60615A: Decision Analysis Session 2 - Probabilistic Modeling I

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Outline

- 1 Characterizing uncertainty with probabilities
- 2 Estimating probabilities using an expert

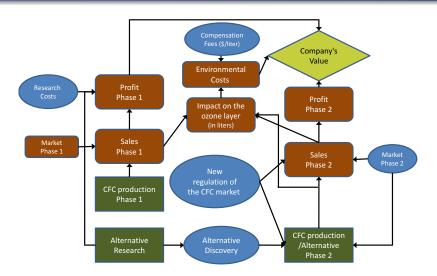
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The presence of uncertainty

- It often happens that certain parameters of a problem are unknown at the moment of making a decision.
 - E.g.: weather, demand for a product, oil prices, amount of time between two volcanic eruptions, etc.
- We say that a <u>parameter is uncertain</u> when it can take different values and that the likelihood of each is computed using probability theory.
 - The uncertainty gives place to the notion of risk: i.e., each
 action has the potential to lead to positive consequences as
 well as negative consequences (with more or less chance).

Example of uncertainty at Dupont



Probability Definitions - Frequentist

- Definition: proportion of times an event will happen if we repeat the experience an infinite number of times.
 - Physical probability: we can study the experience's physical aspect and derive precisely the probability of each event.
 - Example : roll dice without bias, draw a ball from an urn, ...
 - Statistics: we consider the frequencies on a big sample of similar experiences.
 - Example : mortality rates, customer ratings, elections, ...

Probability Definitions - Subjective (Bayesian)

- Definition: the probabilities represent the level of conviction (or belief) of the individual and are updated when new information is available.
- Can be applied to any unconfirmed declaration, whether it relates to a random event or not, repeated or not.
- Example: What is the respective probability of each of the following events?
 - The professor has more than 50\$ in their pockets.
 - The Canadians of Montreal will qualify for the finals this year.

Probability Definitions - The difference

- Consider an urn containing 10 balls. Out of 10 balls drawn randomly from the urn (with replacement), 8 were red.
 What is the probability of randomly drawing a red ball from the urn?
- Frequentist's answer :
 - Necessarily, the proportion is not 0.
 - Other than that, I can't claim anything.
 - I reject more easily the hypothesis that the probability is 0.1 rather than 0.8 (i.e., the proportion is 10% rather than 80%)
- Subjective's answer :
 - Originally, I believed that all proportions were as likely (equal probability on each proportion).
 - After the experience, it is likely to be close to 0.8
 - I perceive that the probability that there is 8 red balls is approximately 0.33
 - However, if the urn was red this probability would be higher.

Fundamental properties of probability

- All probabilities are non-negative and lower or equal to 1.
- 2 The sum of the probabilities of mutually exclusive and collectively exhaustive events is 1.
- The set of conditional probabilities must be coherent with the set of marginal probabilities.

$$\mathbb{P}(A\&B) = \mathbb{P}(A|B)\mathbb{P}(B) \ \Rightarrow \ \mathbb{P}(A|B) = \mathbb{P}(A\&B)/\mathbb{P}(B)$$

Oecomposition of the probability of an event :

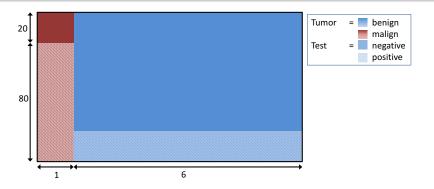
$$\mathbb{P}(A) = \mathbb{P}(A\&B_1) + \mathbb{P}(A\&B_2) + ... + \mathbb{P}(A\&B_n) ,$$

where $B_1, B_2, ..., B_n$ are mutually exclusive and collectively exhaustive events.

Not so easy to avoid violations

- «A certain biopsy test has the reputation of diagnosing properly the benign or malignant nature of a tumor 80% of the time: i.e., a malignant tumor will be accurately identified 80% of the time and a benign tumor will also be, 80% of the time. In general, tumors are 6 times more often benign than malignant.»
- The test just diagnosed a tumor as malignant. What is the probability that the tumor is really malignant?
- Answer: 0.4 not 0.8. Why???

Need to account for basic proportions

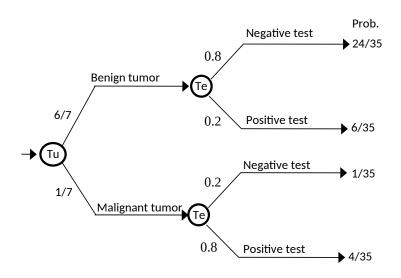


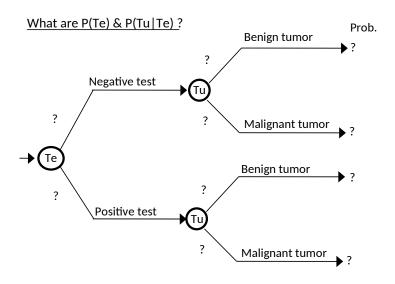
Bayes' theorem

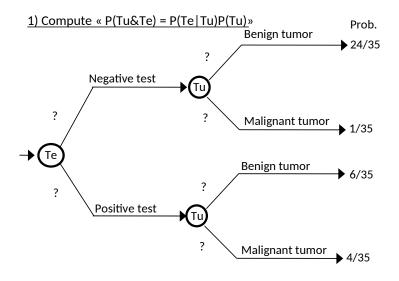
- Bayes' theorem tells us how to account for new information about random variables for which we already had information.
- In a case where we formulated P(A) for $A = A_1, A_2, \dots, A_n$ and receive the new information B
 - Characterize $P(B|A_1)$, $P(B|A_2)$, \cdots , $P(B|A_n)$
 - Apply Bayes' theorem

