

GSOE9210 Engineering Decisions

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Updating belief

- 1 Bayesian updating
 - Airline case study
- 2 Value of information
- 3 Revision of Bayesian beliefs
 - Incorporating additional information
 - Updating reliability likelihood
- 4 Sensitivity analysis

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Outline

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Case study: capital purchase



Example (To purchase or not)

You're the chief engineer of a small commercial airline which, due to increased demand, is considering adding to its fleet by buying (B) a used airliner. Another company is offering to sell one of its airliners for \$400,000. Used airlines range in reliability, which is hard to evaluate without a detailed inspection.

Question: should you purchase?

Problem modelling

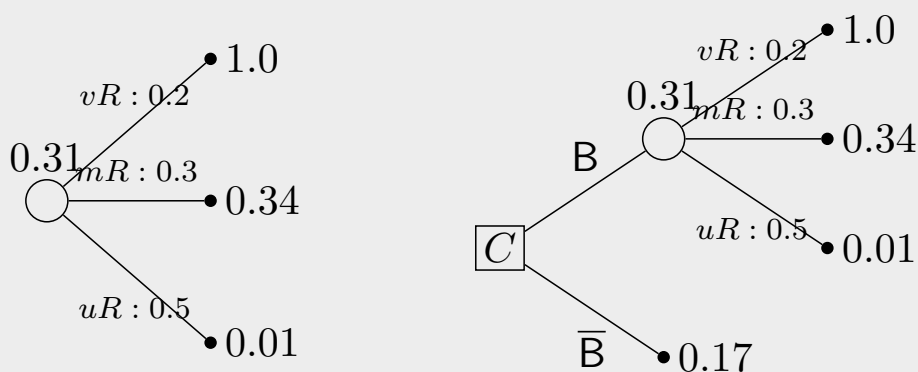
- Problem 1: how to measure reliability? Operating hours
 - Simplification 1: classify airliners as either: very reliable (vR) ($> 90\%$), moderately reliable (mR), or unreliable (uR) ($< 50\%$)
- Beliefs about reliability:

| | Reliability | | |
|-------------|-------------|------|------|
| | vR | mR | uR |
| Probability | 0.2 | 0.3 | 0.5 |
| Utility | 1.0 | 0.34 | 0.01 |

- Simplification 2: assume a very reliable airliner makes \$1M profit (best outcome); an unreliable one makes \$200K loss (worst)
- Simplification 3: utility of not buying airliner—*status quo*: 0.17

Decision C (buy or not)

Buy:



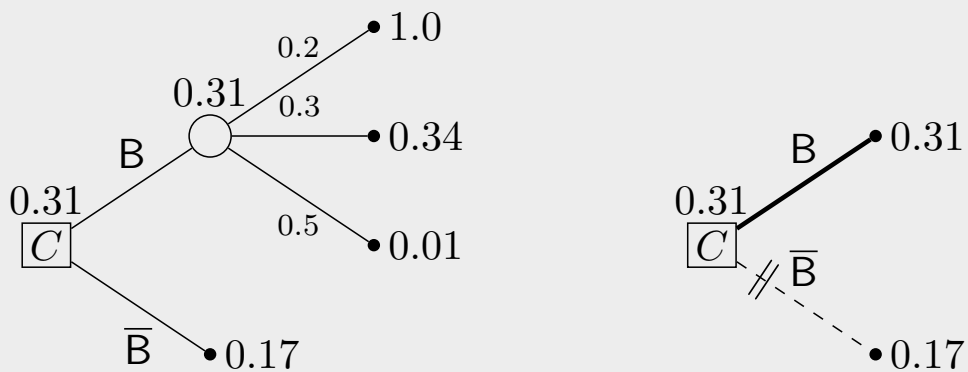
$$U(B) = 0.2(1.0) + 0.3(0.34) + 0.5(0.01) = 0.31$$

$$U(\bar{B}) = 0.17$$

| | 0.2 | 0.3 | 0.5 | U |
|-----------|------|------|------|------|
| | vR | mR | uR | |
| B | 1.00 | 0.34 | 0.01 | 0.31 |
| \bar{B} | 0.17 | 0.17 | 0.17 | 0.17 |

Decision C

- Evaluate decision points/nodes by maximal utility of alternatives (*i.e.*, actions/strategies)
- The value of node C is 0.31, because $0.31 > 0.17$; *i.e.*, $0.31 = \max\{0.17, 0.31\}$



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Get more information?

