

# Decision Trees & Probabilities

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# Learning Objectives and Outline

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# Learning Objectives

- Construct and solve a decision problem by calculating an intervention's **expected value** across competing strategies in a decision tree
- Determine the decision threshold across a range of scenarios
- Differentiate between joint and conditional probabilities and demonstrate their use in decision trees

# Outline

1. Structure
2. Probability review
3. Probabilities within decision trees
4. Strengths/limitations of decision trees

# The Structure of a Decision Analysis

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# Recap: Decision Analysis

- Aims to inform choice under uncertainty using an explicit, quantitative approach
- Aims to identify, measure, & value the **consequences of decisions** (risks/benefits) & uncertainty when a decision needs to be made, most appropriately over time

# Should I go to the beach or stay home?

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# Should I go to the beach or stay home?

## Possible States of the World:

- At the beach with no rain.
- At the beach with rain.
- At home with no rain.
- At home with rain.

# Should I go to the beach or stay home?

## Considerations:

- Likelihood of rain
- My overall well being when
  - At the beach with no rain.
  - At the beach with rain.
  - At home with no rain.
  - At home with rain.

# Should I go to the beach or stay home?

## Considerations:

- Likelihood of rain → probabilities
- My overall well being when
  - At the beach with no rain.
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# Should I go to the beach or stay home?

## Considerations:

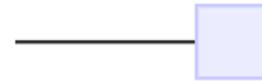
- Likelihood of rain → probabilities
- My overall well being when
  - At the beach with no rain. → payoff
  - At the beach with rain. → payoff
  - At home with no rain. → payoff
  - At home with rain. → payoff

# Decision Trees

- A square decision node indicates a decision point between alternative options.
- A circular chance node shows a point where two or more alternative events for a patient are possible.

# Decision Trees

- A square decision node indicates a decision point between alternative options.
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Decision Node



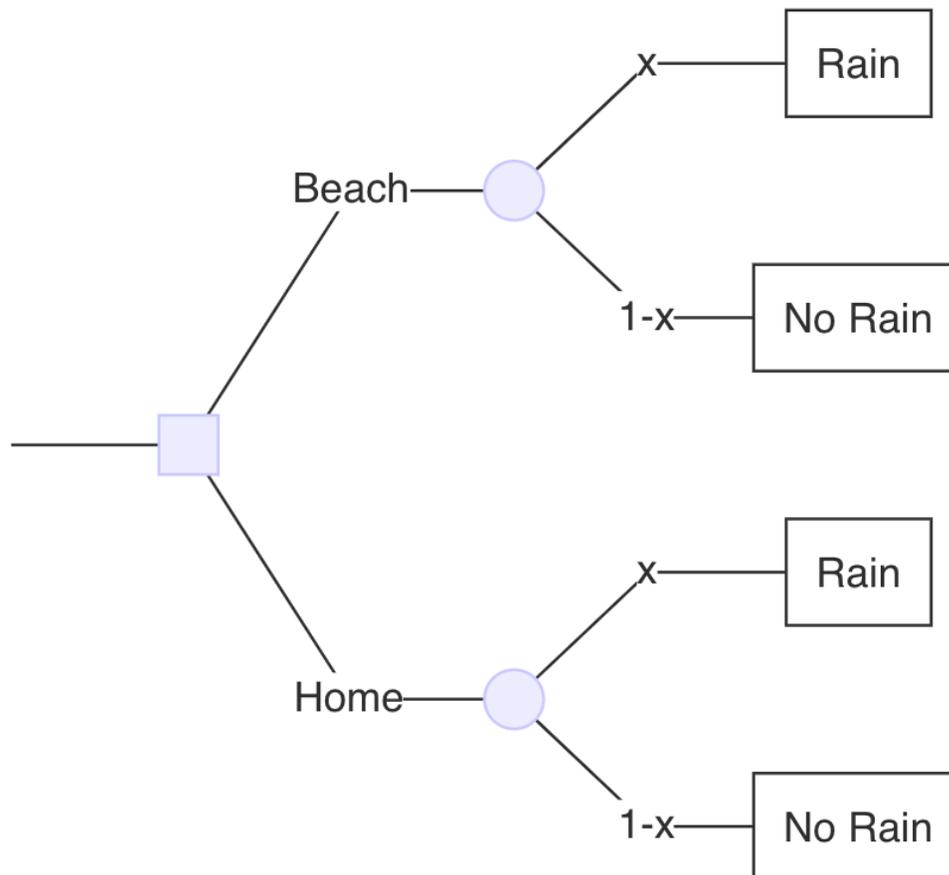
Chance Node

# Decision Trees

- Pathways are mutually exclusive sequences of events and are the routes through the tree.
- Probabilities show the likelihood of particular event occurring at a chance node.

