# Bayes' Rule & Decision Trees



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Incorporating new information in the decision tree

- เ⊯Bayes' Rule
- PROTRAC, Inc. Problem
- 🖙 Farmer Jones' Problem

# Given

 $S_1, S_2, ... S_n$  possible states of nature  $P\{S_i\} \quad \textit{prior} \quad \text{probabilities}$   $O_1, O_2, ... O_m$  possible outcomes of an experiment

 $P\{O_i|S_i\}$  likelihood of an outcome

# Calculate

P{S<sub>i</sub>|O<sub>j</sub>} *posterior* probabilities

By the definition of conditional probability,

$$P\{S_i|O_j\} = \frac{P\{S_i \cap O_j\}}{P\{O_i\}}$$

 $\implies P\{S_i \cap O_j\} = P\{S_i | O_j\} P\{O_j\} = P\{O_j | S_i\} P\{S_i\}$ 

# Bayes' Rule

$$P\{S_i \cap O_i\} = P\{S_i | O_i\} P\{O_i\} = P\{O_i | S_i\} P\{S_i\}$$

$$\Rightarrow P\{S_i|O_j\} = \frac{P\{O_j|S_i\} P\{S_i\}}{P\{O_j\}}$$

## Incorporating New Information

Suppose that in the PROTRAC example, a market research study can be made before deciding which strategy (A, B, or C) to select. The results of this study can then be used to more accurately estimate the probabilities of a "Strong" or "Weak" market.



Test results are either

- Encouraging
- Discouraging

Reliability of the market study: "The past results with our test have tended to be in the 'right direction'. Specifically, in 60% of the instances when the market has been strong, the preceding market study was 'Encouraging', while in 70% of the instances when the market has been weak, the preceding market study was 'Discouraging'."

The statement about "reliability" of the market study provides:

#### Conditional Probabilities:

"In 60% of the instances when the market has been strong, the preceding market study was 'encouraging'"

P{EIS} = 60%

 $P\{DIS\} = 40\%$ 

"In 70% of the instances when the market has been weak, the preceding market study was 'discouraging'"

P{ F|W} = 30%

 $P\{D|W\} = 70\%$ 

**Bayes' Rule** can now be used to find the values for  $P\{S|E\}$ ,  $P\{S|D\}$ , etc.

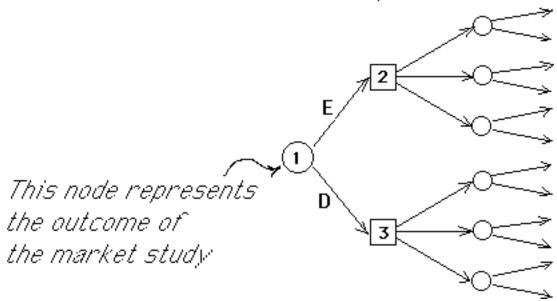
For example,

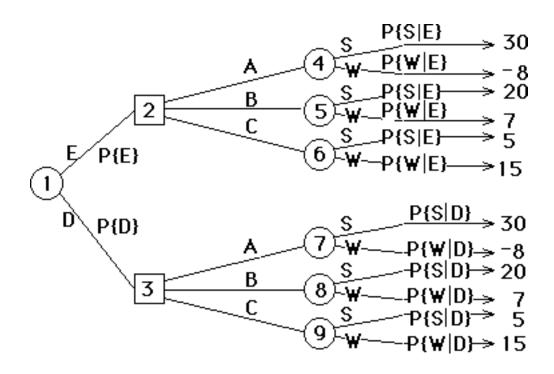
$$P\{S | E\} = \frac{P\{E | S\} P\{S\}}{P\{E\}}$$
$$= \frac{P\{E | S\} P\{S\}}{P\{E | S\} P\{S\}} + P\{E | W\} P\{W\}$$

