

Acting

Decision Making

Maximum Expected Utility

Simple Decision Making

A simple decision making situation \mathcal{D} :

- A set of possible actions $Val(A) = \{a^1, ..., a^K\}$
- A set of states $Val(X) = \{x^1, ..., x^N\}$
- A distribution P(X | A)
- A utility function U(X, A)

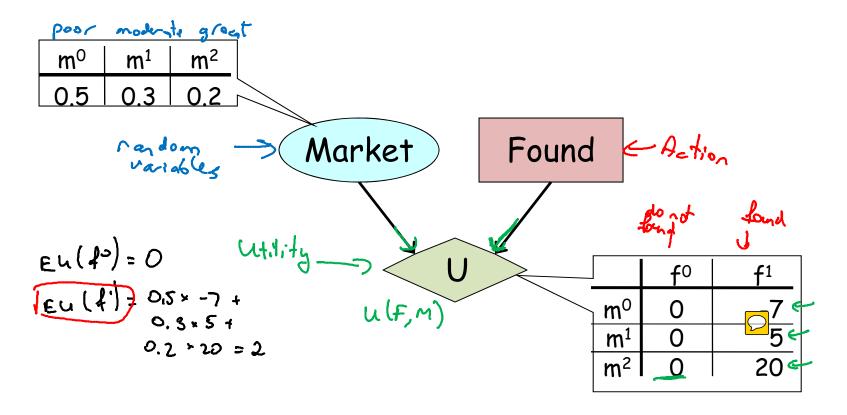
Expected Utility

$$\mathrm{EU}[\underline{\mathcal{D}}[a]] = \sum_{m{x}} P(m{x} \mid a) \underline{U(m{x},a)}$$

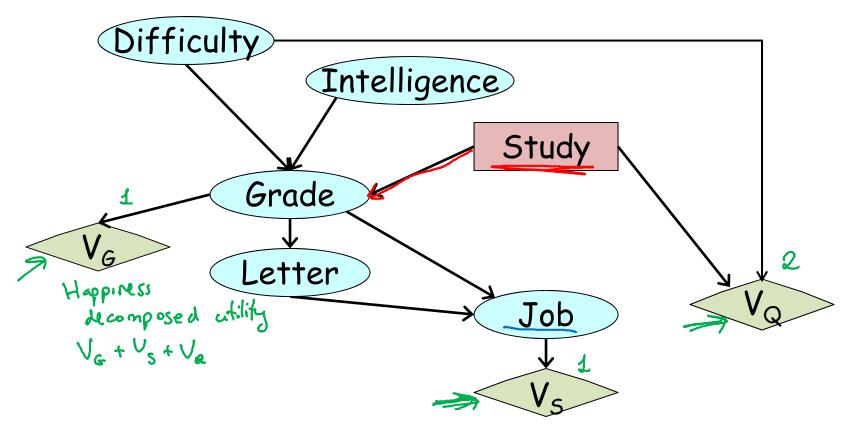
• Want to choose action at that maximizes the expected utility