# Should I go to the beach or stay home?

## Decision Tree:



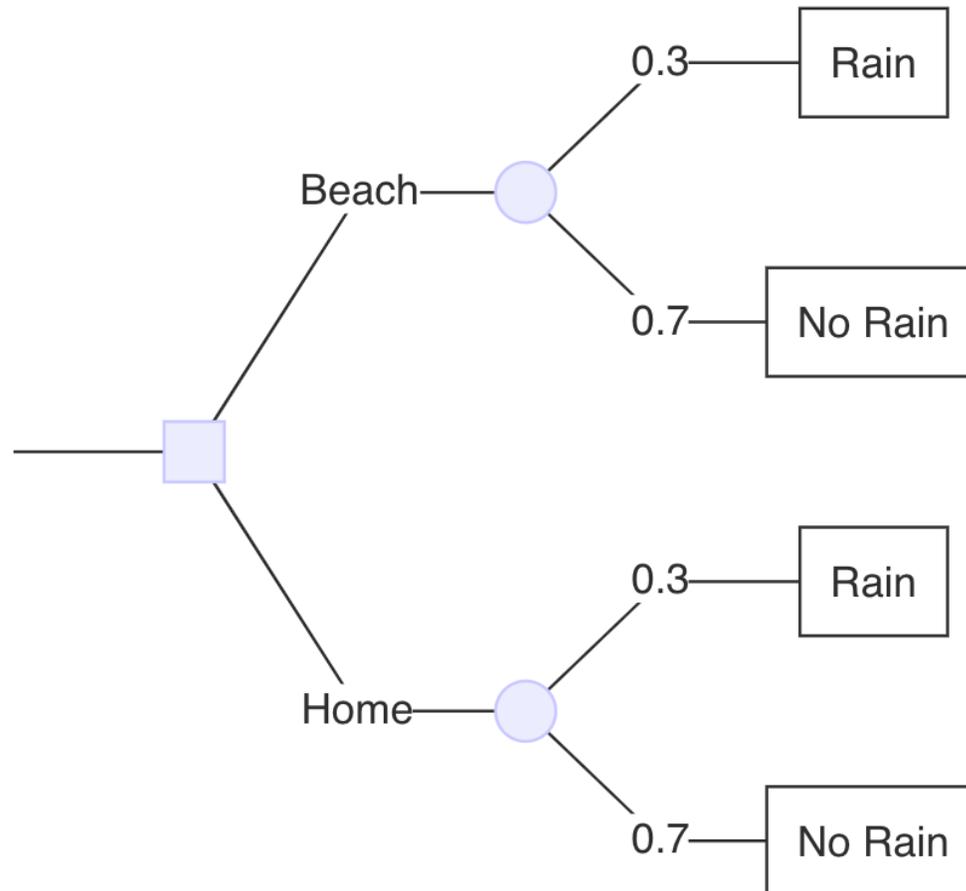
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Probability of rain = 30%

# Should I go to the beach or stay home?

## Decision Tree:



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# Should I go to the beach or stay home?