Example (Additional information)

You have the option to consult an aeronautical engineering firm to conduct an assessment of the airliner for \$10K. The report's will be either favourable (f) or unfavourable (u) as to whether or not to purchase.

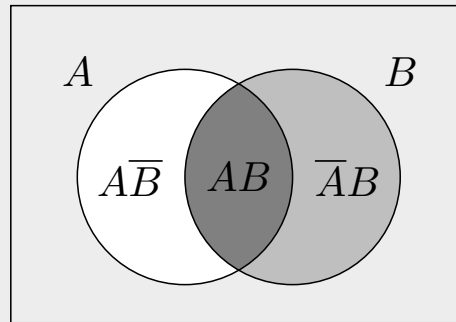
- Firm's assessment reliable?
- Guess/estimate that 90% of very reliable planes receive favourable assessment; *i.e.*, $P(f|vR) = 0.9$

| Probability of: | ... conditional on: | | |
|-----------------|---------------------|------|------|
| | vR | mR | uR |
| f | 0.9 | 0.6 | 0.1 |
| u | 0.1 | 0.4 | 0.9 |

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Conditional probability



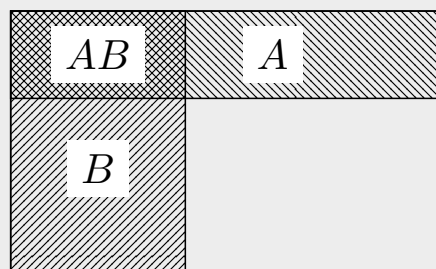
Definition

The *conditional probability* of event A conditional on B (provided B is possible; i.e., $P(B) \neq 0$), written $P(A|B)$, is defined by:

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

In the diagram above, $P(A|B)$ represents the ratio of (the area of) the region AB (the dark region) to that of the whole of B .

Conditional independence



Definition

Event A is (*conditionally*) *independent* of event B if:

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

Event A is (*conditionally*) *dependent* on B if A is not (conditionally) independent of B .

For example, if B is a random sample of a population.

Bayes's rule

- Rearranging the definition of conditional probability:

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

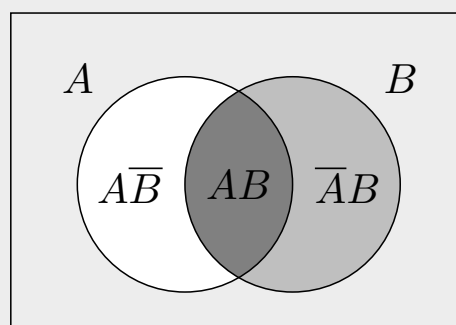
- By symmetry $P(A \cap B) = P(B \cap A)$;
therefore: $P(A|B)P(B) = P(B|A)P(A)$.
Rearranging gives:

Theorem (Bayes's Theorem I)

If A and B are any two events (such that $P(A) \neq 0$), then:

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

Bayes's Venn diagram



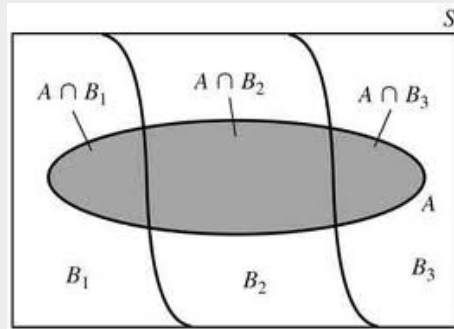
But $A = AB \cup A\bar{B}$. So we get the following:

Theorem (Bayes's Theorem I')

If A and B are any two events ($P(A) \neq 0$), then:

$$P(B|A) = \frac{P(AB)}{P(AB) + P(A\bar{B})}$$

Extending Bayes's rule



$$P(B_1|A) = \frac{P(AB_1)}{P(AB_1) + P(AB_2) + P(AB_3)}$$

$$P(B_2|A) = \frac{P(AB_2)}{P(AB_1) + P(AB_2) + P(AB_3)}$$

$$P(B_3|A) = \frac{P(AB_3)}{P(AB_1) + P(AB_2) + P(AB_3)}$$

Bayes's rule generalised

Events B_1, \dots, B_n are said to be *universally exhaustive* (of Ω) if $\bigcup_{i=1}^n B_i = \Omega$.