$$a^* = \operatorname{argmax}_a \operatorname{EU}[\mathcal{D}[a]]$$

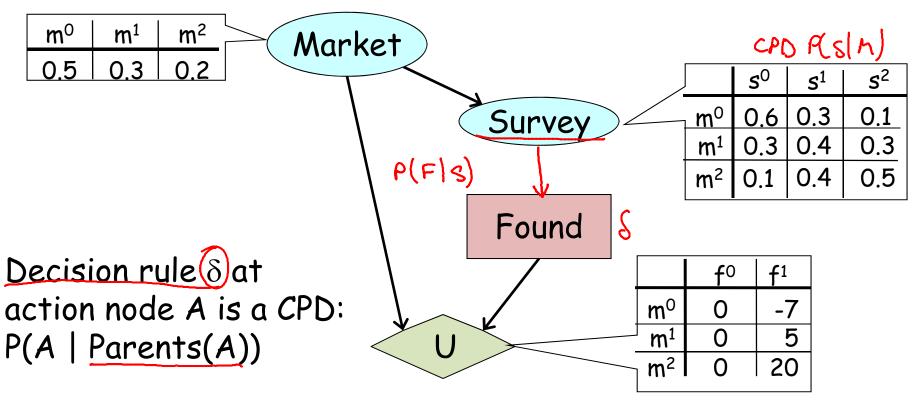
Simple Influence Diagram



More Complex Influence Diagram



Information Edges



Expected Utility with Information

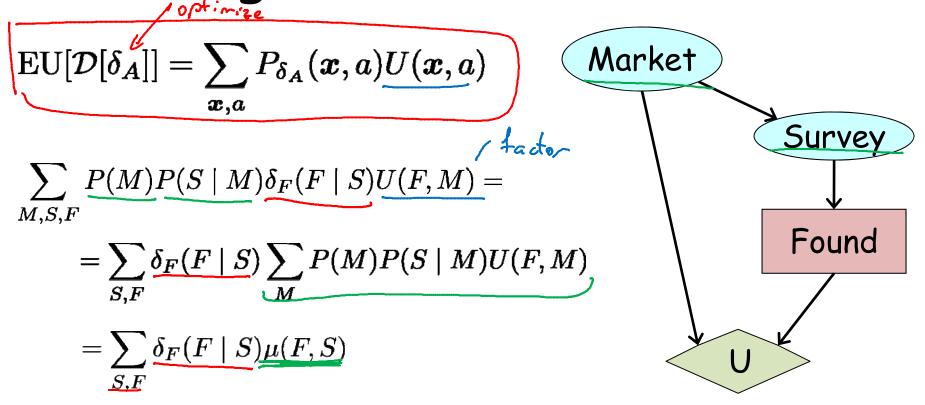
$$\mathrm{EU}[\mathcal{D}[\delta_A]] = \sum_{m{x},a} P_{\delta_A}(m{x},a) U(m{x},a)$$

• Want to choose the decision rule δ_A that maximizes the expected utility

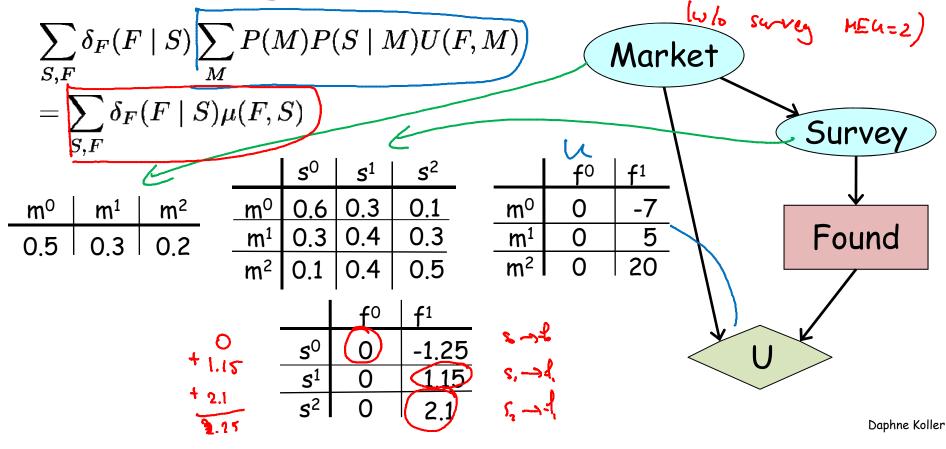
$$\operatorname{argmax}_{\delta_A} \operatorname{EU}[\mathcal{D}[\delta_A]]$$

$$ext{MEU}(\mathcal{D}) = \max_{\delta_A} ext{EU}[\mathcal{D}[\delta_A]]$$

Finding MEU Decision Rules



Finding MEU Decision Rules



More Generally

$$\begin{aligned} & \mathbf{E}\mathbf{U}[\mathcal{D}[\delta_{A}]] = \sum_{\mathbf{x},a} P_{\mathbf{A}}(\mathbf{x},a) \overline{U(\mathbf{x},a)} & \underline{Z} = \mathbf{P}\mathbf{a}_{A} \stackrel{\mathsf{dosevalor}}{prior} h A \\ & \underline{W} = \{X_{1}, \dots, X_{n}\} - Z \end{aligned}$$

$$& = \sum_{X_{1},\dots,X_{n},A} \left(\left(\prod_{i} P(X_{i} \mid \mathbf{P}\mathbf{a}_{X_{i}}) \right) U(\mathbf{P}\mathbf{a}_{U}) \delta_{A}(A \mid \mathbf{Z}) \right)$$

$$& = \sum_{\mathbf{Z},A} \delta_{A}(A \mid \mathbf{Z}) \sum_{\mathbf{W}} \left(\left(\prod_{i} P(X_{i} \mid \mathbf{P}\mathbf{a}_{X_{i}}) \right) U(\mathbf{P}\mathbf{a}_{U}) \right)$$

$$& = \sum_{\mathbf{Z},A} \delta_{A}(A \mid \mathbf{Z}) \mu(A,\mathbf{Z}) \qquad \delta_{A}^{*}(a \mid \mathbf{z}) = \begin{cases} 1 & a = \operatorname{argmax}_{A} \mu(A,\mathbf{z}) \\ 0 & \text{otherwise} \end{cases}$$