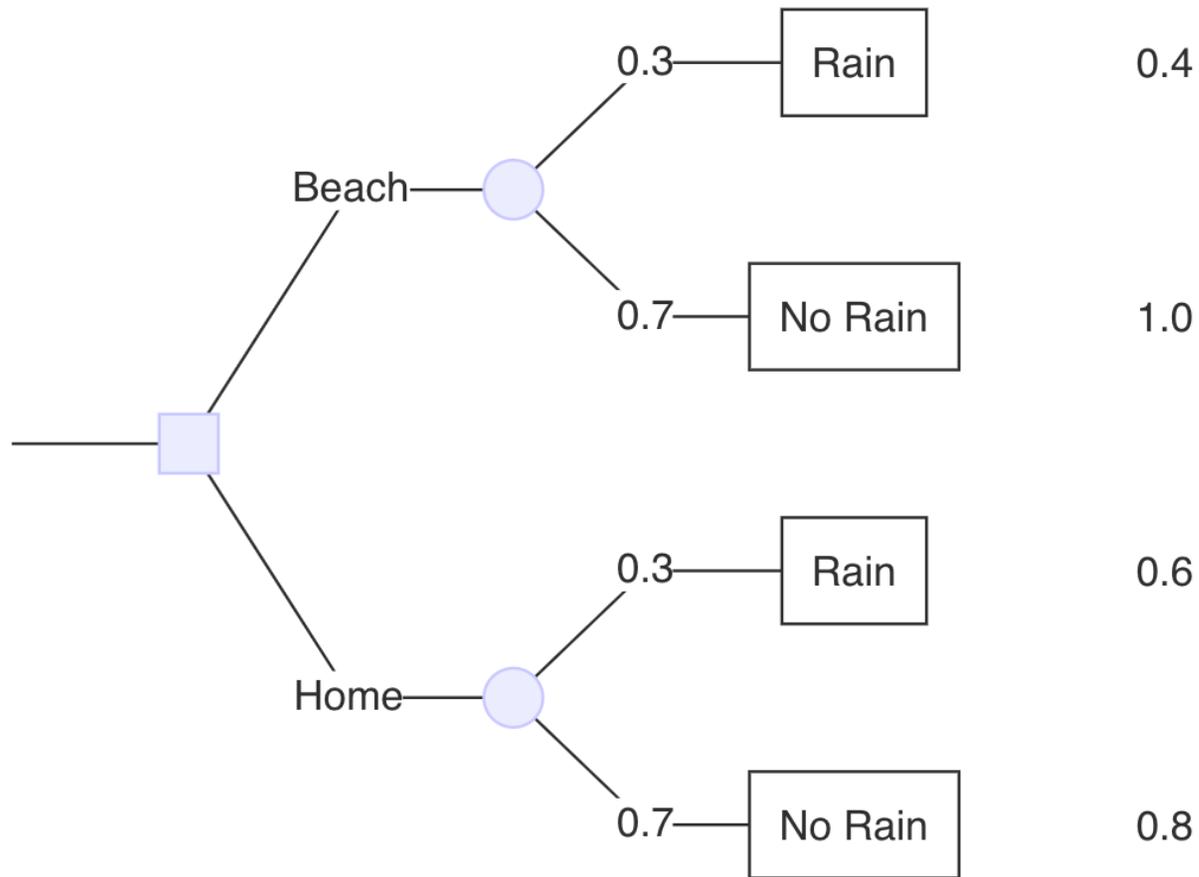
## Payoffs

Scenario	Payoff
At beach, no rain	1.0
At beach, rain	0.4
At home, no rain	0.8
At home, rain	0.6

At beach, no rain > At home, no rain > At home, rain > At beach, rain

# Should I go to the beach or stay home?

## Decision Tree:



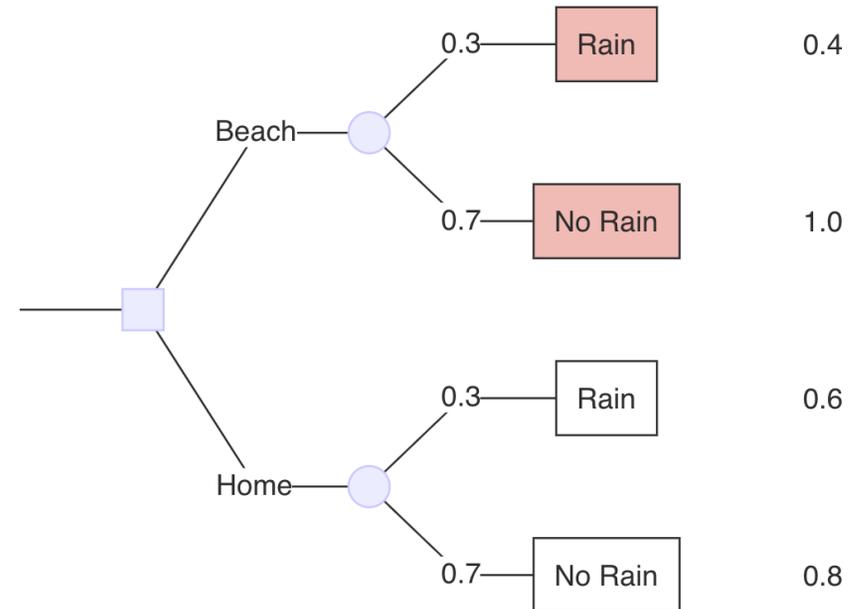
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# What is the expected value of going to the beach?

$$0.82 = \underbrace{0.3 * 0.4}_{\text{Rain}} +$$

$$\underbrace{0.7 * 1.0}_{\text{No Rain}}$$

- Probabilities in **red**.
- Payoffs in **blue**.
- Expected value in **green**.

