$$

Direct calculation

Use the formula

$$P(A \text{ or } B) = P(A) + P(B) - P(A \cap B)$$



Event  $S$  = Smith Oil Profitable

Event  $D$  = David Mining Profitable

$S$  or  $D$  = Smith Oil Profitable  
or David Mining Profitable

We know:  $P(S) = .70$ ,  $P(D) = .48$ ,  $P(S \cap D) = .36$

Thus:  $P(S \text{ or } D) = P(S) + P(D) - P(S \cap D)$

$$\begin{aligned} &= .70 + .48 - .36 \\ &= \textcircled{.82} \end{aligned}$$

(This result is the same as that obtained earlier using the definition of the probability of an event.)

# Multiplication Law

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The multiplication law provides a way to compute the probability of the **intersection** of two events.

The law is written as:  $P(A \cap B) = P(B)P(A|B)$   
 $P(A \cap B) = P(A)P(B|A)$

# Conditional probability

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The probability of an event given that another event has occurred is called a conditional probability.

The conditional probability of  $A$  given  $B$  is denoted by  $P(A|B)$ .

The conditional probability can be obtained from the formula

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

# Independent Events

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If the probability of event  $A$  is not changed by the existence of event  $B$ , we would say that events  $A$  and  $B$  are independent.

Two events  $A$  and  $B$  are independent if:

$$P(A|B) = P(A)$$

or

$$P(B|A) = P(B)$$

**The multiplication law for Independent events becomes**

$$P(A \cap B) = P(A)P(B)$$

# Intersection of Two Events

<u>Exper. Outcome</u>	<u>Net Gain or Loss</u>	<u>Probability</u>
(10, 8)	€18,000 Gain	.20
(10, -2)	€8,000 Gain	.08
(5, 8)	€13,000 Gain	.16
(5, -2)	€3,000 Gain	.26
(0, 8)	€8,000 Gain	.10
(0, -2)	€2,000 Loss	.12
(-20, 8)	€12,000 Loss	.02
(-20, -2)	€22,000 Loss	.06

We are unable to use the following formula yet as we do not know the conditional probability at this moment

The law is written as:  $P(A \cap B) = P(B)P(A|B)$

$$P(A \cap B) = P(A)P(B|A)$$

Event  $S$  = Smith Oil Profitable

Event  $D$  = David Mining Profitable

$S \cap D$  = Smith Oil Profitable

and David Mining Profitable

$$S \cap D = \{(10, 8), (5, 8)\}$$

$$P(S \cap D) = P(10, 8) + P(5, 8)$$

$$= .20 + .16$$

$$= .36$$

# To decide if events are independent

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Event  $S$  = Smith Oil Profitable

Event  $D$  = David Mining Profitable

Are events  $S$  and  $D$  independent?

Does  $P(S \cap D) = P(S)P(D)$  ?

We know:  $P(S \cap D) = .36$ ,  $P(S) = .70$ ,  $P(D) = .48$

But:  $P(S)P(D) = (.70)(.48) = .34$ , not  $.36$

Hence:  $S$  and  $D$  are not independent.



