

# GSOE9210 Engineering Decisions

Victor Jauregui

`vicj@cse.unsw.edu.au`  
`www.cse.unsw.edu.au/~gs9210`

## Decisions under certainty and ignorance

- 1 Decision problem classes
- 2 Decisions under certainty
- 3 Decisions under ignorance

# Outline

- 1 Decision problem classes
- 2 Decisions under certainty
- 3 Decisions under ignorance

## Decision problem classes

Decision problems can be classified based on an agent's epistemic state:

- Decisions under *certainty*: the agent knows the actual state
- Decisions under *uncertainty*:
  - Decisions under *ignorance* (full uncertainty): the agent believes multiple states/outcomes are possible; likelihoods unknown
  - Decisions under *risk*: the agent believes multiple states/outcomes are possible; likelihood information available

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## Decisions under certainty

### Example (Project budgeting)

You are a lead software engineer in a major software company. Your R & D team has proposed three possible projects, A, B, and C, each with a different life-time. The net profits over the life of the projects are listed in the adjacent table.

	profit (\$M)
A	20
B	13
C	17

- Which project would you choose?

## Complex outcomes

Project life-time cash-flows:

- A: three years, big initial set-up costs
- B: one year immediate return
- C: three years, small initial set-up costs

	cashflow (\$M)		
	Year		
	1	2	3
A	−10	5	25
B	13	0	0
C	−5	10	12

New perspective:

- Outcomes described by vectors: e.g., for A:  $(-10, 5, 25)$ .
- What is more important: maximising total return, preserving cash, etc.?
- Which project would you choose?

## Composite outcomes: Net Present Value (NPV)

- The *Net Present Value* (NPV): value of the project in *present* terms
- Future worth less than in the present
- Model by a *discount rate* ( $\gamma$ ); assume discount rate of 20%

$$NPV(A) = -10 + \frac{5}{1.2} + \frac{25}{1.2^2} \approx 11.5$$

$$NPV(B) \approx 13.0$$

$$NPV(C) = -5 + \frac{10}{1.2} + \frac{12}{1.2^2} \approx 11.7$$

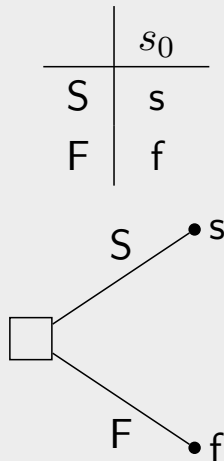
- More generally:

$$v(x_1, x_2, x_3) = x_1 + \frac{x_2}{1 + \gamma} + \frac{x_3}{(1 + \gamma)^2}$$

# Decisions under certainty

## Example (School fund-raising)

A school committee is looking to hold a fund-raiser. It has a choice between holding a fête or a sports day.

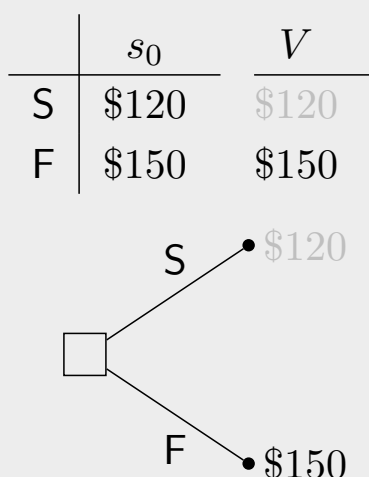


- In this example:  
 $\mathcal{A} = \{S, F\}$   
 $\Omega = \{s, f\}$   
 $\mathcal{S} = \{s_0\}$
- Which action preferred: F or S?
- Which outcome preferred: f or s?

# Decisions under certainty: value functions

## Example (School fund-raising)

A school committee is looking to hold a fund-raiser. It has a choice between holding a fête or a sports day. It expects to make \$150 for a fête but only \$120 for a sports day.



- Value function over *outcomes*:  
 $v : \Omega \rightarrow \mathbb{R}$
- In this example:  
 $v(s) = \$120$   
 $v(f) = \$150$
- Value function over *actions*; *i.e.*,  
 $V : \mathcal{A} \rightarrow \mathbb{R}$ ; here  $V(A) = v(\omega)$ , where  
 $\omega = \omega(A, s_0)$

## Rational decisions

- A *normative* theory of decision-making: *i.e.*, decisions ideal (*rational*) decision-makers *ought* to make
- Which principles govern rational decision-making?