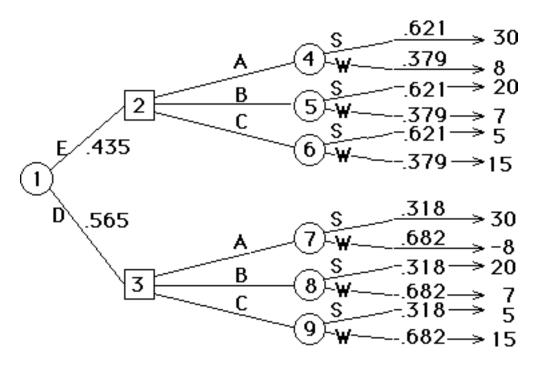
$$=\frac{(0.6)(0.45)}{(0.6)(0.45)+(0.3)(0.55)}$$

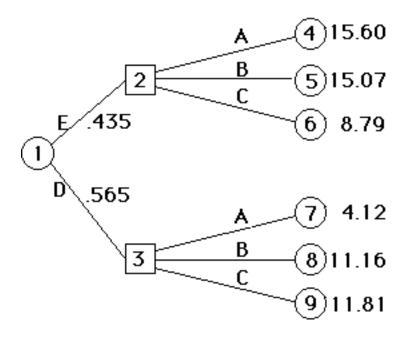
P{EIS} = 60% (0.6)(0.45) + (0.3)(0.55)  
P{DIS} = 40% = 
$$\frac{0.27}{0.27 + 0.165} = \frac{0.27}{0.435} = 0.621$$
  
P{DIW} = 70%

 The decision tree is now drawn with the decision nodes *following* the (random) outcome of the market study:

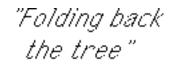


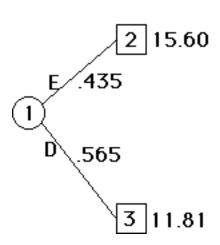


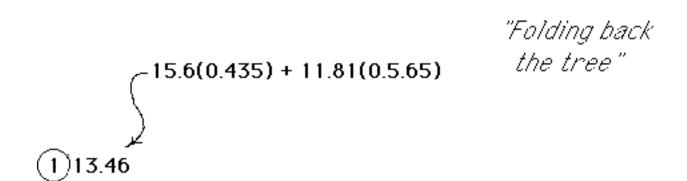




"Folding back the tree"







The maximum expected payoff which can be attained is 13.46

## Expected Value of Sample Information

EVWSI: "Expected Value With Sample Information"

EVWOI: "Expected Value Without Information"

EVSI: "Expected Value of Sample Information"

EVSI = EVWSI-EVWOI

#### **EXAMPLE**

EVSI = EVWSI-EVWOI

In the "PROTRAC" decision problem,

EVWOI = 12.85 

Expected payoff with no market study

Expected payoff using Formarket study Formarket Study

EVSI = 13.46 - 12.85 = 0.61

# EXPECTED VALUE OF PERFECT INFORMATION

**EVWPI**: "Expected Value With

Perfect Information"