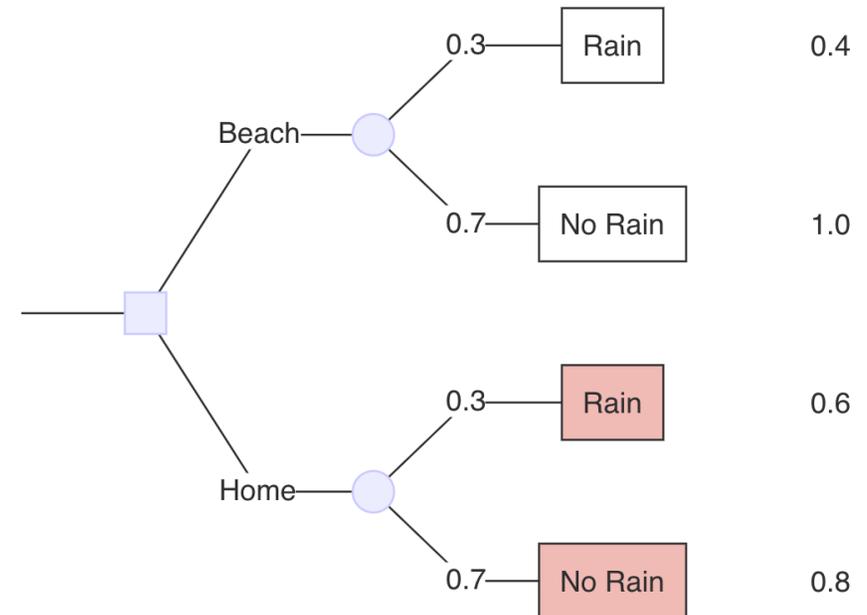


# What is the expected value of staying home?

$$0.74 = \underbrace{0.3 \cdot 0.6}_{\text{Rain}} +$$

$$\underbrace{0.7 \cdot 0.8}_{\text{No Rain}}$$

- Probabilities in **red**.
- Payoffs in **blue**.
- Expected value in **green**.



# Beach

$EV(\text{Beach})=0.82 > EV(\text{Home})=0.74$

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# Expected Values

- Expected value = The sum of the multiplied probabilities for each chance option or intervention

$$E[x] = \sum_{x} \sum_{i=1}^{\infty} X_i P(X_i)$$

# Expected Values

- The expected value within the context of decision trees are the “payoffs” weighted by their preceding probabilities
- What we get is: the result that is expected ON AVERAGE for any one decision alternative (e.g. length of life, quality of life, lifetime costs)
  - Example: On average, patients given Treatment A will live 0.30 years longer than patients given Treatment B

⚠ Maximizing expected value is a reasonable criterion for choice given uncertain prospects; though it does not necessarily promise the best results for any **one individual**

# Determining the Decision Threshold

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# At what probability $p$ are the two choices equal?

Suppose we want to know at what probability of rain  $p$  we are indifferent between going to the beach vs. staying at home...

- Write the equation for each choice using a variable,  $p$ , for the probability in question
- Set the equations equal to to one other and solve for  $p$ .

At what probability  $p$  are the two choices equal?

Beach:  $0.82 = 0.3 \times 0.4 + 0.7 \times 1.0$

Home:  $0.74 = 0.3 \times 0.6 + 0.7 \times 0.8$

At what probability  $p$  are the two choices equal?

$$\underbrace{0.3 * 0.4 + 0.7 * 1.0}_{\text{Beach}} = \underbrace{0.3 * 0.6 + 0.7 * 0.8}_{\text{Home}}$$

# At what probability $p$ are the two choices equal?

Replace probability of rain with  $P$  and  $1-P$  and solve for “ $P$ ”

$$p * 0.4 + (1-p) * 1.0 = p * 0.6 + (1-p) * 0.8$$

$$\underbrace{0.3 * 0.4 + 0.7 * 1.0}_{\text{Beach}} = \underbrace{0.3 * 0.6 + 0.7 * 0.8}_{\text{Home}}$$

At what probability  $p$  are the two choices equal?