### Rationality Principle 1 (Elimination)

Faced with two possible alternatives, rational agents should never choose the less preferred one.

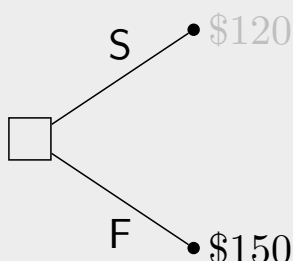
- *i.e.*, rational agents should discard less preferred actions
- Rational decision rules should satisfy this principle

## Rational decisions under certainty

### Rationality Principle 1 (Elimination)

Given a value function  $V : \mathcal{A} \rightarrow \mathbb{R}$  over actions, rational agents should prefer action A to B iff  $V(A) > V(B)$ .

	$s_0$	$V$
S	\$120	\$120
F	\$150	\$150



- Since F is preferred to S ( $V(F) > V(S)$ ), S is eliminated (by elimination), hence the rational choice is the remaining option: F

### Corollary

A rational agent should not choose any action which is not preference maximal.

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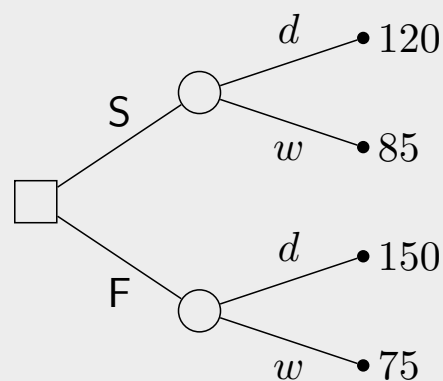
## Decisions under uncertainty

### Example (Uncertain school fund-raising)

Proceeds of the school fund-raiser depend on the weather; on a dry day ( $d$ ) the school expects to make \$150 for a fête (F) but only \$120 for a sports day (S). However, on a wet day ( $w$ ) the sports day will net \$85 and the fête only \$75.

	$d$	$w$
S	120	85
F	150	75

$d$  day is dry  
 $w$  day is wet



# Decisions under uncertainty

## Problem

How to assign values to uncertain actions?

	$d$	$w$
S	120	85
F	150	75

- How should  $V(S)$  and  $V(F)$  be defined?
- How should  $V$  (over actions) depend on  $v$  (over outcomes)?

# Lotteries

## Definition (Lottery)

A *lottery* over a finite set of states  $\mathcal{S}$ , and outcomes, or *prizes*,  $\Omega$ , is a function  $\ell : \mathcal{S} \rightarrow \Omega$ . The lottery  $\ell$  is written:

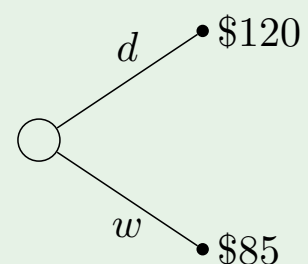
$$\ell = [s_1 : \omega_1 | s_2 : \omega_2 | \dots | s_n : \omega_n]$$

where for each  $s_i \in \mathcal{S}$ ,  $\omega_i = \ell(s_i)$ .

## Example (Dry or wet?)

The uncertain situation in which the weather on a given future day is unknown represented by the lottery:

$$\ell_S = [d : \$120 | w : \$85]$$





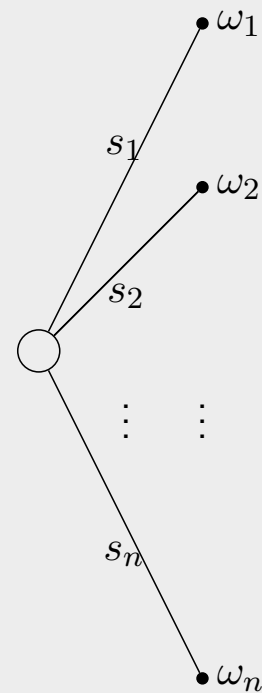
# Decision problems and lotteries

$$[s_1 : \omega_1 | s_2 : \omega_2 | \dots | s_n : \omega_n]$$

- Uncertain action = lottery
- Choosing an action = choosing a lottery

## Problem

The problem of evaluating actions amounts to the problem of determining how to compare and/or evaluate lotteries.



# Decisions under uncertainty: ignorance

## Example (Raffle)

There are four raffle tickets in a hat. Each ticket is either blue or red, but you don't know how many of each there are. Blue tickets win \$3; red ones lose (\$0). The cost of entering the raffle is \$1.