**EVWOI:** "Expected Value Without Information"

EVPI = EVWPI - EVWOI

#### **EXAMPLE**

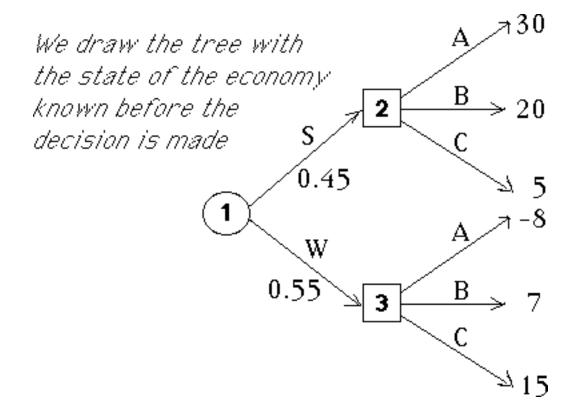
PROTRAC decision problem

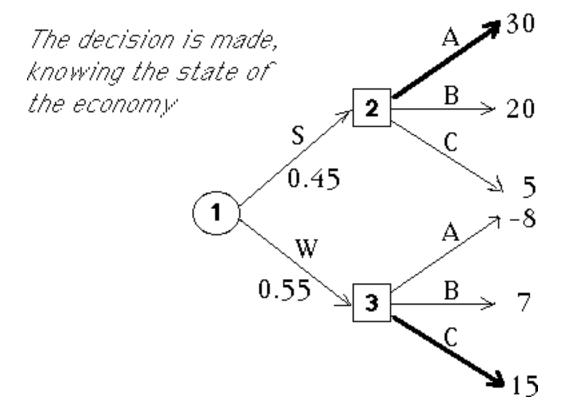
To calculate EVWPI ("Expected Value With Perfect Information"), we draw the decision tree in which the decision-maker has full knowledge of which state has occurred before the decision must be made.

# The Payoff Table

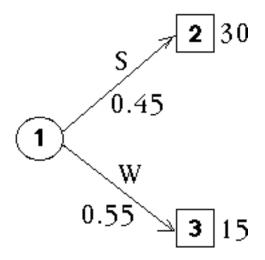
	State of "Nature"	
	S: strong	W: weak
Decision	0.45	0.55
A	30	-8
В	20	7
С	5	15

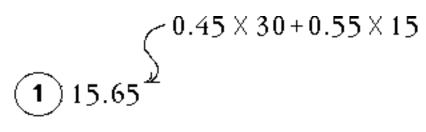
Probability





## Folding back:





#### EVWPI

#### EVPI = EVWPI - EVWOI

### **EXAMPLE**

Farmer Jones must determine whether to plant corn or soybeans on a certain piece of land.

His "payoff" depends upon the weather conditions during the summer growing season:



- If he plants corn and the weather is warm, he earns \$8000
- If he plants corn and the weather is cold, he earns \$5000
- If he plants soybeans and the weather is warm, he earns \$7000
- If he plants soybeans and the weather is cold, he earns \$6500.

In the past,

prior probabilities

40% of all years have been **cold**, and 60% have been **warm**.

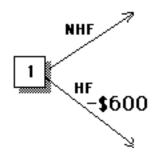
Before planting, farmer Jones can pay \$600 for an expert weather forecast.

If the year will actually be cold, there is a 90% chance that the forecaster will be correct, i.e., predict a cold year.

If the year will actually be warm, there is a 80% chance that the forecaster will be correct, i.e., will predict a warm year.

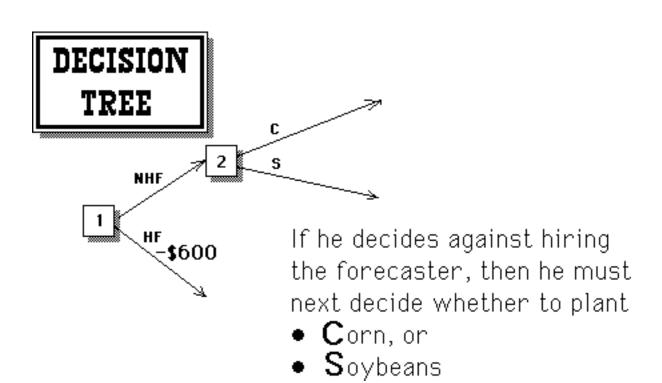
- CONSTRUCTING DECISION TREE
- FOLDING BACK TREE
- P OPTIMAL DECISIONS
- EXPECTED VALUE OF FORECAST
- EXPECTED VALUE OF PERFECT INFORMATION