$$p * 0.4 + (1-p) * 1.0 = p * 0.6 + (1-p) * 0.8$$

**At what probability  $p$  are the two choices equal?**

$$p * 0.4 + (1-p) * 1.0 = p * 0.6 + (1-p) * 0.8$$

$$0.4p + 1-p = 0.6p + 0.8 - 0.8p$$

$$1-0.6p = 0.8 - 0.2p$$

$$1-0.8 = 0.6p - 0.2p$$

$$0.2 = 0.4 * p$$

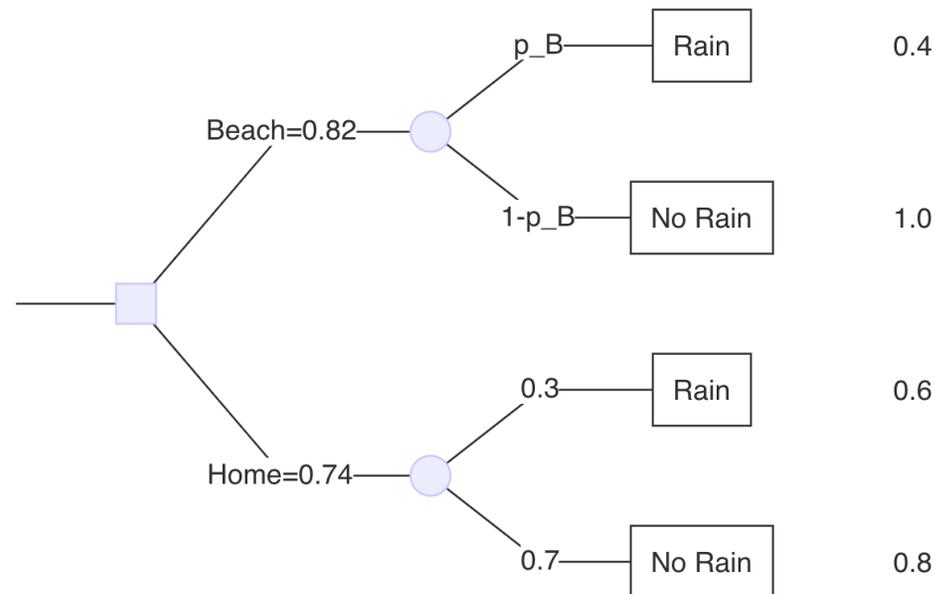
$$0.5 = p$$

# At what probability $p$ are the two choices equal?

When the probability of rain is 50% at BOTH the beach and home, given how we weighted the outcomes, going to the beach would be the same as staying at home

In other words, you would be indifferent between the two – staying at home or going to the beach

# At what probability ( $p_B$ ) of rain for the beach are you indifferent between the two options?



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# At what probability ( $p_B$ ) of rain for the beach are you indifferent between the two options?

- Earlier, we solved for the expected payoff of remaining at home: 0.74 (which was a lower expected value than going to the beach when the chance of rain at both was 30%)
- **What would  $p_B$  need to be to yield an expected payoff *at the beach* of 0.74?**
  - In other words, at what probability of rain at the beach would you be indifferent between staying at home & going to the beach?

**At what probability ( $p_B$ ) of rain for the beach are you indifferent between the two options?**

Set 0.74 (expected value of remaining at home) equal to the **beach** payoffs and solve for  $p_B$

$$p_B * 0.4 + (1 - p_B) * 1.0 = 0.74$$

**At what probability ( $p_B$ ) of rain for the beach are you indifferent between the two options?**

$$p_B * 0.4 + (1 - p_B) * 1.0 = 0.74$$

$$p_B * 0.4 + 1 - p_B = 0.74$$

**At what probability ( $p_B$ ) of rain for the beach are you indifferent between the two options?**

$$p_B * 0.4 + 1 - p_B = 0.74$$

$$p_B * -0.6 = -0.26$$

At what probability ( $p_B$ ) of rain for the beach are you indifferent between the two options?

$$p_B * -0.6 = -0.26$$

$$p_B = -0.26 / -0.6 = \mathbf{0.43}$$

# At what probability ( $p_B$ ) of rain for the beach are you indifferent between the two options?

When the probability of rain at the beach is 43% (probability of rain at home remains at 30%), we would be indifferent between staying at home & going to the beach.

If the probability of rain at the beach is  $> 43\%$ , then we would stay home

# Probability Review

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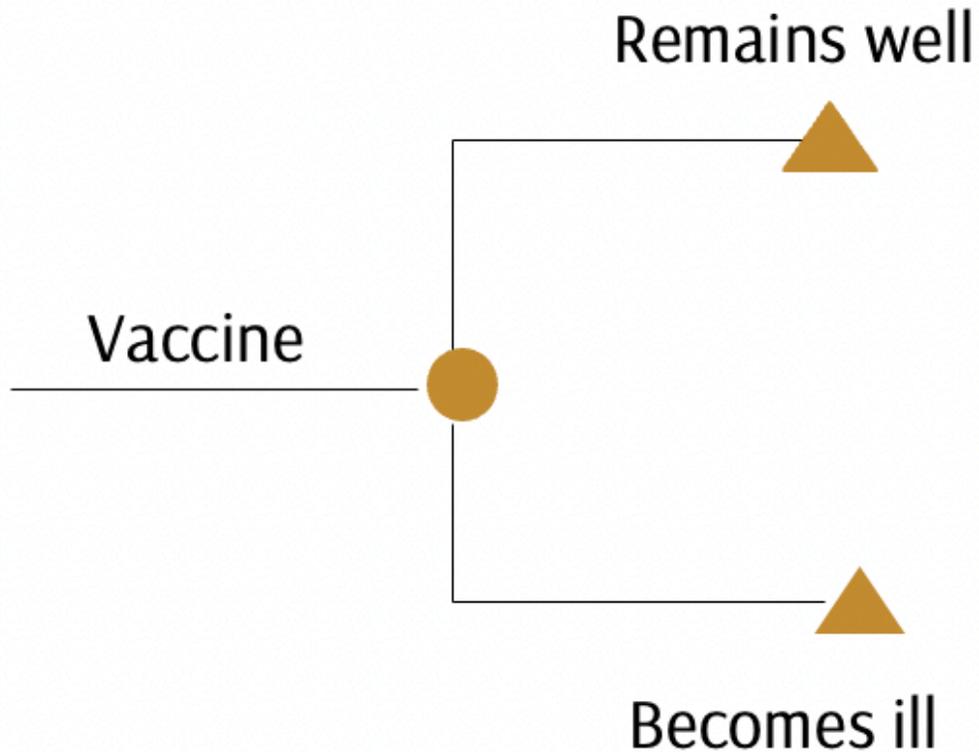
# Probabilities within Decision Trees