## Exercises

- Draw the decision tree and table for this problem
- Should you draw a ticket in the raffle?
- What if you knew there were three blue tickets? Four? None?
- How many blue tickets would there have to be to make it worth entering?
- If there were  $n$  blue tickets ( $0 \leq n \leq 4$ ), what would the prize have to be to make it worthwhile entering?

# Decisions under ignorance

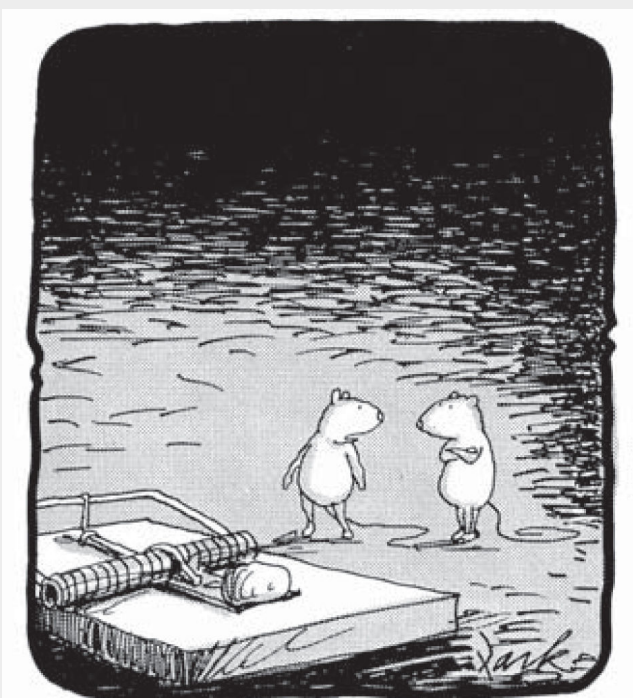
## Definition (Decision rule)

A *decision rule* is a way of choosing, for each decision problem, an action or set of actions.

Rational decision rules under ignorance:

- Optimistic *MaxiMax* rule: “nothing ventured, nothing gained”
- Wald’s pessimistic *Maximin* rule: “its better to be safe than sorry”
- *Hurwicz*’s mixed optimistic-pessimistic rule: use an *optimism index*  $\alpha$
- Savage’s *miniMax Regret* rule; least *opportunity loss*
- Laplace’s *principle of insufficient reason* rule

## MaxiMax



“Modern technology being what it is, there’s a good chance it won’t work anyway.”

*MaxiMax* associates with each action the state which yields the most preferred outcome (*i.e.*, preference maximal)

The *MaxiMax* decision rule selects the action(s) which yield a preference-maximal outcome among these

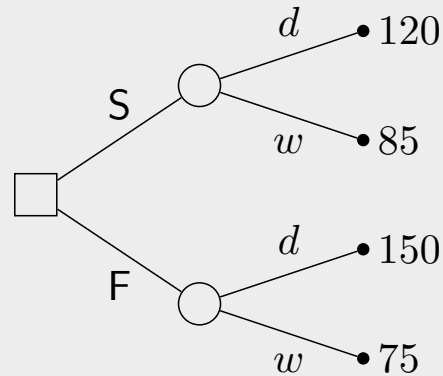
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	$d$	$w$
S	120	85
F	150	75

$d$  day is dry  
 $w$  day is wet



## MaxiMax (MM): aim for the best

	$s_1$	$s_2$	$s_3$	$V$
$A_1$	6	0	4	6
$A_2$	2	5	1	5
$A_3$	4	3	2	4

- For each action find best possible outcome over all possible states; *i.e.*, for each row find the maximum value:  

$$V_{MM}(A) = M(A) = \max\{v(\omega(A, s)) \mid s \in \mathcal{S}\}$$
- Choose the actions/rows with maximal value:  $A_1$
- Equivalently: find the maximum value of the entire table, choose the row/action with this value
- $r_{MM}(\omega) = \arg \max\{M(A) \mid A \in \mathcal{A}\}$

## MaxiMax

	$s_1$	$s_2$
A	10	0
B	9	9

- Which action is better?
- How could ties be broken?