Theorem (Bayes's Theorem II)

If events $B_1, \dots, B_k, \dots, B_n$ are mutually exclusive and universally exhaustive, and A is a possible event ($P(A) \neq 0$), then:

$$P(B_k|A) = \frac{P(AB_k)}{\sum_{i=1}^n P(AB_i)}$$

Theorem (Bayes's Theorem II')

If $B_1, \dots, B_k, \dots, B_n$ are mutually exclusive and universally exhaustive events and A is a possible event ($P(A) \neq 0$) then:

$$P(B_k|A) = \frac{P(A|B_k)P(B_k)}{\sum_{i=1}^n P(A|B_i)P(B_i)}$$

Example: Bayes's rule

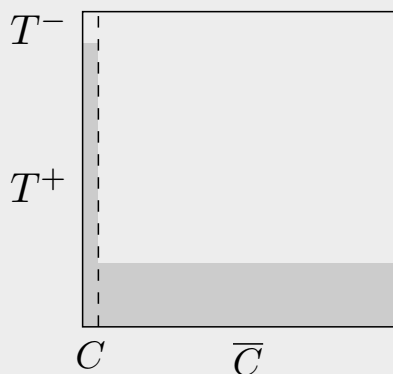
Example (Medical diagnostics)

In a given population of people, one in every thousand have hypo-cytocitic cancer. A certain pathology test is used to detect the disease. The test is 'good' but not perfect; it returns a positive result in 98% of persons with the disease, and registers a *false positive* (i.e., gives a positive result for a person free of the disease) 5% of the time.

Exercise

A random person comes in to get tested and the test returns a positive result. What is the probability that the person has cancer?

Example: solution



Given information:

$$\begin{aligned} P(C) &= \frac{1}{1000} & P(\bar{C}) &= \frac{999}{1000} \\ P(T^+|C) &= \frac{98}{100} & P(\bar{T}^+|C) &= \frac{2}{100} \\ P(T^+|\bar{C}) &= \frac{5}{100} & P(\bar{T}^+|\bar{C}) &= \frac{95}{100} \end{aligned}$$

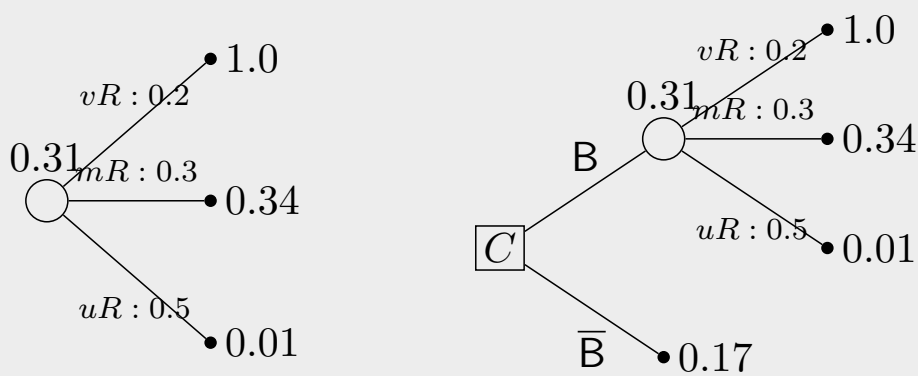
What is $P(C|T^+)$?

$$\begin{aligned} P(C|T^+) &= \frac{P(CT^+)}{P(T^+)} = \frac{P(CT^+)}{P(CT^+ \cup \bar{C}T^+)} = \frac{P(CT^+)}{P(CT^+) + P(\bar{C}T^+)} \\ &= \frac{P(T^+|C)P(C)}{P(T^+|C)P(C) + P(T^+|\bar{C})P(\bar{C})} \\ &= \frac{\frac{98}{100} \times \frac{1}{1000}}{\frac{98}{100} \times \frac{1}{1000} + \frac{5}{100} \times \frac{999}{1000}} = \frac{98}{98 + 5 \times 999} \approx \frac{100}{5000} = 0.02 \end{aligned}$$

Patient only has 2% chance of having cancer despite testing positive?!