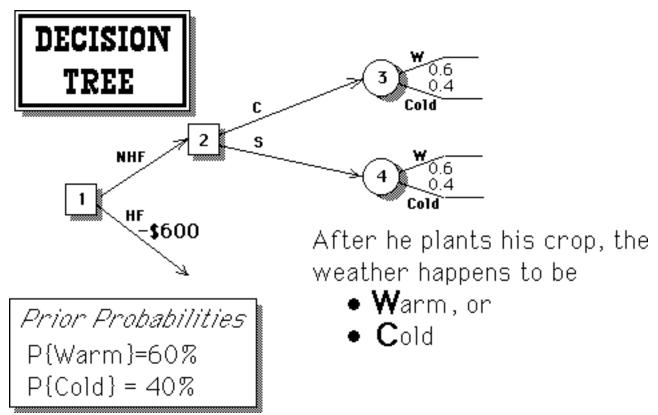


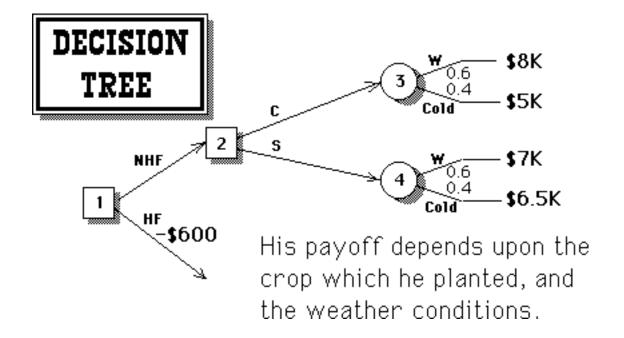


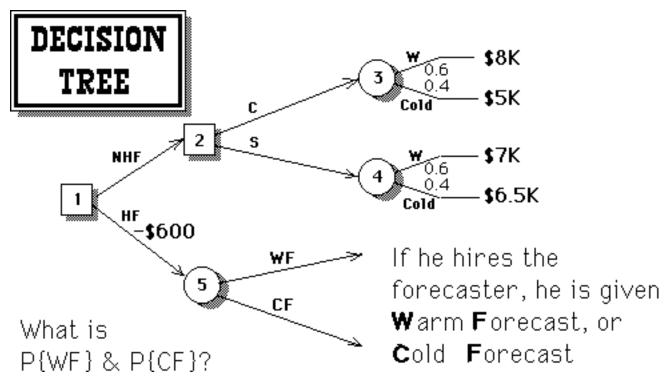
Jones must first decide whether to **H**ire **F**orecaster (**HF**), or **N**ot **H**ire **F**orecaster (**NHF**)











Condition the event "Warm Forecast" on the events "Warm weather" and "Cold weather":

P{Warm Forecast}

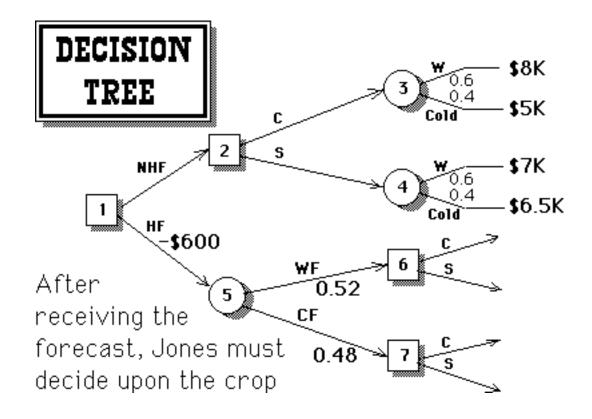
- = P{Warm Forecast|Warm}P{Warm weather} (correct in warm season)
- + P{Warm Forecast | Cold} P{Cold weather}