	$s_1$	$s_2$	$s_3$	$V$
$A_1$	6	0	4	6
$A_2$	2	6	1	6
$A_3$	4	3	2	4

## Risk attitudes

- **MaxiMax** is a decision rules for extreme *risk takers*
- Some agents may *prefer* risks if the favourable outcomes are sufficiently desirable
- *MaxiMax* would be a rational decision rule decision-makers with risk-taking attitudes/preferences
- In many cases it is wise to be *risk averse* (dislike risk): *i.e.*, avoid, reduce, or protect against risk
- What might a risk averse decision rule look like?

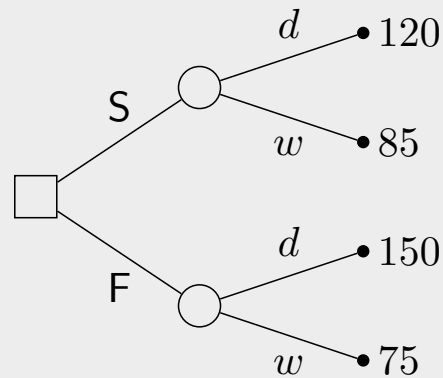
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	$d$	$w$
S	120	85
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$d$  day is dry  
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## Maximin (Mm): best in the worst case

- Assume the worst case/state will occur for each action

	$s_1$	$s_2$	$s_3$	$V$
$A_1$	6	0	4	0
$A_2$	2	5	1	1
$A_3$	4	3	2	2

- For each action find the worst possible outcome under all possible cases/states; i.e., for each row find the minimum value:  

$$V(A) = m(A) = \min\{v(\omega(A, s)) \mid s \in \mathcal{S}\}$$
- $m(A)$  is sometimes called the *security level* of action  $A$
- Choose the action/row with the maximum of these:  $A_3$
- $r_{Mm}(\omega) = \arg \max\{m(A) \mid A \in \mathcal{A}\}$

## Maximin

	$s_1$	$s_2$
A	10	0
B	1	1

- Which action is better?
- How could ties be broken?

	$s_1$	$s_2$	$s_3$	$V$
$A_1$	6	0	4	0
$A_2$	2	5	2	2
$A_3$	4	3	2	2

## Hurwicz's optimism index

	$s_1$	$s_2$	$s_3$	$M$	$m$	$\alpha M + (1 - \alpha)m$
$A_1$	6	0	4	6	0	$\frac{9}{2}$
$A_2$	2	5	1	5	1	$\frac{8}{2}$
$A_3$	4	3	2	4	2	$\frac{7}{2}$

- For each action/row, find the minimum ( $m$ ) and maximum ( $M$ ) values
- Calculate a weighted sum based on the *optimism index*  $\alpha$  (e.g.,  $\alpha = \frac{3}{4}$ ); i.e.,  $V(A) = \alpha M(A) + (1 - \alpha)m(A) = \frac{3}{4}M + \frac{1}{4}m$ .
- Choose the row/action that maximises this value:  $A_1$

### Exercise

What happens when  $\alpha = 1$ ?  $\alpha = 0$ ?

## MaxiMax and Maximin

	$s_1$	$s_2$
A	100	0
B	99	99

	$s_1$	$s_2$
A	100	4
B	5	5

	$s_1$	$s_2$
A	$a_1$	$a_2$
B	$b_1$	$b_2$

Compare problems above:

- *MaxiMax* and *Maximin* choose the same action for any values of  $a_1, a_2, b_1, b_2$ , provided  $a_1 > b_1 \geq b_2 > a_2$  is preserved; since  $M(A) = a_1 > M(B)$ ,  $m(B) > a_2 = m(A)$  remain unchanged; *i.e.*, the actual numbers are irrelevant for the rules
- In this case the differences  $a_1 - b_1$  and  $b_2 - a_2$  are irrelevant provided  $M(A) > M(B)$  and  $m(B) > m(A)$

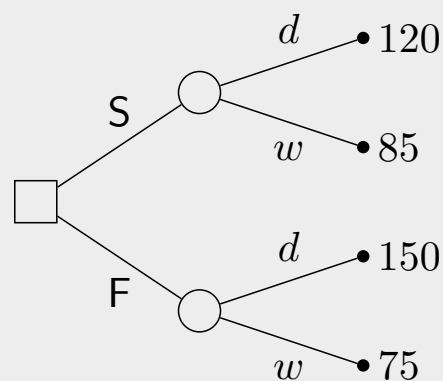
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	$d$	$w$
S	120	85
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$d$  day is dry  
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## Best response actions

### Definition (Better actions)

An action  $A$  is *strictly (weakly) better* than action  $B$  in a given state  $s$  iff  $A$ 's outcome in  $s$  strictly (weakly) preferred to  $B$ 's.