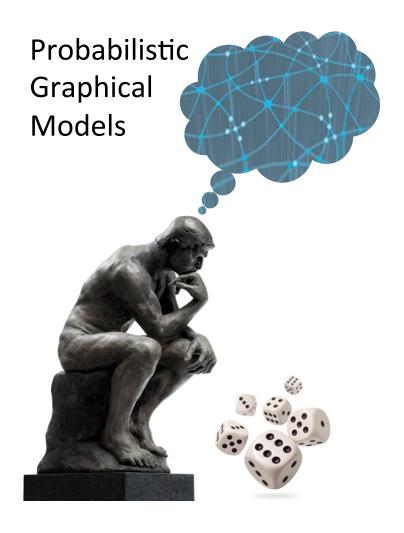
MEU Algorithm Summary

- To compute MEU & optimize decision at A:
 - Treat A as random variable with arbitrary CPD
 - Introduce utility factor with scope Pa_{\cup}
- Eliminate all variables except A, Z (A's parents) to produce factor $\mu(A, Z)$,
 - For each z, set:

$$\delta_{A}^{*}(a \mid oldsymbol{z}) = \left\{egin{array}{ll} 1 & a = \mathrm{argmax}_{A} \mu(A, oldsymbol{z}) \ 0 & \mathrm{otherwise} \end{array}
ight.$$

Decision Making under Uncertainty

- MEU principle provides rigorous foundation
- PGMs provide structured representation for probabilities, actions, and utilities
- PGM inference methods (VE) can be used for
 - Finding the optimal strategy
 - Determining overall value of the decision situation
- Efficient methods also exist for:
 - Multiple utility components
 - Multiple decisions



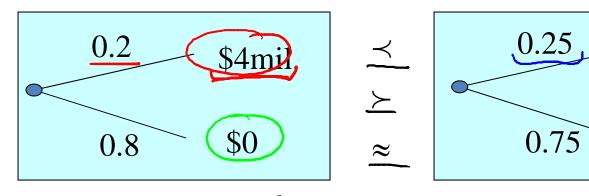
Acting

Decision Making

Utility Functions

Utilities and Preferences

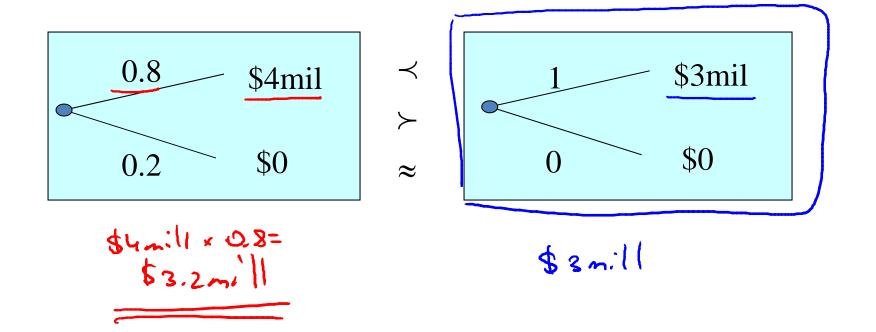
l.Heries



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Utility = Payoff?