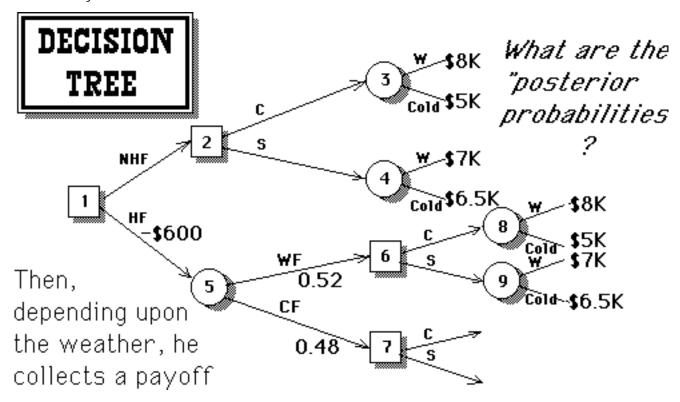
$$P\{WF\} = P\{WF|W\}P\{W\} + P\{WF|C\}P\{C\}$$
  
= 0.8 × 0.6 + 0.1 × 0.4  
= 0.52

#### P{ Cold Forecast}

- = P{Cold Forecast|Warm}P{Warm weather}
- + P{Cold Forecast|Cold}P{Cold weather}

$$P\{CF\} = P\{CF|W\}P\{W\} + P\{CF|C\}P\{C\}$$
  
= 0.2 × 0.6 + 0.9 × 0.4  
= 0.48 = 1 - P{WF}





#### Revised probabilities after receiving forecast

P{Warm weather | Warm Forecast}

$$P\{W | WF\} = \frac{P\{WF | W\} P\{W\}}{P\{WF\}} = \frac{0.8 \times 0.6}{0.52} = 0.9231$$

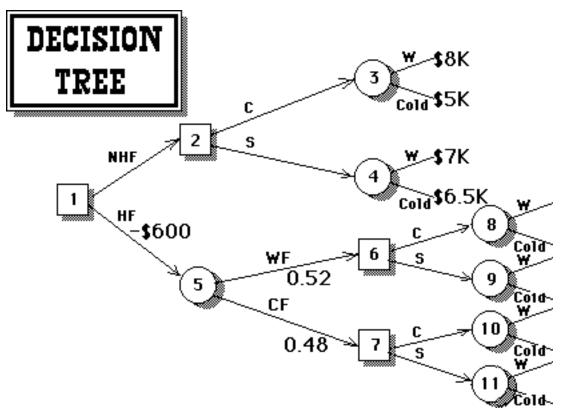
$$Bayes' Rule$$

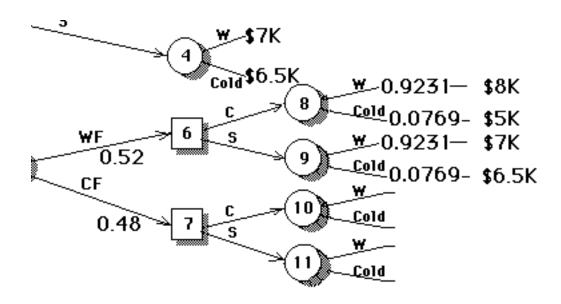
$$posterior$$

$$probabilities$$

P{Cold weather|Warm Forecast}

$$P\{C \mid WF\} = 1 - P\{W \mid WF\} = 0.0769$$





$$P\{W \mid WF\} = 0.9231$$
  
 $P\{C \mid WF\} = 0.0769$ 

#### Revised probabilities after receiving forecast

P{Cold | Cold Forecast} =

$$P\{C \mid CF\} = \frac{P\{CF \mid C\} P\{C\}}{P\{CF\}} = \frac{0.9 \times 0.4}{0.48} = 0.75$$

