

# Bounded Rationality Lecture 2

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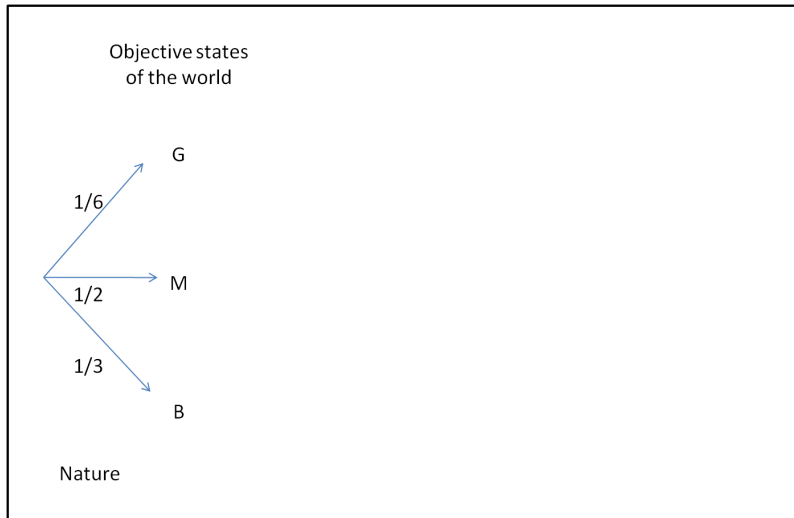
- Last week we introduced on model of costly information search/attention: Satisficing
- Examined optimal behavior with search costs
- Assumed a particular form of information search
  - Sequential Search
- Seems unnecessarily restrictive

- People choose information to acquire to maximize utility net of information costs
- People free to choose
  - *How much* information to acquire
  - *What type* of information to acquire
- Has been used to examine
  - Consumption and Savings behavior - e.g: Sims 2006
  - Price Setting - e.g. Matejka 2010, Martin 2012
  - Portfolio Choice - Modria 2010

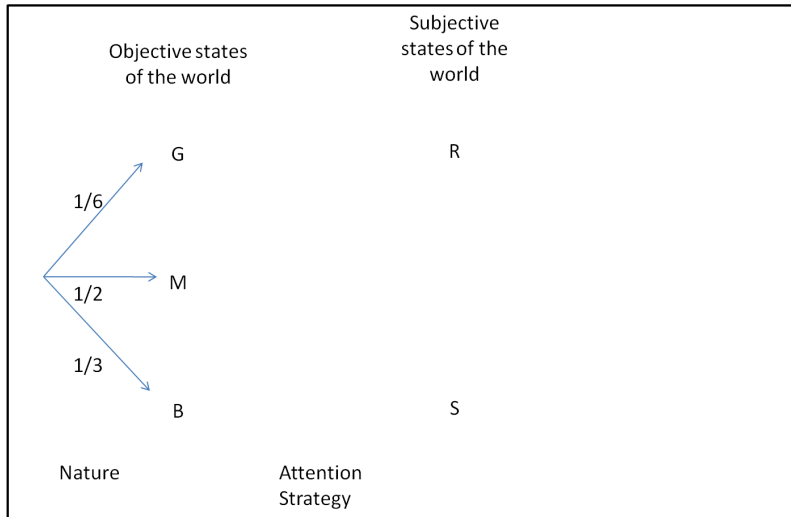
- A canonical model of rational inattention
- Implications of the canonical model

- Objective states of the world
  - e.g. Demand could be 'good', 'medium' or 'bad'
- Decision maker chooses an action
  - Set price to be high, medium or low
- Gross payoff depends on act and state
  - Quantity sold depends on price and demand