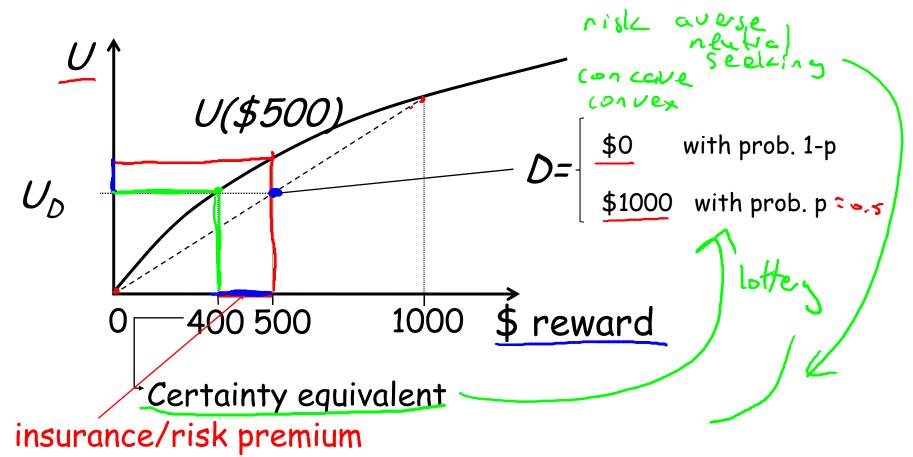


St. Petersburg Paradox

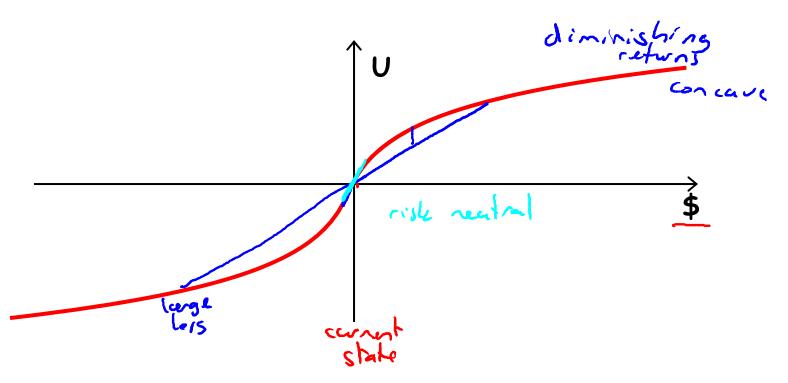


- Fair coin is tossed repeatedly until it comes up heads, say on the nth toss
- Payoff = $$2^n$

$$\frac{1}{2} \times 2 + \frac{1}{4} \times 4 + \frac{1}{8} \times 8 + \dots = \infty$$
most people valle 2×82



Typical Utility Curve



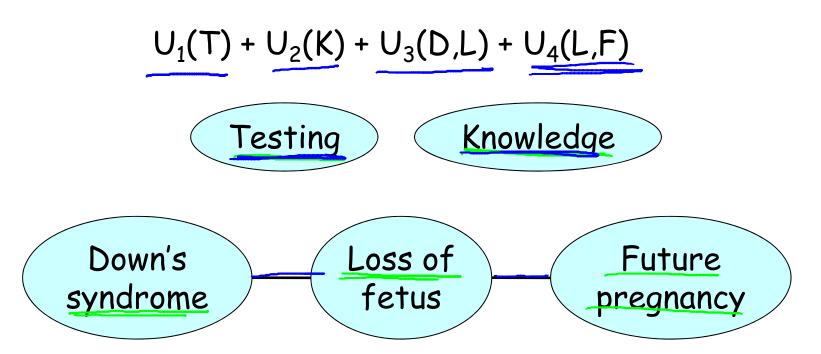
Multi-Attribute Utilities

 All attributes affecting preferences must be integrated into one utility function

money , time, pleasure, ...

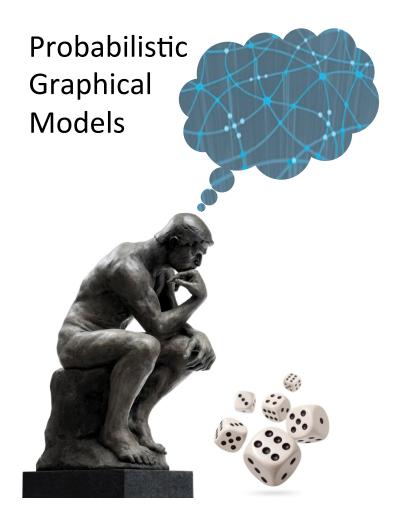
- · Human life
 - Micromorts 1/2000000 chance I death 2 \$20 1930
 - QALY (quality-adjusted life year)

Example: Prenatal diagnosis



Summary

- Our utility function determines our preferences about decisions that involve uncertainty
- Utility generally depends on multiple factors
 - Money, time, chances of death, ...
- Relationship is usually nonlinear
 - Shape of utility curve determines attitude to risk
- Multi-attribute utilities can help decompose high-dimensional function into tractable pieces



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Value of Perfect Information

Value of Information

alue of perfect information

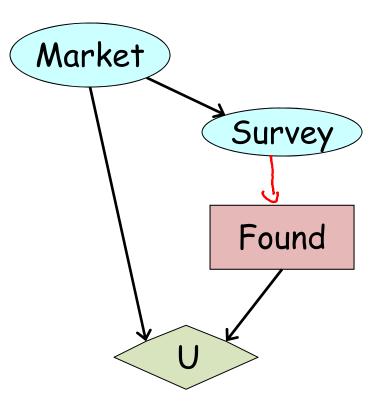
- VPI(A | X) is the value of observing X before choosing an action at A
- D = original influence diagram
- $\mathcal{D}_{X \to A}$ = influence diagram with edge $X \to A$

$$\mathrm{VPI}(A \mid X) := \mathrm{MEU}(\mathcal{D}_{X \rightarrow A}) - \mathrm{MEU}(\mathcal{D})$$