# Conditional probability

$$P(A|B) = \frac{P(A \cap B)}{P(B)}$$

Event  $S$  = Smith Oil Profitable

Event  $D$  = David Mining Profitable

$P(D|S)$  = David Mining Profitable given Smith Oil Profitable

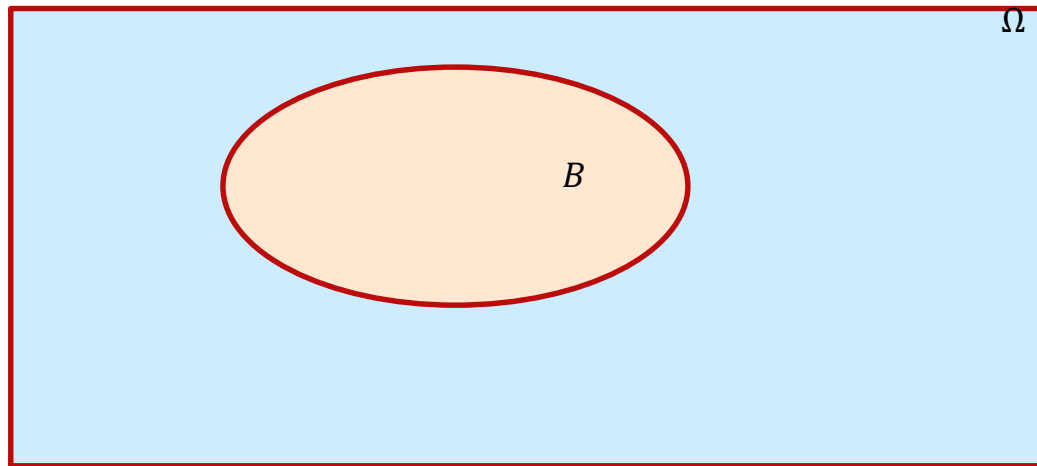
We know:  $P(S \cap D) = .36$ ,  $P(S) = .70$

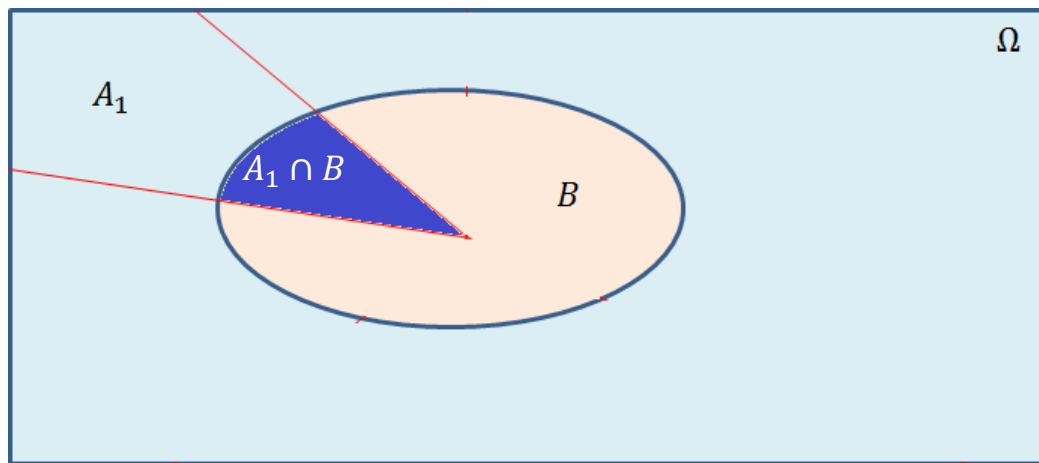
$$\text{Thus: } P(D|S) = \frac{P(S \cap D)}{P(S)} = \frac{0.36}{0.70} = 0.5143$$

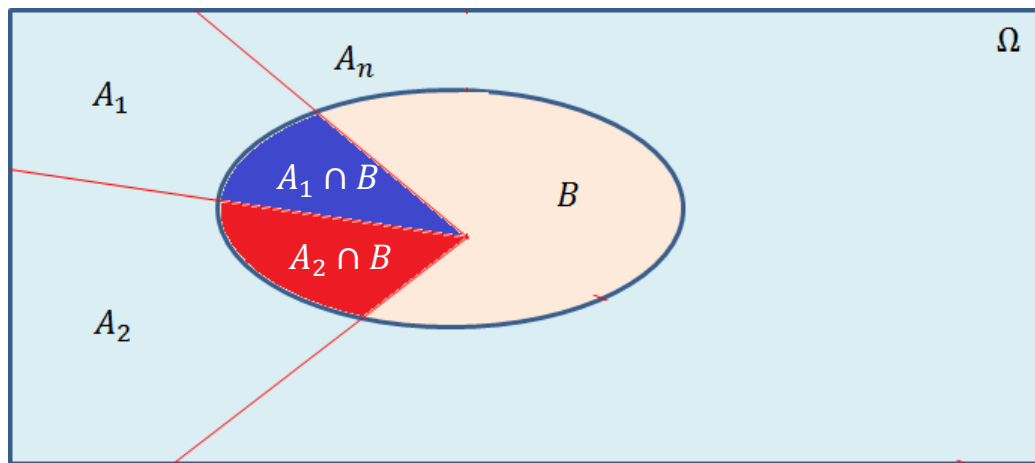
# **The theorem of total probability, Bayes theorem**