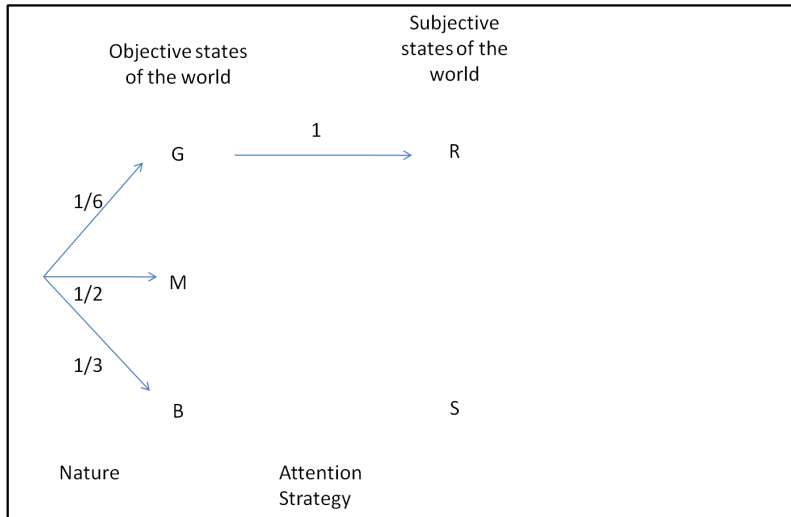
# The Choice Problem



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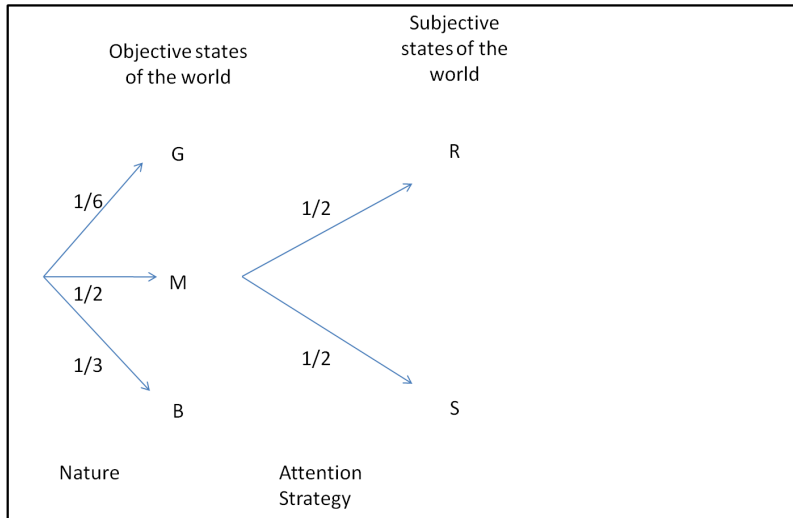


