## Decision Networks and Value of Perfect Information



## Decision Networks



## Decision Networks



## Decision Networks

- MEU: choose the action which maximizes the expected utility given the evidence
- Can directly operationalize this with decision networks
- Bayes nets with nodes for utility and actions
- Lets us calculate the expected utility for each action
- New node types:
- Chance nodes (just like BNs)
- Actions (rectangles, cannot have parents, act as observed evidence)
- Utility node (diamond, depends on action and chance nodes)



## Decision Networks

- Action selection
- Instantiate all evidence
- Set action node(s) each possible way
- Calculate posterior for all parents of utility node, given the evidence
- Calculate expected utility for each action
- Choose maximizing action



## Decision Networks

Umbrella = leave

$$
\begin{aligned}
& \mathrm{EU}(\text { leave })=\sum_{w} P(w) U(\text { leave }, w) \\
& =0.7 \cdot 100+0.3 \cdot 0=70
\end{aligned}
$$

Umbrella = take

$$
\mathrm{EU}(\text { take })=\sum_{w} P(w) U(\text { take }, w)
$$

$$
=0.7 \cdot 20+0.3 \cdot 70=35
$$

## Optimal decision = leave

| $W$ | $P(W)$ |
| :---: | :---: |
| sun | 0.7 |
| rain | 0.3 |



| $A$ | $W$ | $U(A, W)$ |
| :---: | :---: | :---: |
| leave | sun | 100 |
| leave | rain | 0 |
| take | sun | 20 |
| take | rain | 70 |

$$
\operatorname{MEU}(\varnothing)=\max _{a} \operatorname{EU}(a)=70
$$

## Decisions as Outcome Trees



- Almost exactly like expectimax / MDPs
- What's changed?


## Example: Decision Networks

Umbrella = leave

$$
\begin{aligned}
& \mathrm{EU}(\text { leave } \mid \text { bad })=\sum_{w} P(w \mid \text { bad }) U(\text { leave, } w) \\
& \quad=0.34 \cdot 100+0.66 \cdot 0=34
\end{aligned}
$$

Umbrella = take

$$
\begin{aligned}
& \mathrm{EU}(\text { take } \mid \text { bad })=\sum_{w} P(w \mid \text { bad }) U(\text { take }, w) \\
& =0.34 \cdot 20+0.66 \cdot 70=53
\end{aligned}
$$

Optimal decision = take


## Decisions as Outcome Trees



## Single-Stage Decision

|  |  |  |  |
| :--- | :--- | :--- | :--- | :--- |

# Diagnosis Decision Network 



## Markov Decision Process (MDP)



Figure 9.12: Decision network representing a finite part of an MDP

## Ghostbusters Decision Network

Demo: Ghostbusters with probability


## Value of Information



## Value of Information

- Idea: compute value of acquiring evidence
- Can be done directly from decision network
- Example: buying oil drilling rights
- Two blocks A and B, exactly one has oil, worth k
- You can drill in one location
- Prior probabilities 0.5 each, \& mutually exclusive
- Drilling in either $A$ or $B$ has $E U=k / 2, M E U=k / 2$
- Question: what's the value of information of O?
- Value of knowing which of A or B has oil
- Value is expected gain in MEU from new info
- Survey may say "oil in a" or "oil in b," prob 0.5 each
- If we know OilLoc, MEU is $k$ (either way)
- Gain in MEU from knowing OilLoc?
- VPI(OilLoc) $=\mathrm{k} / 2$
- Fair price of information: $k / 2$



## VPI Example: Weather

MEU with no evidence

$$
\operatorname{MEU}(\not)=\max _{a} \operatorname{EU}(a)=70
$$

MEU if forecast is bad

$$
\operatorname{MEU}(F=\mathrm{bad})=\max _{a} \operatorname{EU}(a \mid \mathrm{bad})=53
$$

MEU if forecast is good

$$
\operatorname{MEU}(F=\operatorname{good})=\max _{a} \operatorname{EU}(a \mid \text { good })=95
$$



| $A$ | $W$ | $U$ |
| :---: | :---: | :---: |
| leave | sun | 100 |
| leave | rain | 0 |
| take | sun | 20 |
| take | rain | 70 |

Forecast distribution

| $F$ | $P(F)$ |
| :---: | :---: |
| good | 0.59 |
| bad | 0.41 |

$$
0.59 \cdot(95)+0.41 \cdot(53)-70
$$

$$
\operatorname{VPI}\left(E^{\prime} \mid e\right)=\left(\sum_{e^{\prime}} P\left(e^{\prime} \mid e\right) \operatorname{MEU}\left(e, e^{\prime}\right)\right)-\operatorname{MEU}(e)
$$



## Value of Information

- Assume we have evidence $\mathrm{E}=\mathrm{e}$. Value if we act now:

$$
\operatorname{MEU}(e)=\max _{a} \sum_{s} P(s \mid e) U(s, a)
$$

- Assume we see that $\mathrm{E}^{\prime}=\mathrm{e}^{\prime}$. Value if we act then:
$\operatorname{MEU}\left(e, e^{\prime}\right)=\max _{a} \sum_{s} P\left(s \mid e, e^{\prime}\right) U(s, a)$
- BUTE' is a random variable whose value is unknown, so we don't know what e' will be
- Expected value if $E$ ' is revealed and then we act:

$$
\operatorname{MEU}\left(e, E^{\prime}\right)=\sum_{e^{\prime}} P\left(e^{\prime} \mid e\right) \operatorname{MEU}\left(e, e^{\prime}\right)
$$

- Value of information: how much MEU goes up by revealing E' first then acting, over acting now:


$$
\operatorname{VPI}\left(E^{\prime} \mid e\right)=\operatorname{MEU}\left(e, E^{\prime}\right)-\operatorname{MEU}(e)
$$

## VPI Properties

- Nonnegative

$$
\forall E^{\prime}, e: \operatorname{VPI}\left(E^{\prime} \mid e\right) \geq 0
$$



- Nonadditive
(think of observing $\mathrm{E}_{\mathrm{j}}$ twice)

$$
\operatorname{VPI}\left(E_{j}, E_{k} \mid e\right) \neq \mathrm{VPI}\left(E_{j} \mid e\right)+\operatorname{VPI}\left(E_{k} \mid e\right)
$$



- Order-independent

$$
\begin{aligned}
\mathrm{VPI}\left(E_{j}, E_{k} \mid e\right) & =\operatorname{VPI}\left(E_{j} \mid e\right)+\operatorname{VPI}\left(E_{k} \mid e, E_{j}\right) \\
& =\operatorname{VPI}\left(E_{k} \mid e\right)+\operatorname{VPI}\left(E_{j} \mid e, E_{k}\right)
\end{aligned}
$$



## Quick VPI Questions

- The soup of the day is either clam chowder or split pea, but you wouldn' t order either one. What's the value of knowing which it is?

- There are two kinds of plastic forks at a picnic. One kind is slightly sturdier. What's the value of knowing which?
- You' re playing the lottery. The prize will be $\$ 0$ or $\$ 100$. You can play any number between 1 and 100 (chance of winning is $1 \%)$. What is the value of knowing the winning number?


## Value of Imperfect Information?

- No such thing

- Information corresponds to the observation of a node in the decision network
- If data is "noisy" that just means we don't observe the original variable, but another variable which is a noisy version of the original one


## VPI Question

- VPI(OilLoc) ?
- VPI(ScoutingReport) ?
- VPI(Scout) ?
- VPI(Scout | ScoutingReport) ?
- Generally:

If Parents(U) $\Perp$ Z | CurrentEvidence Then VPI(Z|CurrentEvidence) $=0$