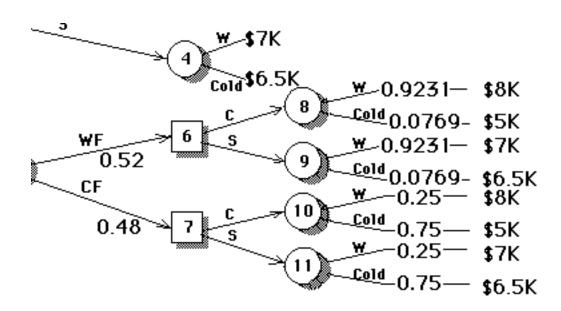
$$Bayes' Rule$$

$$posterior$$

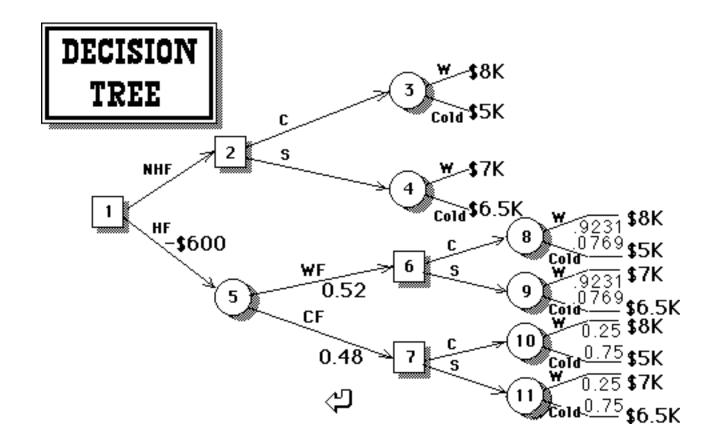
$$probabilities$$

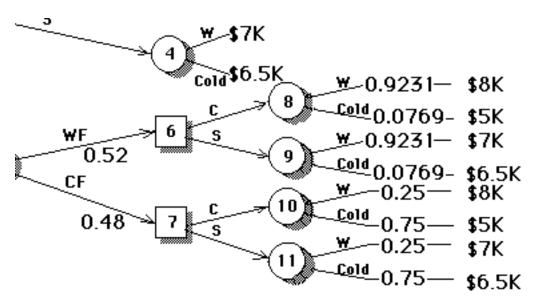
P{Warm | Cold Forecast}

$$P\{W|CF\} = 1 - P\{C|CF\} = 1 - 0.75 = 0.25$$



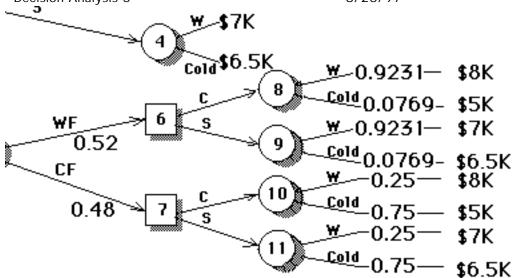
$$P\{C|CF\} = 0.75$$
  
 $P\{W|CF\} = 0.25$ 





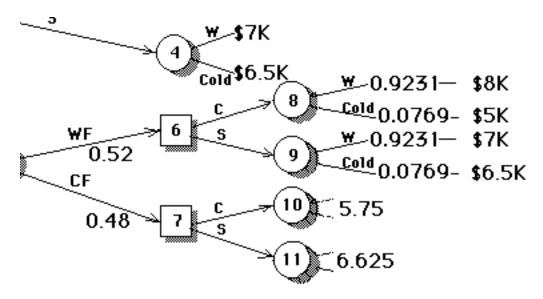
Now we begin "folding back" the nodes of the tree...

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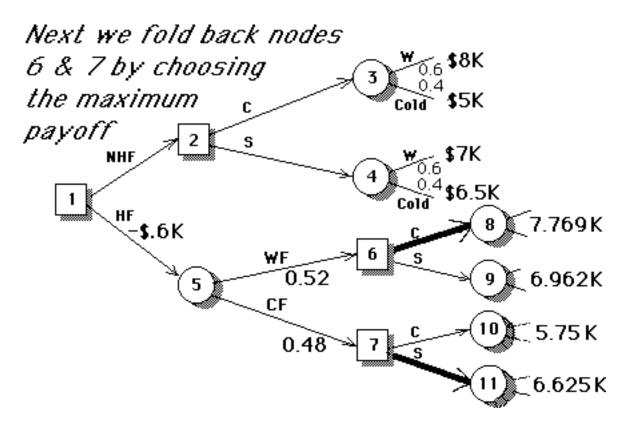
#### Folding back nodes 10 & 11:

$$8(0.25) + 5(0.75) = 5.75$$
  
 $7(0.25) + 6.5(0.75) = 6.625$ 

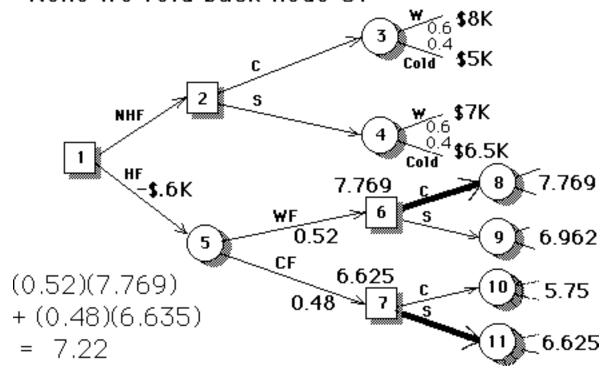


#### Folding back nodes 8 & 9:

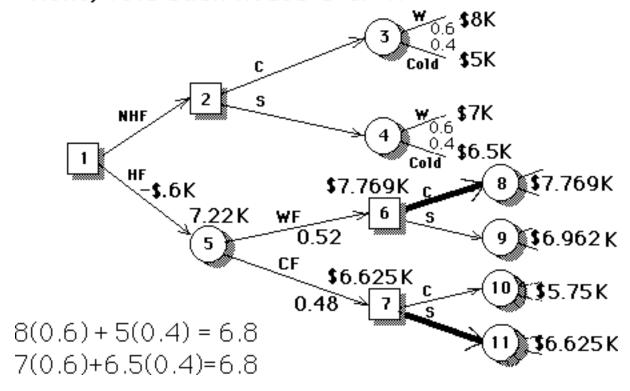
$$8(0.9231) + 5(0.0769) = 7.769$$
  
 $7(0.9231) + 6.5(0.0769) = 6.962$ 

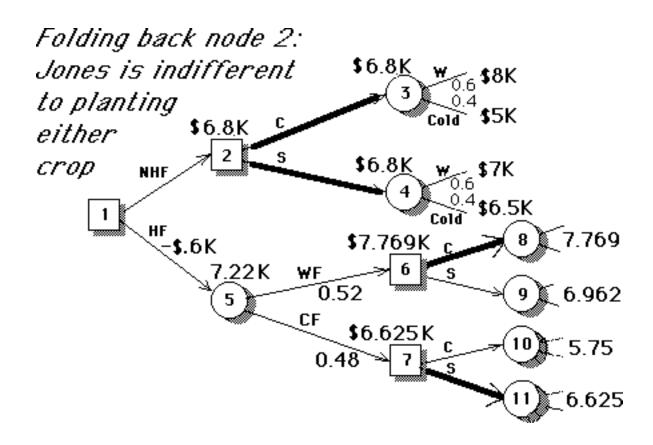


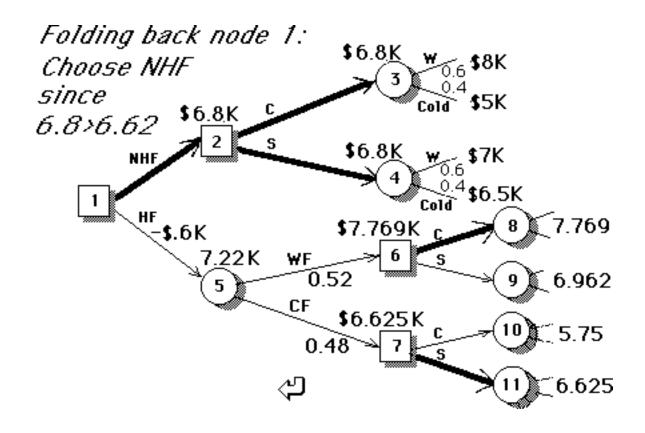
#### Next we fold back node 5:

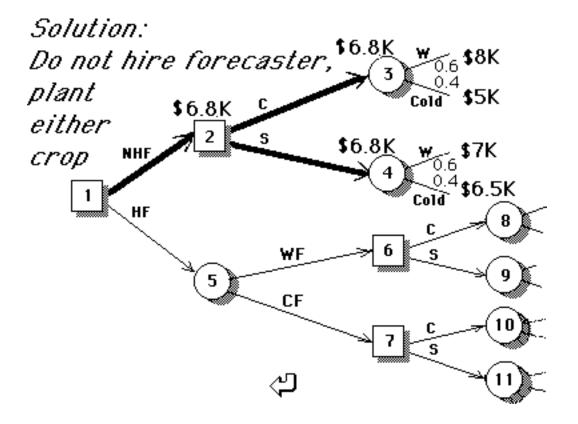


#### Next, fold back nodes 3 & 4:









#### EVSI

What is the expected value of the forecast?

If the forecast were "free", Jones' expected payoff, using the forecast, would be \$7.22K, or \$420 more than his expected payoff without the forecast.

EVSI = \$420

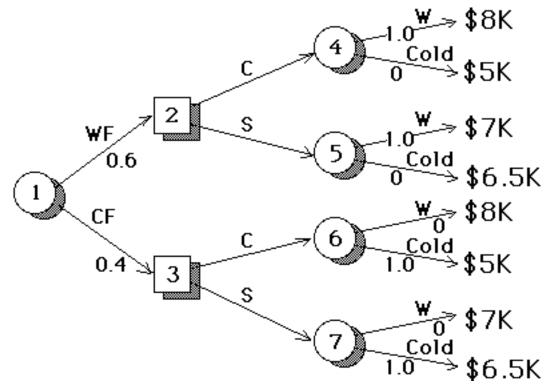


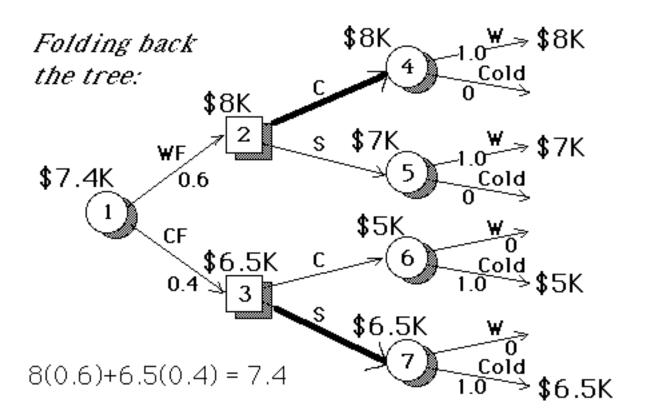
#### EVPI

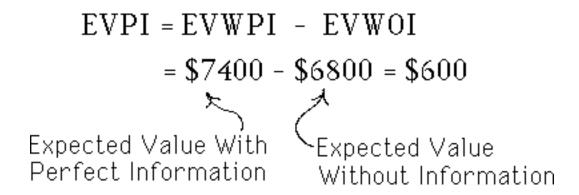
What is the expected value of perfect information?

Imagine that Jones obtained a forecast which was 100% accurate









The NBS TV network earns an average of \$400K from a hit show, and loses an average of \$100K on a flop.

Of all shows reviewed by the network, 25% turn out to be hits and 75% flops.

For \$40K, a market research firm will have an audience view a prospective show and give its view about whether the show will be a hit or flop.

If a show is actually going to be a hit, there is a 90% chance that the market research firm will predict a hit; if the show is actually going to be a flop, there is an 80% chance that the firm will predict a flop.

What is the optimal strategy?

What is EVSI?

What is EVPI?