# The Choice Problem

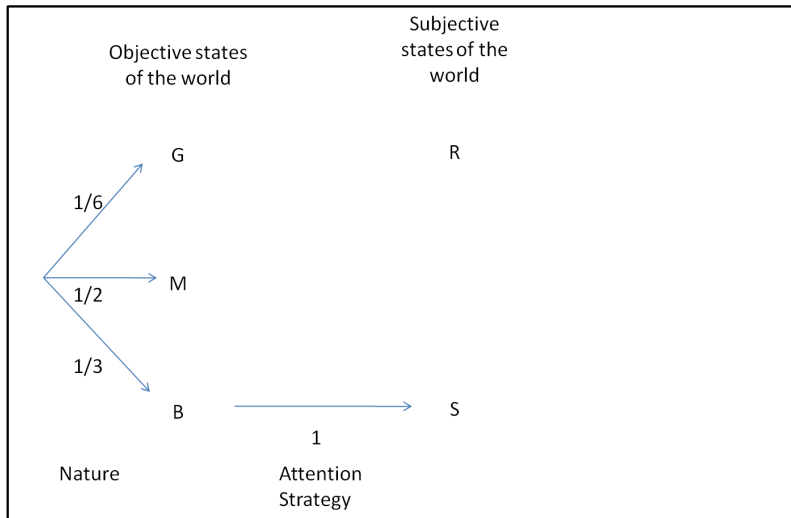




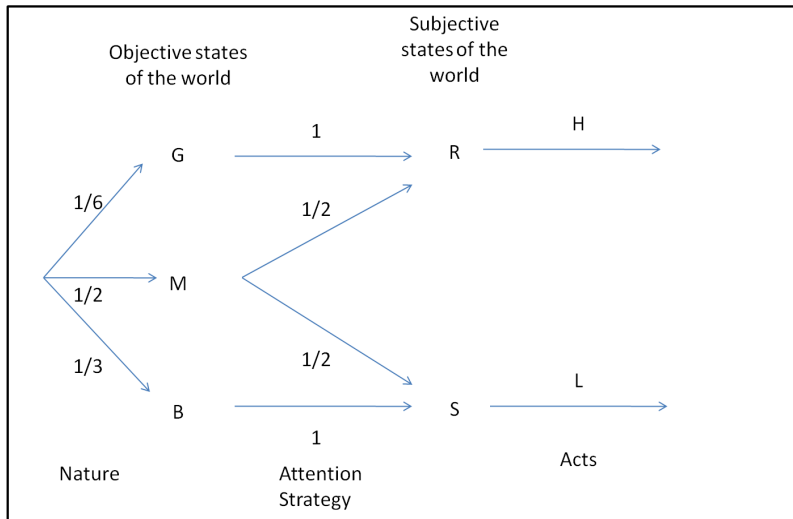
# The Choice Problem



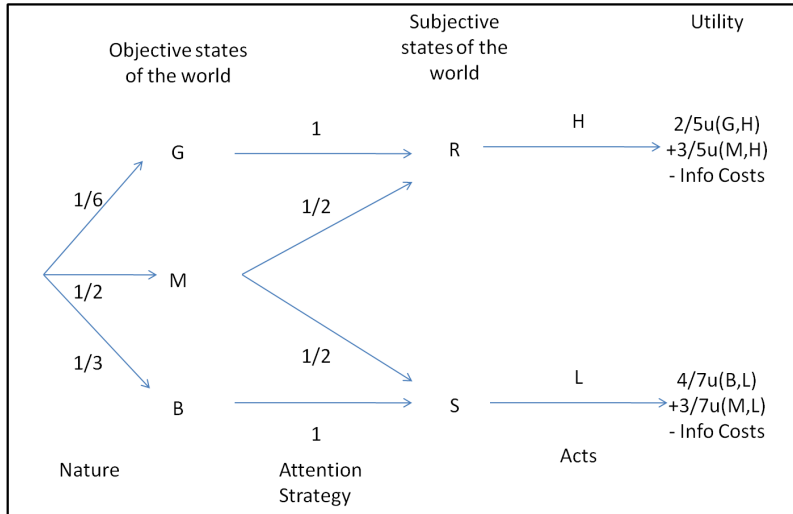
# The Choice Problem



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# The Choice Problem



- $\Omega = \{\omega_1, \dots, \omega_M\}$ : Objective States
- $\Delta(\Omega)$ : Probability distributions over  $\Omega$
- $X$ : Finite set of outcomes
- $U : X \rightarrow \mathbb{R}$ : (Expected) utility function over outcomes
- $f : \Omega \rightarrow X$ : Act, with  $F$  the set of all acts and  $\mathcal{F}$  non-empty subsets of  $F$
- $\{\beta, A\}$ : Decision problem:  $\beta \in \Delta(\Omega)$ ,  $A \in \mathcal{F}$

- A strategy consists of a set of subjective states

$$\{t^1, \dots, t^N\} = T(\lambda) \in \Delta(\Omega)^N$$

- and conditional probabilities linking objective and subjective states

$$\lambda : \Omega \times T(\lambda) \rightarrow [0, 1]$$

$\lambda_m(t^n)$  is the probability of subjective state  $t^n$  conditional on objective state  $\omega_m$

- Which obey Bayes law

$$t_m^n = \frac{\beta_m \lambda_m(t^n)}{\sum_{k=1}^M \beta_k \lambda_k(t^n)}$$

- For a decision problem  $\{\beta, A\}$  decision maker must choose
  - An attention strategy  $\lambda \in \Lambda(\beta)$
  - A choice function for acts  $C : T(\lambda) \rightarrow A$
- In order to maximize

Expected utility from acts - cost of information

$$\sum_{m=1}^M \beta_m \sum_{t \in T(\lambda)} \lambda_m(t) U(C_t(\omega_m)) - K(\lambda, \beta)$$

- where
  - $\Lambda(\beta)$ : set of attention strategies available from prior  $\beta$
  - $K : \beta \times \Lambda(\beta) \rightarrow \mathbb{R}$ : cost of attention strategy

- Decision maker can choose any form of information structure
- Nests other models of information acquisition
  - Shannon Entropy
  - Fixed signals
  - Partitions
  - Fixed capacity
  - Sequential Search



- Data: State dependant stochastic choice
- For each  $\omega_i \in \Omega$  and  $f \in A$  :
- $D_i(f)$ : probability of choosing act  $f$  in state  $\omega_i$
- Easy to observe in the laboratory
- With assumptions, can be observed outside the lab

- Question: If we don't make explicit assumptions about the costs of information, can we make any predictions about rationally inattentive behavior?
- Answer: Yes, if we assume that more information is more costly