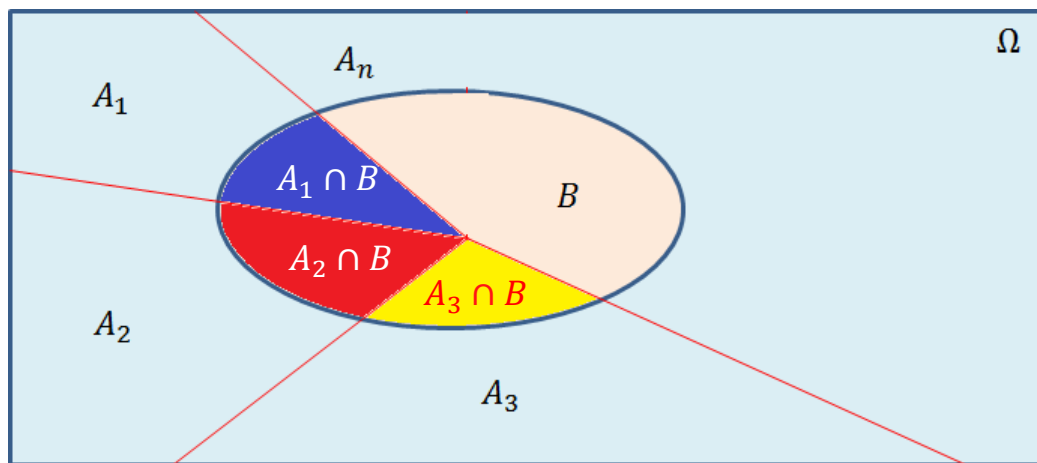
# The Theorem of Total Probability

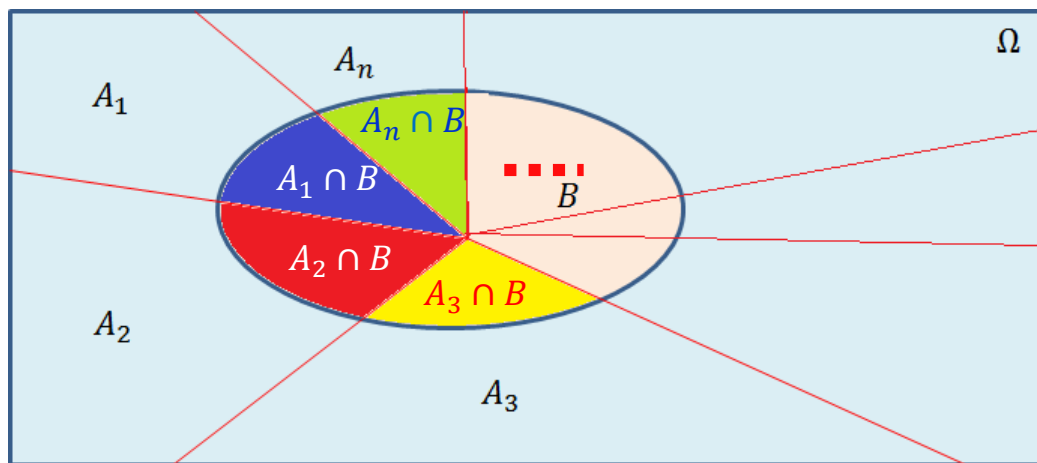
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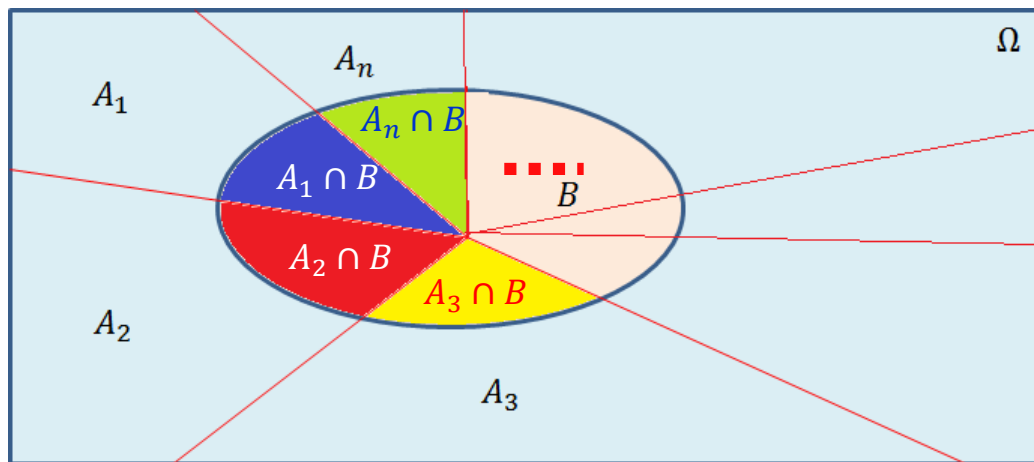








# Law of total probability



- $(A_i \cap B) \cap (A_j \cap B) = \emptyset, \text{ where } i \neq j$
- $B = (A_1 \cap B) \cup (A_2 \cap B) \cup \dots \cup (A_n \cap B)$

## Law of total probability:

- Then 
$$P(B) = P(A_1 \cap B) + P(A_2 \cap B) + \dots + P(A_n \cap B)$$
$$= P(A_1)P(B|A_1) + P(A_2)P(B|A_2) + \dots + P(A_n)P(B|A_n)$$



# Bayes' Theorem

- Then  $P(B) = P(A_1 \cap B) + P(A_2 \cap B) + \dots + P(A_n \cap B)$   
 $= P(A_1)P(B|A_1) + P(A_2)P(B|A_2) + \dots + P(A_n)P(B|A_n)$

■ To find the posterior probability,  $P(A_i|B)$ , that event  $A_i$  will occur given that event  $B$  has occurred, we apply

Bayes' theorem.

$$P(A_i|B) = \frac{P(A_i \cap B)}{P(B)} = \frac{P(A_i)P(B|A_i)}{P(A_1)P(B|A_1) + P(A_2)P(B|A_2) + \dots + P(A_n)P(B|A_n)}$$

- $P(A_i)$ : Prior probability
- $P(B|A_i)$ : Likelihood
- $P(A_i|B)$ : Posterior probability

Calculating the posterior distribution can be difficult. Some methods such as MCMC (Markov chain Monte Carlo) and Variational Inference can be applied