Decision C (buy or not)

Buy:



$$U(B) = 0.2(1.0) + 0.3(0.34) + 0.5(0.01) = 0.31$$

$$U(\bar{B}) = 0.17$$

| | 0.2 | 0.3 | 0.5 | |
|-----------|------|------|------|------|
| | vR | mR | uR | U |
| B | 1.00 | 0.34 | 0.01 | 0.31 |
| \bar{B} | 0.17 | 0.17 | 0.17 | 0.17 |

Post-report (posterior) probabilities

- If report favourable (f):

$$\begin{aligned}
 P(vR|f) &= \frac{P(f|vR)P(vR)}{P(f|vR)P(vR) + P(f|mR)P(mR) + P(f|uR)P(uR)} \\
 &= \frac{0.9(0.2)}{0.9(0.2) + 0.6(0.3) + 0.1(0.5)} \\
 &= \frac{0.18}{0.41} \approx 0.44
 \end{aligned}$$

Similarly: $P(mR|f) \approx 0.44$ and $P(uR|f) \approx 0.12$

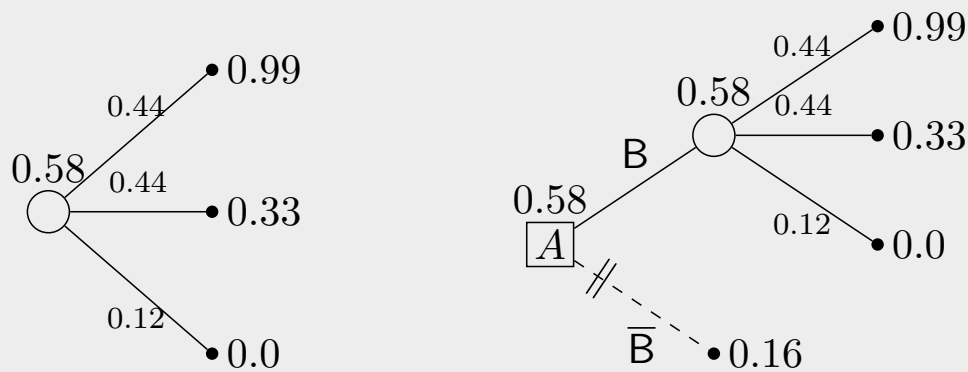
- If report unfavourable (u):

$$P(vR|u) = \frac{0.02}{0.59} \approx 0.04$$

$$P(mR|u) \approx 0.20$$

$$P(uR|u) \approx 0.76$$

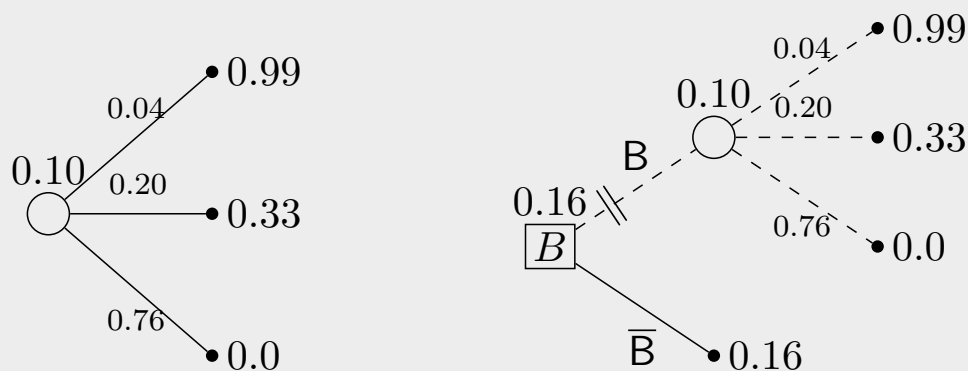
Decision A (report favourable)