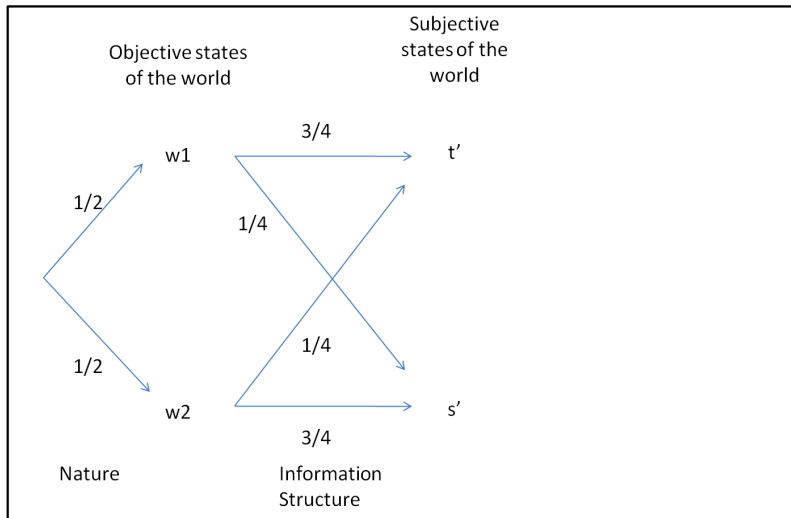
## An Aside: Blackwell Information Ordering

- $\lambda$  is sufficient for  $\lambda'$  if there exists a  $|T(\lambda)| \times |T(\lambda')|$  matrix  $B$  that

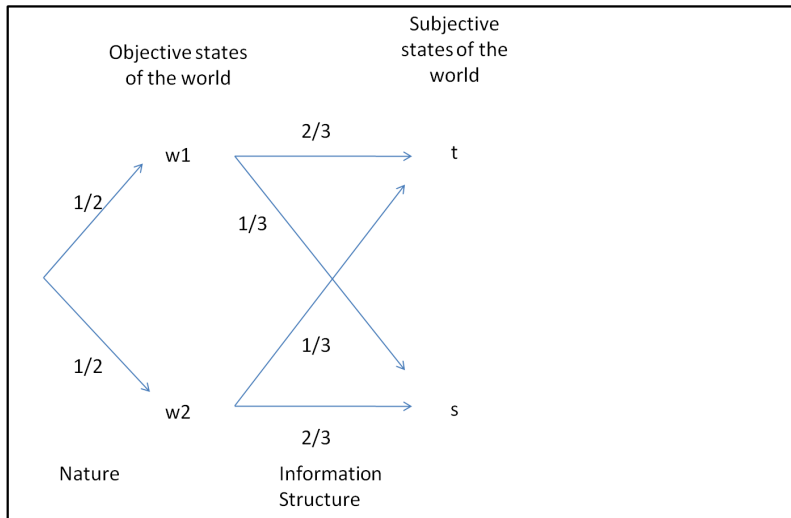
$$\sum_j B^{ij} = 1 \quad \forall j$$
$$\lambda'_m(t^j) = \sum_i B^{ij} \lambda_m(t^i) \quad \forall j$$

- The matrix  $B$  'scrambles' the information in  $\lambda$  in order to get to  $\lambda'$
- $B^{ij}$  probability of going from subjective state  $t^i$  to the subjective state  $t^j$

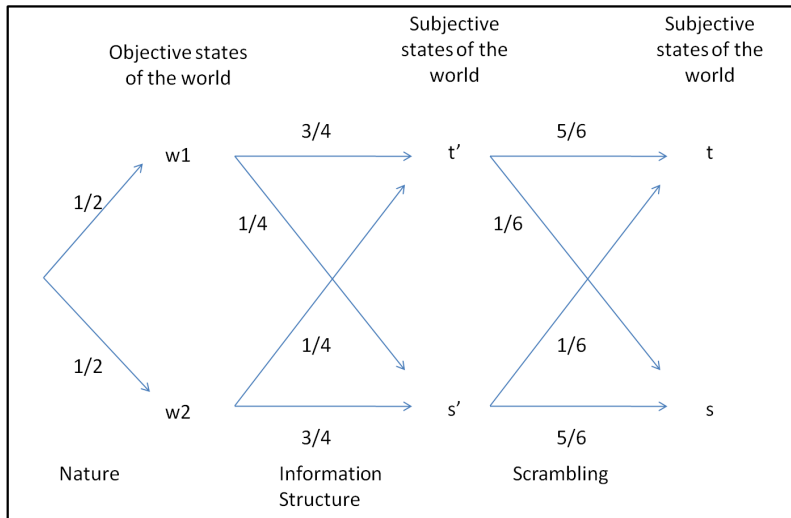
# Sufficiency: An Example



# Sufficiency: An Example



# Sufficiency: An Example



## An Aside: Blackwell's Theorem

- Let  $V(\beta, A, \lambda)$  be the *value* of using attentional structure  $\lambda$  in environment  $\{A, \lambda\}$

$$V(\beta, A, \lambda) = \max_{\{C_t\}_{T(\lambda)} \in A} \sum_{m=1}^M \beta_m \sum_{t \in T(\lambda)} \lambda_m(t) U(C_t(\omega_m))$$