## Mutually exclusive events

- 2 things that cannot occur together (one event cannot occur at the same time as the other event)
  - Example: 2 events, survive or die; mutually exclusive because a person cannot be both at the same time
  - Example: 2 events, cured or not cured

# Probabilities within Decision Trees

## Mutually exclusive events



# Probabilities within Decision Trees

## Mutually exclusive events

- Assuming events are mutually exclusive, then the probability of 2 events occurring is the **sum of the probability of each event occurring individually**

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

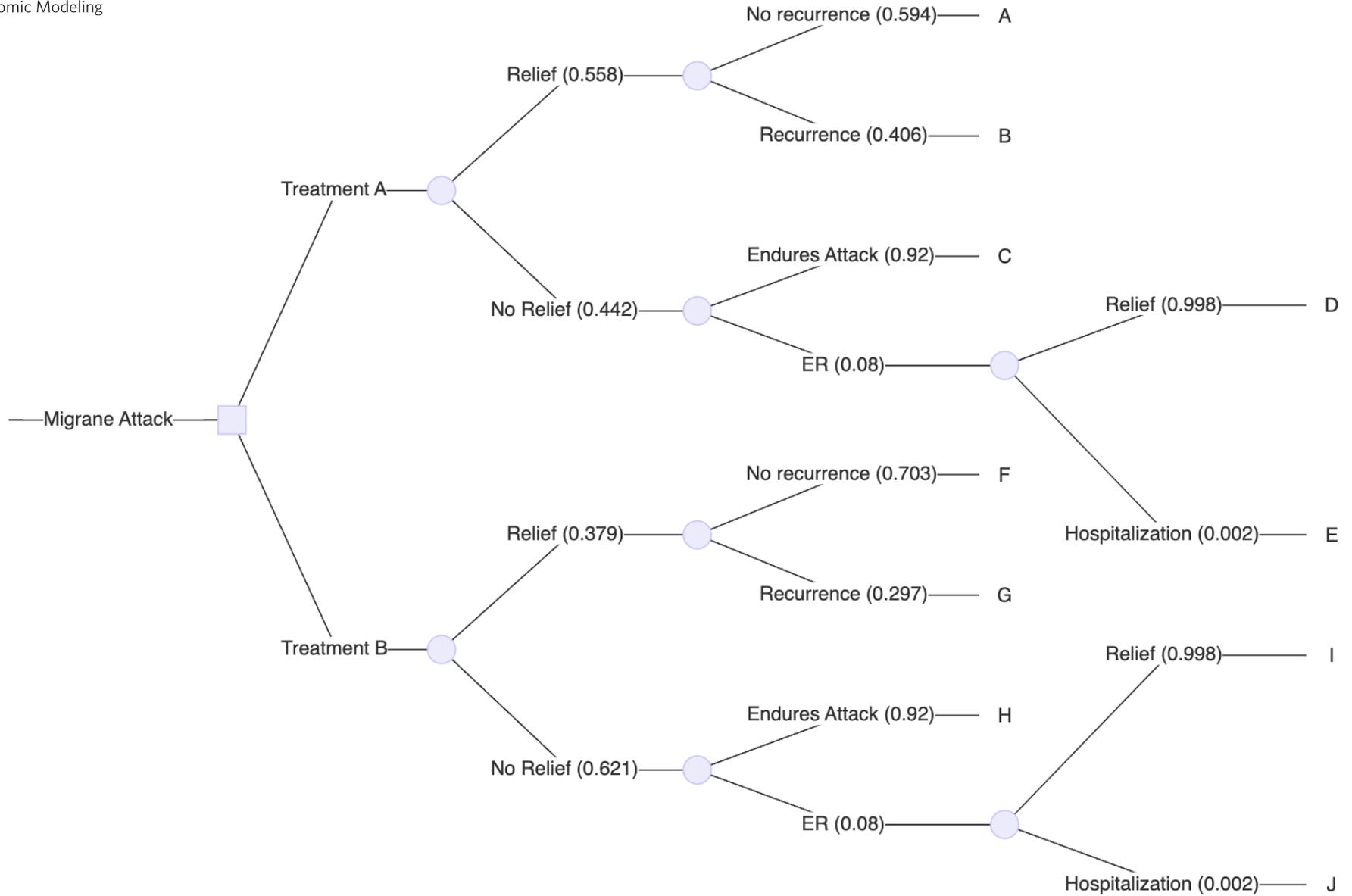
# Probabilities within Decision Trees

## Joint probability

$P(A \text{ and } B)$ : The probability of two events occurring at the same time.

## Conditional probability

$P(A|B)$ : The probability of an event A given that an event B is known to have occurred.



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# Pathways

- Pathways = sequence of events that lead to a subsequent “pay off”
- In other words, a sequence of events leads to an outcome/consequence, or payoff
- Example below: Our beach example has 4 pathways.



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# Conditional probability

Example: What is the conditional probability of death within a year of birth, given the infant has a mother who smokes?

(Probability of an event occurring (B) given that another event occurred (A))

$$P(A|B) = P(A \text{ and } B) / P(B)$$

# Conditional probability, 2x2 review

What is the conditional probability of death within a year of birth, given the infant has a mother who smokes?

	Mother's smoking status		
Infant mortality	Smoker	Non-smoker	Total
Death	16,712	18,784	35,496
Live at 1-year	1,197,142	2,878,421	4,075,563
Total	1,213,854	2,897,205	4,111,059

*\*A= death in first year; B=mother who smokes*

$$P(A|B) = 16,712 / (1,197,142 + 16,712)$$

$$= 14 \text{ per } 1,000 \text{ births}$$

# Conditional probability, 2x2 review

Or, if we wanted to use the conditional probability equation