### Definition (Best response)

An action  $A$  is a *strictly (weakly) best response* in state  $s$  iff  $A$  is strictly (weakly) better than every other action in state  $s$ .

- A best response action is preference maximal over all actions in that state.

## Best response actions

	$d$	$w$
S	120	85
F	150	75

- On dry days ( $d$ ) the best response is F;  $v(\omega(F, d)) > v(\omega(S, d))$
- On wet days ( $w$ ) the best response is S;  $v(\omega(S, w)) > v(\omega(F, w))$
- Choosing an action which turns out to not be a best response action (i.e., when later discover actual state our choice of action was not a best response action) causes 'regret'
- A best response means 'no regret' in the given state ...



# Regret

## Definition (Regret)

The *regret*, or *opportunity loss*, of an outcome in a given state is the difference between the outcome's value and that of the best possible outcome for that state.

Consider the fund-raising problem discussed earlier:

	$d$	$w$		$d$	$w$	$M_R$
S	120	85	S	30	0	30
F	150	75	F	0	10	10
$M_s$	150	85				

- The maximum regret for the sports day is 30 but only 10 for the fête
- The action which minimises the maximum regret is F

## miniMax Regret

	$s_1$	$s_2$	$s_3$		$s_1$	$s_2$	$s_3$	$V$
A <sub>1</sub>	6	0	4	A <sub>1</sub>	0	5	0	5
A <sub>2</sub>	2	5	1	A <sub>2</sub>	4	0	3	4
A <sub>3</sub>	4	3	2	A <sub>3</sub>	2	2	2	2
	6	5	4					

- For each column/state  $s$ , find its maximum value ( $M_s$ )
- Construct the regret table:  $R(\omega) = M_s - v(\omega)$
- For each action/row find the maximum regret:  

$$V(A) = \max\{R(\omega(A, s)) \mid s \in \mathcal{S}\}$$
- Choose the row/action that **minimises the regret**: A<sub>3</sub>

## Laplace's insufficient reason

	$s_1$	$s_2$	$s_3$	$V$
$A_1$	6	0	4	$\frac{6}{3} + 0 + \frac{4}{3} = 3\frac{1}{3}$
$A_2$	2	5	1	$\frac{2}{3} + \frac{5}{3} + \frac{1}{3} = 2\frac{2}{3}$
$A_3$	4	3	2	$\frac{4}{3} + \frac{3}{3} + \frac{2}{3} = 3$

- Assume each state is equally likely
- For each row/action calculate the mean value:  

$$V(A) = \frac{1}{n}v(\omega(A, s_1)) + \dots + \frac{1}{n}v(\omega(A, s_n))$$
- Choose the row/action with **maximum value:  $A_1$**

### Exercise

How could you simplify this decision rule?

## Decision rules

	$s_1$	$s_2$	$s_3$	$s_4$	$V$
$A_1$	2	2	0	1	
$A_2$	1	1	1	1	
$A_3$	0	4	0	0	
$A_4$	1	3	0	0	

- For a value function  $V$  on actions, a decision rule  $r_V$  is defined by:

$$r_V(p) = \arg \max \{V(A) \mid A \in \mathcal{A}\}$$

- On the decision problem above, which rules agree (*i.e.*, choose the same actions)?

$$\begin{aligned} V_{MM}(A) &= \max \{v(\omega(A, s)) \mid s \in \mathcal{S}\} \\ V_{Mm}(A) &= \min \{v(\omega(A, s)) \mid s \in \mathcal{S}\} \\ V_{mMR}(A) &= \max \{R(A, s) \mid s \in \mathcal{S}\} \end{aligned}$$

## Summary: decisions under certainty and ignorance

- Certainty: only one possible state
- Ignorance: more than one possible state; likelihoods unknown
- Rationality Principles: elimination and best response
- Lotteries: uncertain situations
- Some decision rules under uncertainty (ignorance):
  - Optimistic *MaxiMax* rule: “nothing ventured, nothing gained”
  - Wald’s pessimistic *Maximin* rule: “its better to be safe than sorry”
  - *Hurwicz*’s mixed optimistic-pessimistic rule: use an *optimism index*  $\alpha$
  - Savage’s *miniMax Regret* rule; least *opportunity loss*
  - Laplace’s *principle of insufficient reason* rule