- The revised expected utility of buying the airliner is $U(B) = 0.44(0.99) + 0.44(0.33) + 0.12(0.0) = 0.58$
- The utility of not buying it is $U(\bar{B}) = 0.16$.

| | 0.44 | 0.44 | 0.12 | |
|-----------|------|------|------|------|
| | vR | mR | uR | U |
| B | 0.99 | 0.33 | 0.0 | 0.58 |
| \bar{B} | 0.16 | 0.16 | 0.16 | 0.16 |

Decision B (report unfavourable)



- The revised expected utility of buying the airliner is $U(B) = 0.04(0.99) + 0.20(0.33) + 0.76(0.0) = 0.10$
- The utility of not buying it is $U(\bar{B}) = 0.16$.

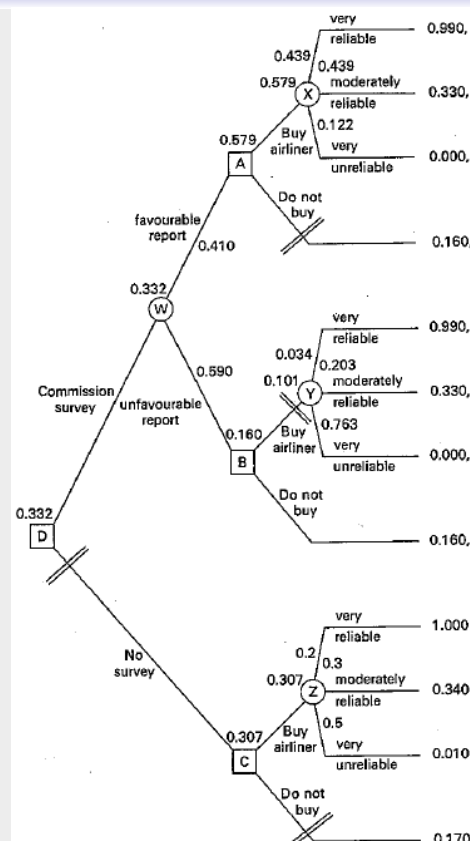
| | 0.04 | 0.20 | 0.76 | |
|-----------|------|------|------|------|
| | vR | mR | uR | U |
| B | 0.99 | 0.33 | 0.0 | 0.10 |
| \bar{B} | 0.16 | 0.16 | 0.16 | 0.16 |

Utility adjustments

- Problem: cost associated with report?
Question: How does report's cost (\$10K) affect utility?
- Observation: report cost small relative to other monetary quantities: potential profit \$1M; *i.e.*, $\$10K \ll \$1M$
- Simplification 3: model effect by constant shift; *i.e.*, for report costing \$ x ($x \ll 1M$), change of utility is $\Delta u = \frac{x}{1M}$; \$1M
- That is, every \$10K is worth 0.01 *utils*

Combined decision

- Combine all three possible cases into one big decision problem
- Introduce new decision: commission survey/report, and no survey/report
- Introduce new event: report outcome (f or u)
- If consultant good, report likely to be good predictor of (*i.e.*, correlated to) aircraft reliability
- Consultant's increased predictive accuracy is *valuable* in making decision



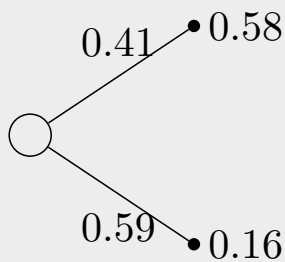
Combined decision

- From the denominators in the earlier calculations:

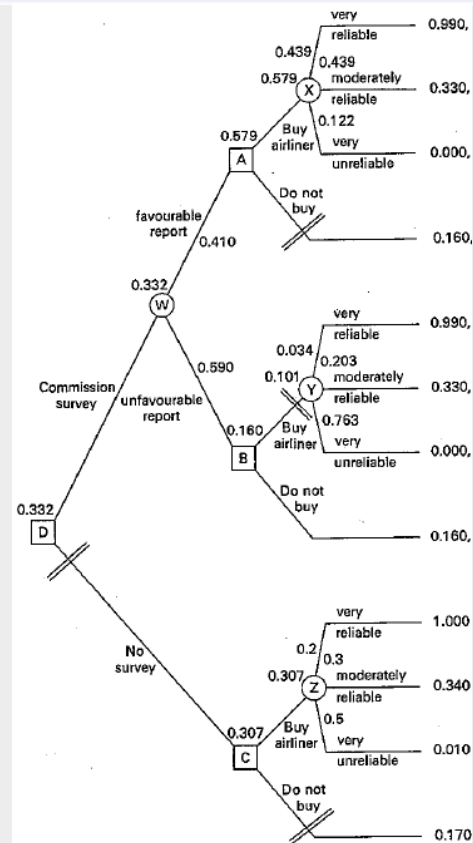
$$P(f) = 0.41$$

$$P(u) = 0.59$$

- Therefore, if report commissioned:



- Utility of report: 0.33



Decision table

| | f, vR | f, mR | f, uR | u, vR | u, mR | u, uR | U |
|----------|----------|----------|---------|---------|---------|---------|------|
| A_1 | 1.0 | 0.34 | 0.01 | 1.0 | 0.34 | 0.01 | 0.31 |
| A_2 | 0.17 | 0.17 | 0.17 | 0.17 | 0.17 | 0.17 | 0.17 |
| A_3 | 0.99 | 0.33 | 0 | 0.99 | 0.33 | 0 | |
| \vdots | \vdots | \vdots | | \dots | | | |
| A_6 | | | | | | | |

where

A_1 no survey; buy airliner

A_2 no survey; don't buy airliner

A_3 commission survey; buy airliner

A_4 commission survey; don't buy

A_5 commission survey; if favourable, buy airliner; else don't buy

A_6 commission survey; if favourable, don't buy airliner; else buy

Value of information

- Optimal policy if report commissioned:

Policy for C : report commissioned

If report favourable, buy airliner, if not don't buy.

- Value of policy is $U(C) = 0.33$, inclusive of the 0.01 fee
- Optimal policy if report not commissioned:

Policy for \bar{C} : report not commissioned

Buy the airliner.

- $U(\bar{C}) = 0.31$
- How much is report worth?
- $U(C) = 0.34 - u_r \geq 0.31 = U(\bar{C})$; i.e., should commission report for fee up to $u_r = 0.03$; i.e., for any fee up to \$30K

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Production and demand

Example (Production)

Alice is the CTO at a company and Bob is the CFO. They're considering two possible production processes for a product. Process A is expected to net \$40K if demand increases, \$30K if demand remains stable, and \$20K if demand falls. Process B requires a greater initial capital expenditure; it will only net \$10K if demand drops, and \$40K otherwise.

Future estimates of demand are: 20% of an increase, 30% chance of staying level, and 50% of a decrease.

Which process should Alice implement?

Example

The decision table is:

| | $\frac{5}{10}$ ↓ | $\frac{3}{10}$ — | $\frac{2}{10}$ ↑ | $V_{\$}$ |
|---|---------------------|---------------------|---------------------|----------|
| A | \$20 | \$30 | \$40 | \$27 |
| B | \$10 | \$40 | \$40 | \$25 |

$$V_{\$}(A) = \frac{5}{10}(20) + \frac{3}{10}(30) + \frac{2}{10}(40)$$

$$= 10 + 9 + 8 = \$27$$

$$V_{\$}(B) = \frac{5}{10}(10) + \frac{3}{10}(40) + \frac{2}{10}(40)$$

$$= 5 + 12 + 8 = \$25$$

Alternative A has greater expected monetary value

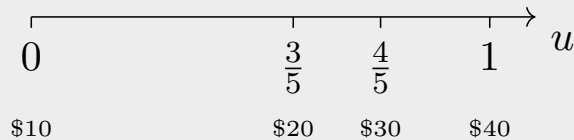
Example

Alice consults Bob who advises her that, under its current financial position, the company's preferences are:

$$\$20 \sim \left[\frac{3}{5} : \$40 \mid \frac{2}{5} : \$10 \right]$$

$$\$30 \sim \left[\frac{4}{5} : \$40 \mid \frac{1}{5} : \$10 \right]$$

The company's utility for money is:



The utility table:

| | $\frac{5}{10}$ ↓ | $\frac{3}{10}$ — | $\frac{2}{10}$ ↑ | U |
|---|---------------------|---------------------|---------------------|------------------|
| A | $\frac{3}{5}$ | $\frac{4}{5}$ | 1 | $\frac{74}{100}$ |
| B | 0 | 1 | 1 | $\frac{50}{100}$ |

$$\begin{aligned} U(A) &= \frac{5}{10} \left(\frac{3}{5} \right) + \frac{3}{10} \left(\frac{4}{5} \right) + \frac{2}{10} (1) \\ &= \frac{1}{50} (15 + 12 + 10) = \frac{74}{100} \end{aligned}$$

$$\begin{aligned} U(B) &= \frac{5}{10} (0) + \frac{3}{10} (1) + \frac{2}{10} (1) \\ &= \frac{1}{50} (0 + 15 + 10) = \frac{50}{100} \end{aligned}$$

Therefore, A will also have greater utility

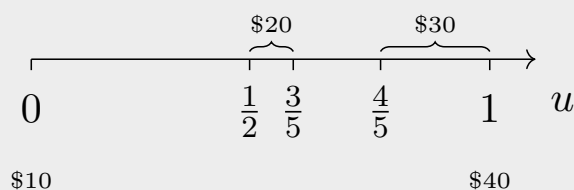
Sensitivity analysis

Suppose Bob cannot give precise assessments on values of \$20 and \$30, only bounds:

$$\left[\frac{3}{5} \$40 \right] \succ \$20 \succ \left[\frac{1}{2} \$40 \right]$$

$$\$40 \succ \$30 \succ \left[\frac{4}{5} \$40 \right]$$

The utility for money is:



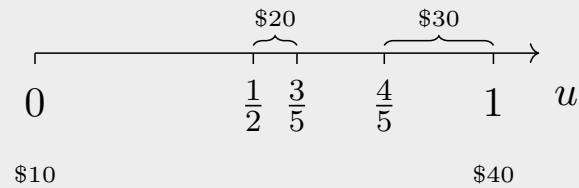
Lower bound for A:

| | $\frac{5}{10}$ ↓ | $\frac{3}{10}$ — | $\frac{2}{10}$ ↑ | U |
|---|---------------------|---------------------|---------------------|------------------|
| A | $\frac{1}{2}$ | $\frac{4}{5}$ | 1 | $\frac{69}{100}$ |
| B | 0 | 1 | 1 | $\frac{50}{100}$ |

Upper bound for A:

| | $\frac{5}{10}$ ↓ | $\frac{3}{10}$ — | $\frac{2}{10}$ ↑ | U |
|---|---------------------|---------------------|---------------------|------------------|
| A | $\frac{3}{5}$ | 1 | 1 | $\frac{80}{100}$ |
| B | 0 | 1 | 1 | $\frac{50}{100}$ |

Sensitivity analysis



Bounds on A:

$$\begin{aligned}
 U(A) &> \frac{5}{10} \left(\frac{1}{2} \right) + \frac{3}{10} \left(\frac{4}{5} \right) + \frac{2}{10} (1) \\
 &= \frac{1}{100} (25 + 24 + 20) \\
 &= \frac{69}{100} \\
 U(A) &< \frac{5}{10} \left(\frac{3}{5} \right) + \frac{3}{10} (1) + \frac{2}{10} (1) \\
 &= \frac{1}{100} (30 + 30 + 20) \\
 &= \frac{80}{100}
 \end{aligned}$$

That is:

$$\frac{69}{100} < U(A) < \frac{80}{100}$$

Conclusion:

A is guaranteed to be preferred to B ($U(B) = \frac{50}{100}$) regardless of the uncertainty over the precise preference for \$20 and \$30.

Summary

- Explored decision problems in greater depth:
 - actions that affect epistemic state (value of information-gathering actions)
 - dealing with uncertainty in preferences (sensitivity analysis)
- Updating beliefs (epistemic state) via *Bayes's* theorem
- Value of information: cost of gathering more information versus increase in expected utility due to new information
- Sensitivity analysis:
 - decisions under imprecise preferences
 - how might preference uncertainty affect a decision?