	Mother's smoking status		
Infant mortality	Smoker	Non-smoker	Total
Death	16,712	18,784	35,496
Live at 1-year	1,197,142	2,878,421	4,075,563
Total	1,213,854	2,897,205	4,111,059

$$P(A|B) = P(A \text{ and } B)/P(B)$$

*\*A= death in first year; B=mother who smokes*

$$P(A \text{ and } B) = 16,712 / 4,111,059 = 0.0041$$

$$P(B) = 1,213,854/4,111,059 = 0.295$$

$$P(A|B) = 0.0041/0.295 = 0.014$$

# Conditional probability, 2x2 review

Probability of  $A|B$  is different from that of  $B|A$

	Birthweight		
Infant mortality	Low birthweight	Normal birthweight	Total
Death	21,054	14,442	35,496
Live at 1-year	271,269	3,804,294	4,075,563
Total	292,323	3,818,736	4,111,059

If  $A$  = death in first year;  $B$ =normal birth weight infant, then the conditional probability of  $P(A|B)$  = the probability of an infant death, given that the child has a normal birth weight

# Conditional probability, 2x2 review

What is the conditional probability of an infant death, given that the child has a normal birth weight

	Birthweight		
Infant mortality	Low birthweight	Normal birthweight	Total
Death	21,054	14,442	35,496
Live at 1-year	271,269	3,804,294	4,075,563
Total	292,323	3,818,736	4,111,059

If A = death in first year; B=normal birth weight infant

$$P(A|B) = 14,442 / (14,442 + 3,804,294)$$

$$= 3.8 \text{ deaths per } 1,000 \text{ births}$$

# Conditional probability, 2x2 review

Or, if we wanted to use equation

	Birthweight		
Infant mortality	Low birthweight	Normal birthweight	Total
Death	21,054	14,442	35,496
Live at 1-year	271,269	3,804,294	4,075,563
Total	292,323	3,818,736	4,111,059

$$P(A|B) = P(A \text{ and } B)/P(B)$$

$$P(A \text{ and } B) = 14,442$$

$$P(B) = 3,818,736$$

$$P(A|B) = 14,442/3,818,736 = 0.0038$$

# Conditional probability, 2x2 review

On the other hand, the conditional probability of  $P(B|A)$  is the probability that an infant had normal birth weight, given that the infant died within 1 year from birth

	Birthweight		
Infant mortality	Low birthweight	Normal birthweight	Total
Death	21,054	14,442	35,496
Live at 1-year	271,269	3,804,294	4,075,563
Total	292,323	3,818,736	4,111,059