- An information structure  $\lambda$  is sufficient for information structure  $\lambda'$  if and only if

$$V(\beta, A, \lambda) \geq V(\beta, A, \lambda') \quad \forall \{\beta, A\}$$

# Observing Attentional Strategies

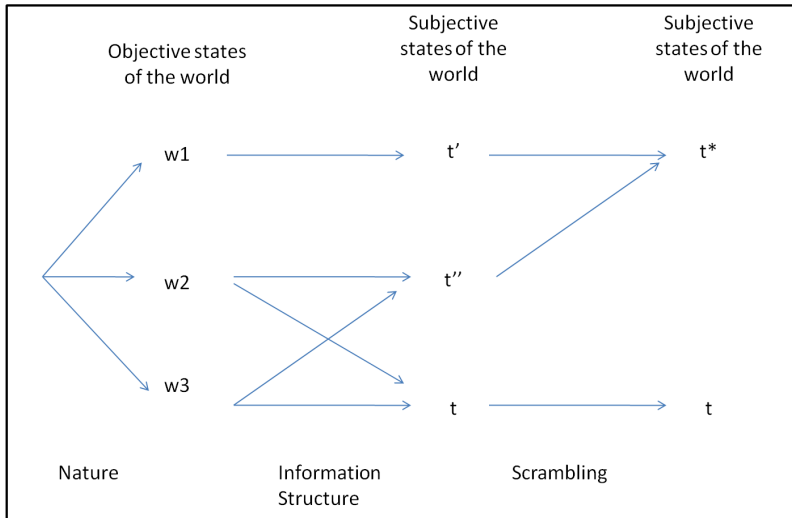
- Observation: Assuming more information is more costly, one subjective state per act
  - Will never have  $t'$  and  $t''$  such that  $C(t') = C(t'')$
- Why? Let  $\{\lambda, C\}$  be such a strategy, with  $C(t') = C(t'')$  for  $t', t'' \in T(\lambda)$
- Now consider alternative strategy  $\{\lambda', C\}$  such that
  - $\lambda(t) = \lambda'(t) \forall t \neq t', t''$
  - $\lambda'$  'merges'  $t$  and  $t''$  to state  $t^*$  defined by

$$t_m^* = \frac{\beta_m (\lambda_m(t') + \lambda_m(t''))}{\sum_{k=1}^M \beta_k (\lambda_k(t') + \lambda_k(t''))}$$

- Clearly  $\lambda$  is sufficient for  $\lambda'$



# Observing Attentional Strategies



- Means  $\lambda'$  is cheaper than  $\lambda$ , yet

$$V(\beta, A, \lambda) = V(\beta, A, \lambda')$$

(assuming choices were optimal under  $\lambda$ )

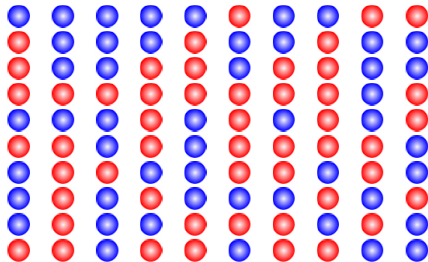
- Thus, never optimal to choose the same act in two states
- Means that we can recover attentional strategies from stochastic choice
  - For each chosen act  $f$  a subjective state  $t^f$  such that

$$\lambda_i(t^f) = D_i(f)$$

# Optimal Behavior Under Blackwell Costs

- Choice of act optimal given attentional strategy
- Choice of attention strategy optimal