## Example 2: Application of Bayes' Theorem

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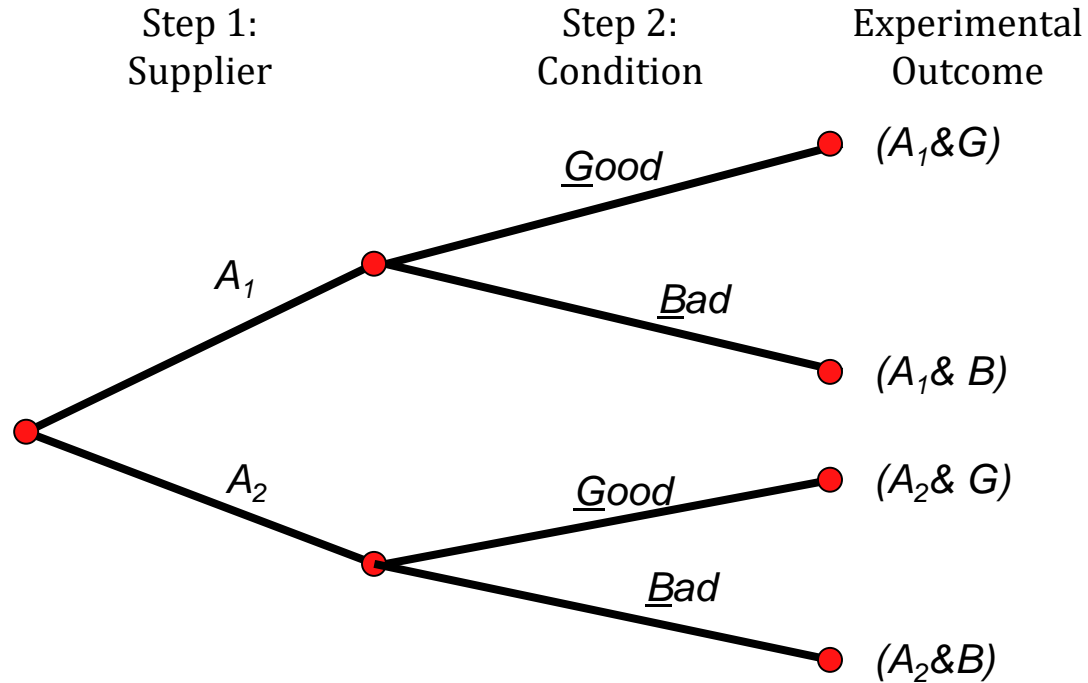
- Consider a manufacturing firm that receives shipment of parts from two suppliers.
- Let
  - $A_1$ : event that a part is received from supplier 1; and
  - $A_2$ : event that a part is received from supplier 2
- Suppose that
  - 65% of our parts are supplied by supplier 1 and
  - 35% of our parts are supplied by supplier 2.
  - That is:
    - $P(A_1) = 0.65$  and
    - $P(A_2) = 0.35$ .
- **The above gives prior probabilities**