Finding MEU Decision Rules

$$neu(0_{s\to F}) - meu(0)$$

3.25 2 = 1.25



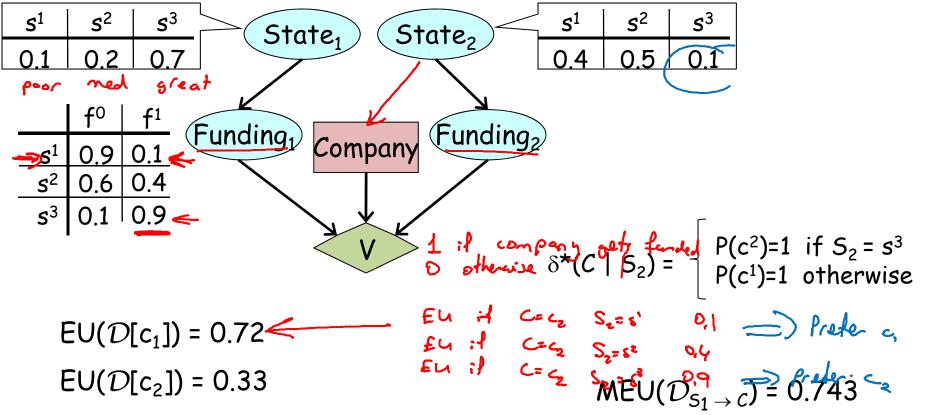
Value of Information

 $VPI(A \mid X) := MEU(\mathcal{D}_{X \to A}) - MEU(\mathcal{D})$

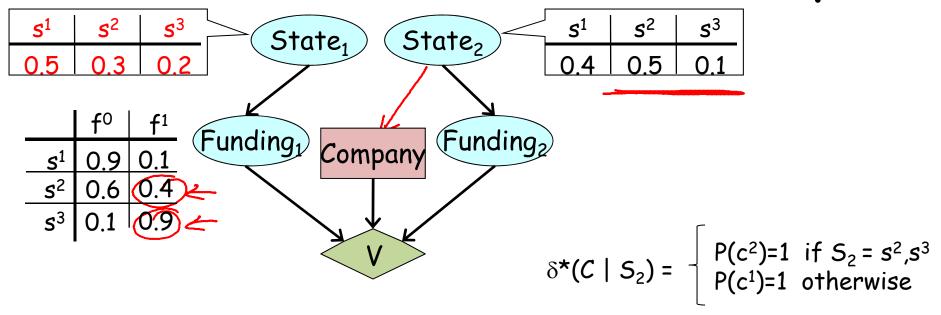
optimizing S(A) = x) optimizing S(A) =

- Theorem:
 - $-VPI(A \mid X) \geq 0$
 - $VPI(A \mid X) = 0$ if and only if the optimal decision rule for \mathcal{D} is still optimal for $\mathcal{D}_{X \to A}$

Value of Information Example



Value of Information Example

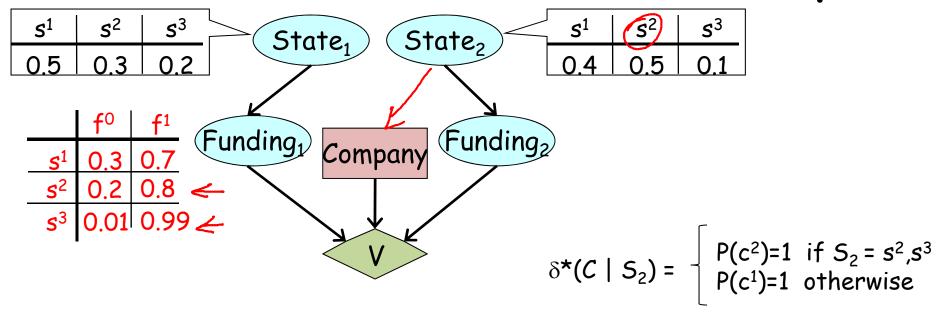


$$EU(\mathcal{D}[c_1]) = 0.35$$

$$EU(\mathcal{D}[c_2]) = 0.33$$

$$MEU(\mathcal{D}_{S_2 \to C}) = 0.43$$

Value of Information Example



$$EU(D[c_1]) = 0.788$$

$$\mathsf{EU}(\mathcal{D}[\mathsf{c}_2]) = 0.779$$

$$\mathsf{MEU}(\mathcal{D}_{\mathbf{S}_1 \to \mathcal{C}}) = 0.8142$$

Summary

- Influence diagrams provide clear and coherent semantics for the value of making an observation
 - Difference between values of two IDs
- Information is valuable if and only if it induces a change in action in at least one context