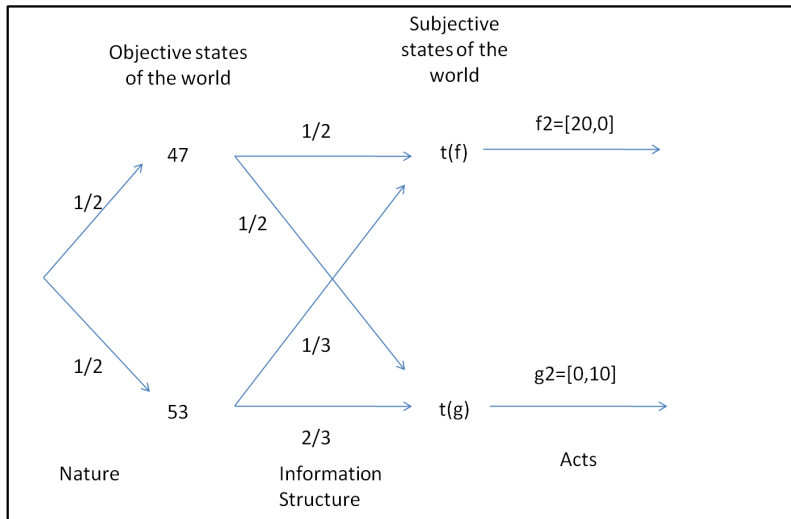
# Optimal Choice of Act



Act	Payoff 47 red dots	Payoff 53 red dots
$f_2$	20	0
$g_2$	0	10

Prior:  $\{0.5, 0.5\}$

# Optimal Choice of Acts



- Posterior probability of 47 red balls when act  $g$  was chosen

$$\begin{aligned}t_{47}^g &= \frac{P(\omega = 47, g \text{ chosen})}{P(g \text{ chosen})} \\ &= \frac{\frac{1}{4}}{\frac{1}{4} + \frac{2}{6}} = \frac{3}{7}\end{aligned}$$

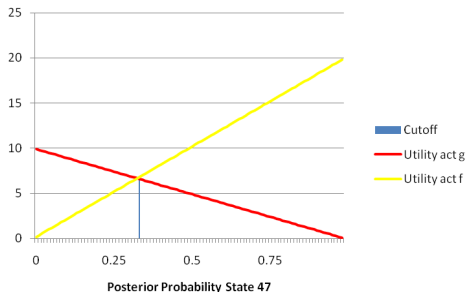
- But for this posterior

$$\begin{aligned}\frac{3}{7}U(f_{47}) + \frac{4}{7}U(f_{53}) &= \frac{3}{7}20 + \frac{4}{7}0 = 8.6 \\ \frac{3}{7}U(g_{47}) + \frac{4}{7}U(g_{53}) &= \frac{3}{7}0 + \frac{4}{7}10 = 5.7\end{aligned}$$

Condition 1 At every state  $s \in T(\lambda)$ , it must be the case that

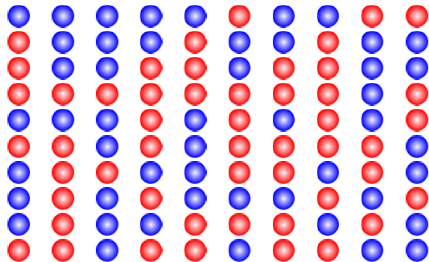
$$C(s) \in \arg \max_{f \in A} \sum_{m=1}^M s_m U(f(\omega_m))$$

- In  $2 \times 2$  case generates a cutoff  $c$  such that  $t^f \geq c$  and  $t^g \leq c$



# Optimal Choice of Attention Strategy

Question 1



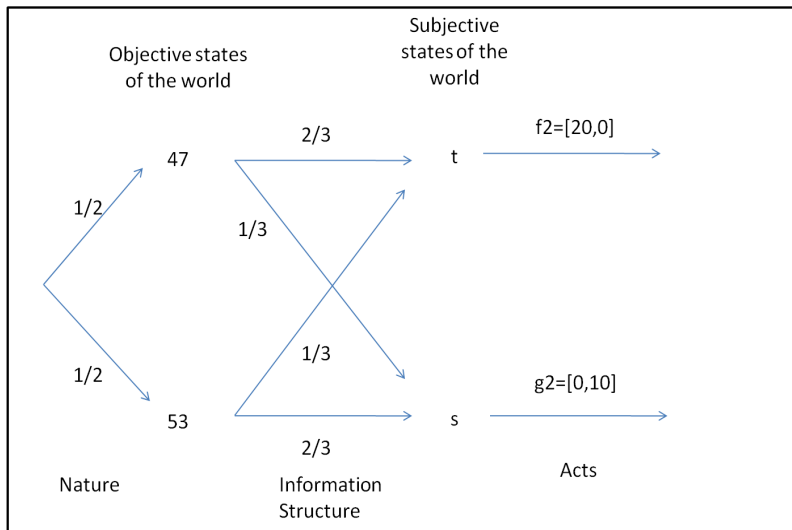
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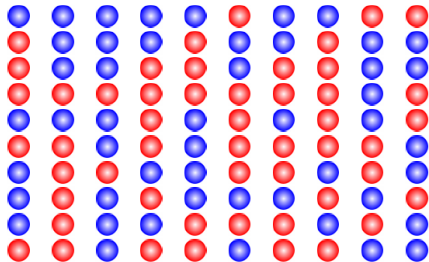
# Optimal Choice of Attention Strategy