# Quality levels differ between suppliers

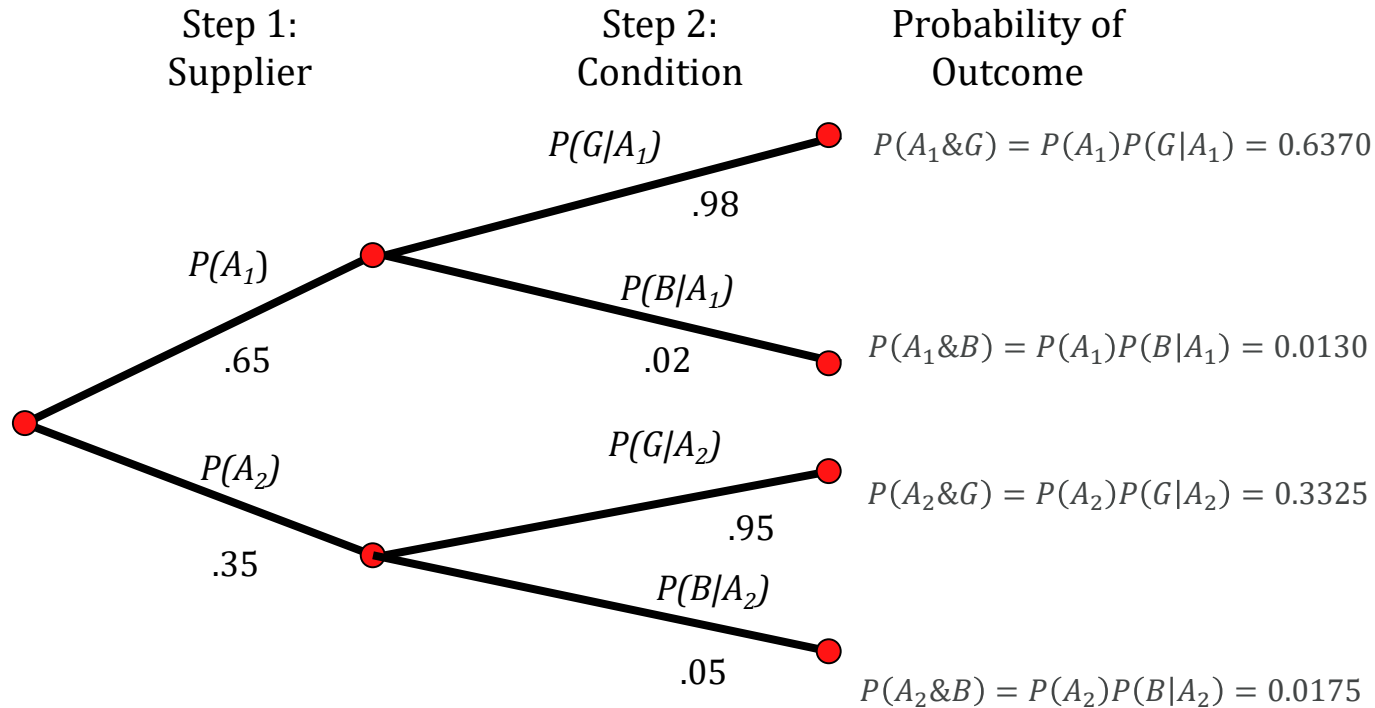
	Percentage Good Parts	Percentage Bad Parts
Supplier 1	98	2
Supplier 2	95	5

- Let
  - G: a part is good, and
  - B: a part is bad
- According to the above table, we have
  - $P(G|A_1) = 0.98$  and  $P(B|A_1) = 0.02$
  - $P(G|A_2) = 0.95$  and  $P(B|A_2) = 0.05$
- **The above gives new information, or likelihood**

# Tree Diagram for Two-Supplier Example



# Probability Tree for the Two-Supplier Example



# Application of Bayes' theorem

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$$\begin{aligned}P(A_1 | B) &= \frac{P(A_1)P(B | A_1)}{P(A_1)P(B | A_1) + P(A_2)P(B | A_2)} \\&= \frac{(.65)(.02)}{(.65)(.02) + (.35)(.05)} = \frac{.0130}{.0305} = .4262\end{aligned}$$

$$\begin{aligned}P(A_2 | B) &= \frac{P(A_2)P(B | A_2)}{P(A_1)P(B | A_1) + P(A_2)P(B | A_2)} \\&= \frac{(.35)(.05)}{(.65)(.02) + (.35)(.05)} = \frac{.0175}{.0305} = .5738\end{aligned}$$

**The above obtains the posterior probabilities, which updates our knowledge**

# Summary I

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- Sample space
- Events
  - Null events
  - Complementary events
  - Mutually exclusive events
  - Independent events
- Compound events
  - Union
  - Intersection
- Conditional probability
- Bayes' theorem

# Summary II

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- OR

- $P(A \cup B) = P(A) + P(B) - P(A \cap B)$

- AND

- $P(A \cap B) = P(A)P(B|A)$

- $P(A \cap B) = P(B)P(A|B)$

- Exclusive

- If A and B are mutually exclusive, then  $P(A \cap B) = 0$

- Independent

- If A is not independent of B, then

- $P(A \cap B) = P(A)P(B|A)$

- If A is independent of B, then

- $P(A \cap B) = P(A)P(B)$