*\*B=normal birth weight infant; A = death in first year*

# Conditional probability, 2x2 review

Solve  $P(B|A)$  – the probability that an infant had normal birthweight, given that the infant died within 1 year from birth

	Birthweight		
Infant mortality	Low birthweight	Normal birthweight	Total
Death	21,054	14,442	35,496
Live at 1-year	271,269	3,804,294	4,075,563
Total	292,323	3,818,736	4,111,059

$$P(B|A) = 14,442 / (14,442 + 21,054) = 0.41$$

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# Other Concepts in Decision Analysis

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# Decision Philosophies

❗ Maximizing expected value is a reasonable criterion for choice given uncertain prospects; though it does not necessarily promise the best results for any one individual.

- Mini-max regret
- Maxi-max
- Expected utility

# Decision Philosophies

⚠ Maximizing expected value is a reasonable criterion for choice given uncertain prospects; though it does not necessarily promise the best results for any one individual.

- Mini-max regret
  - Never go to the beach unless 0% rain.
- Maxi-max gain
- Expected utility

# Decision Philosophies

⚠ Maximizing expected value is a reasonable criterion for choice given uncertain prospects; though it does not necessarily promise the best results for any one individual.

- Mini-max regret
  - Never go to the beach unless 0% rain.
- Maxi-max
  - Always go to the beach unless 100% rain.
- Expected utility

# Decision Philosophies

❗ Maximizing expected value is a reasonable criterion for choice given uncertain prospects; though it does not necessarily promise the best results for any one individual.

- Mini-max regret
  - Never go to the beach unless 0% rain.
- Maxi-max
  - Always go to the beach unless 100% rain.
- Expected utility
  - Depends on the weather.

# Payoffs

- Each state of the world is assigned a cost or outcome.
- Our goal is often to calculate the expected value of these payoffs.
- We'll cover more on the theories and frameworks underlying various payoffs in the next few lectures

# Strengths/limitations of decision trees

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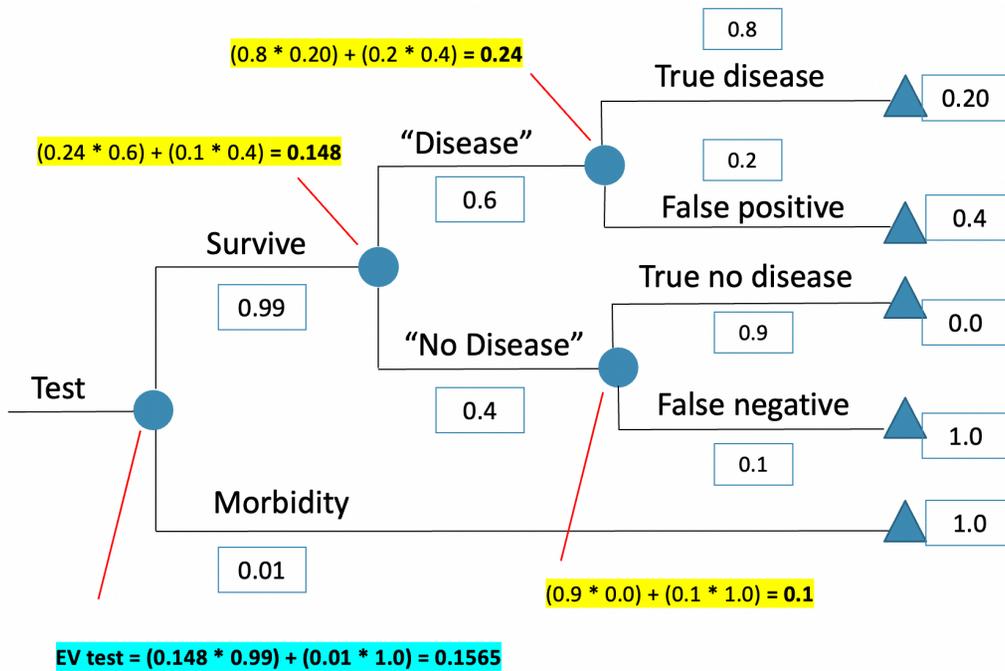
# Strengths

- They are easy to describe and understand
- Works well with limited time horizon
- Decision trees are a powerful framework for analyzing decisions and can provide rapid/useful insights, but they have limitations.

# Limitations

- No explicit accounting for the elapse of time.
  - Recurrent events must be separately built into model.
  - Fine for short time cycles (e.g., 12 months) but we often want to model over a lifetime.
- Difficult to incorporate real clinical detail - Tree structure can quickly become complex.

# Preview of what's to come



- More complex tree!