Question 1



# Optimal Choice of Attention Strategy

Question 2

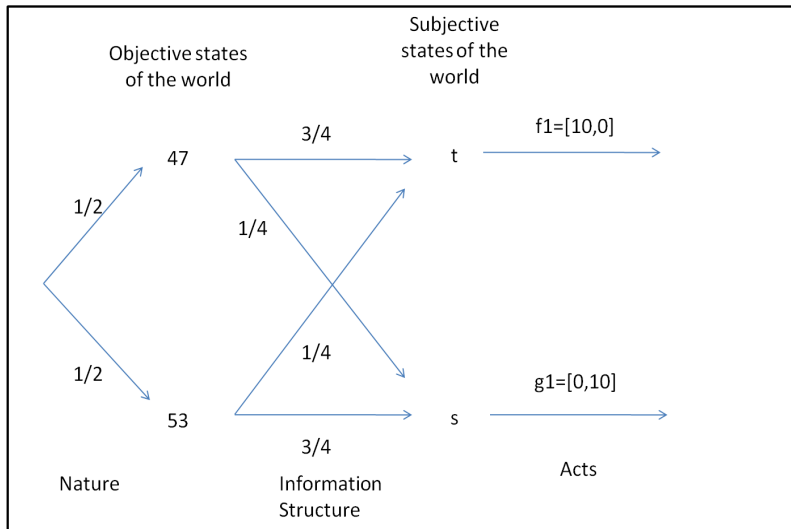


Act	Payoff 47 red dots	Payoff 53 red dots
$F_1$	10	0
$G_1$	0	10

Prior:  $\{0.5, 0.5\}$

# Optimal Choice of Attention Strategy

Question 2



# Optimal Choice of Attention Strategy

Example 2

$V$	$\lambda_1$	$\lambda_2$
$\{f_1, g_1\}$	$7\frac{1}{2}$	$6\frac{2}{3}$
$\{f_2, g_2\}$	$11\frac{1}{4}$	10

- Cost function must satisfy

$$V(\beta, \{f_1, g_1\}, \lambda_1) - K(\lambda_1, \beta) \geq V(\beta, \{f_1, g_1\}, \lambda_2) - K(\lambda_2, \beta)$$

$$V(\beta, \{f_2, g_2\}, \lambda_2) - K(\lambda_2, \beta) \geq V(\beta, \{f_2, g_2\}, \lambda_1) - K(\lambda_1, \beta)$$

- Which implies

$$\frac{5}{6} = V(\beta, \{f_1, g_1\}, \lambda_1) - V(\beta, \{f_1, g_1\}, \lambda_2) \geq$$

$$K(\lambda_1, \beta) - K(\lambda_2, \beta) \geq$$

$$V(\beta, \{f_2, g_2\}, \lambda_1) - V(\beta, \{f_2, g_2\}, \lambda_2) = 1\frac{1}{4}$$

# Optimal Choice of Attention Strategy

- Surplus must be maximized by correct assignments

$$\begin{aligned} & V(\beta, \{f_1, g_1\}, \lambda_1) - V(\beta, \{f_1, g_1\}, \lambda_2) \\ & + V(\beta, \{f_2, g_2\}, \lambda_2) - V(\beta, \{f_2, g_2\}, \lambda_1) \\ & \geq 0 \end{aligned}$$

- To guarantee the existence of a cost function require a stronger condition

**Condition 2** For any  $\beta$  and observed sequence of acts  $A^1, \dots, A^K$  and associated information structures  $\lambda^1, \dots, \lambda^K$

$$\begin{aligned} & V(\beta, A^1, \lambda^1) - V(\beta, A^1, \lambda^2) \\ & + V(\beta, A^2, \lambda^2) - V(\beta, A^2, \lambda^3) \\ & + \dots \\ & + V(\beta, A^K, \lambda^K) - V(\beta, A^K, \lambda^1) \\ & \geq 0 \end{aligned}$$

## Theorem

For a sequence of decision problems  $\{\beta, A_l\}_{l=1}^L$ , attention strategies  $\{\lambda^l\}_{l=1}^L$  and choice functions  $\{C^l\}_{l=1}^L$  such that  $C^l : T(\lambda^l) \rightarrow A^l$  the following two statements are equivalent

- ①  $\{\beta, A^l\}_{l=1}^L$  and  $\{\lambda^l\}_{l=1}^L$  satisfy conditions 1 and 2
- ② there exists a  $K : \beta \times \Lambda(\beta) \rightarrow \mathbb{R}$  such that  $\lambda^l$  and  $C^l$  solve the decision problem for each  $\{\beta, A^l\}$

## Proof.

2  $\rightarrow$  1 Trivial

1  $\rightarrow$  2 Rochet [1987]



- This problem is familiar from the implementation literature
- Say there were a set of environments  $X_1 \dots X_N$  and actions  $B_1 \dots B_M$  such that the utility of each environment and each state is given by

$$u(X_i, B_j)$$

- Say we want to implement a mechanism such that action  $Y(X_i)$  is taken at in each environment.
- We need to find a taxation scheme  $\tau : B_1 \dots B_M \rightarrow \mathbb{R}$  such that

$$u(X_i, Y(X_i)) - \tau(Y(X_i)) \geq u(X_i, B) - \tau(B) \\ \forall B_1 \dots B_M$$

- This is the same as our problem.

- Taxation theorem: this is the equivalent problem to the following:

- Find  $\theta : \{A_i\}_{i=1}^L \rightarrow \mathbb{R}$  such that, for all  $A_i, A_j$

$$V(\beta, A_i, \lambda^i) - \theta(A_i) \geq V(\beta, A_j, \lambda^j) - \theta(A_j)$$

- Just define  $K(\lambda) = \theta(A_i)$  if  $\lambda = \lambda^i$  for some  $i$ , or  $= \infty$  otherwise



- Now, pick some arbitrary  $A_0$  and define

$$T(A) = \sup_{\text{all chains s.t } A_0 \text{ to } A=A_M} \sum_{n=1}^{M-1} V(\beta, A_{i+1}, \lambda^i) - V(\beta, A_i, \lambda^i)$$

- Condition 2 implies that  $T(A_0) = 0$
- It also implies that

$$T(A_0) \geq T(A_i) + V(\beta, A_0, \lambda^i) - V(\beta, A_i, \lambda^i)$$

- So  $T(A_i)$  is bounded

- Furthermore, for any  $A_i, A_j$  we have

$$T(A_i) \geq T(A_j) + V(\beta, A_i, \lambda^j) - V(\beta, A_j, \lambda^j)$$

- So, setting  $\theta(A_j) = V(\beta, A_j, \lambda^j) - T(A_j)$ , we get

$$V(\beta, A_i, \lambda^j) - \theta(A_i) \geq V(\beta, A_i, \lambda^j) - \theta(A_j)$$

# Cost Function and Blackwell Ordering

- Observation: if  $\lambda$  is sufficient for  $\lambda'$ , then

$$V(\beta, A, \lambda) \geq V(\beta, A, \lambda')$$

- for all  $\beta, A$
- Thus, for any cost function that rationalizes behavior

$$\begin{aligned} V(\beta, A', \lambda') - K(\beta, \lambda') &\geq V(\beta, A', \lambda) - K(\beta, \lambda) \\ 0 &\geq V(\beta, A', \lambda') - V(\beta, A', \lambda) \geq \\ &K(\beta, \lambda') - K(\beta, \lambda) \\ \Rightarrow K(\beta, \lambda) &\geq K(\beta, \lambda') \end{aligned}$$

- Cost function will weakly obey Blackwell
- For unchosen information structures, assume cost is equal to lowest cost chosen Blackwell dominant option

# Cost Function and Blackwell Ordering

- Does not guarantee strict observance with Blackwell
- Example: two states,  $\beta_1 = 0.5$ , 3 acts

<b>Act</b>	<b>Payoff 47 red dots</b>	<b>Payoff 53 red dots</b>
<i>f</i>	10	0
<i>g</i>	0	10
<i>h</i>	7.5	7.5

- Strategy  $\lambda_1(t^f) = 0.75$ ,  $\lambda_2(t^f) = 0.75$ .
- $V(\beta, A, \lambda) = 7.5$