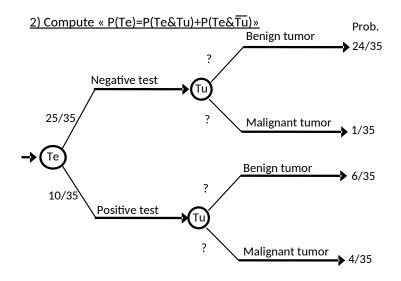
$$P(A|B) = \frac{P(B|A)P(A)}{P(B)}$$

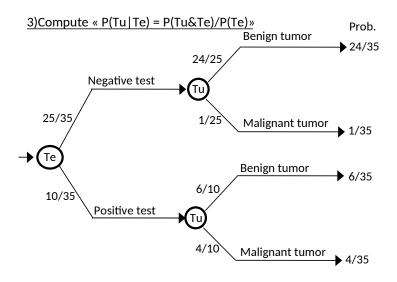
$$= \frac{P(B|A)P(A)}{P(B|A_1)P(A_1) + P(B|A_2)P(A_2) + \dots + P(B|A_n)P(A_n)}$$







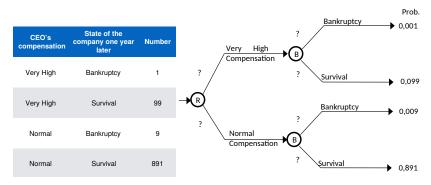




Independence between two events : $A \perp B$

The event B is independent of event A if knowing the status of A doesn't have an effect on the probability that B happens.

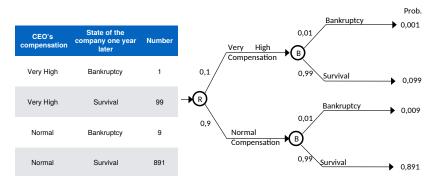
- «A independent of B» implies that «B independent of A»
- E.g.: Is the upcoming bankruptcy of a business independent of the compensation of its CEO?



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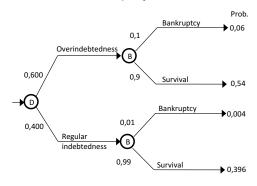
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Dependence between two events : $A \not\perp B$

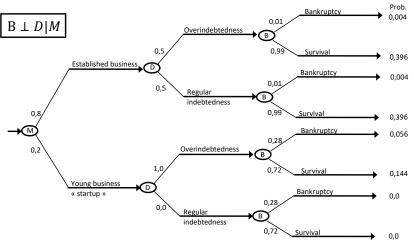
The event A is dependent on the event B if knowing the status of A influences the probability that B happens.

- «A is dependent on B» implies that «B is dependent on A»
- E.g.: Is the bankruptcy of a business dependent on the over-indebtedness of a company?



Conditional independence : $A \perp B \mid C$

The relationship between over-indebtedness and bankruptcy can possibly be explained by the maturity of company :



Conditional independence : $A \perp C \mid B \Rightarrow A \perp C$

Conditional independence doesn't imply non-conditional independence. Take for example the following case :

- Consider throwing three unbaised six-faced dice, let A be the value of die #1, B the sum of dice #1 and #2, C the sum of the three dice.
- $C \perp A|B$ since if we know the sum of the first two dice, the value of die #1 doesn't provide us any additional information about the sum of the three dice.

$$P(C = k|A = i, B = j) = P(C = k|B = j)$$

$$= \begin{cases} 1/6 & \text{if } C - B \in \{1, \dots, 6\} \\ 0 & \text{otherwise.} \end{cases}$$

• $C \not\perp A$ since knowing the value of die #1 informs us of the result of the sum of the three : if A = 6 then C > 8.

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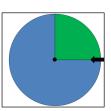
Determining a discrete probability - Verbal method

- You wish to learn the opinion of an expert on the probability that event A happens
- It is greatly discouraged to use the following method :
 - Ask «How probable is event A to happen?»
 - Assign the probability according to the following table :

Answer	Value	
Small chances	0,1	
Not likely	0,15	
I am Doubtful	0,25	
There is a chance	0,5	
It's possible	0,5	
It's likely	0,65	
There is a good chance	0,75	
Most likely	0,85	
Almost certain	0,95	

Determining a discrete probability - Wheel of fortune

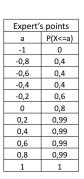
- You wish to learn the opinion of an expert on the probability that event A happens
- Consider two prizes of different values.
 - Prize 1 = Trip to Europe
 - Prize 2 = Nothing
- Present two bets :
 - First Bet :
 - He wins Prize #1 if A happens
 - He wins Prize #2 if A doesn't happen
 - Second Bet (wheel of fortune) :
 - He wins Prize #1 if the arrow stops on the green
 - He wins Prize #2 if the arrow stops on the blue
- Adjust the size of the green region so that the two bets are equivalent



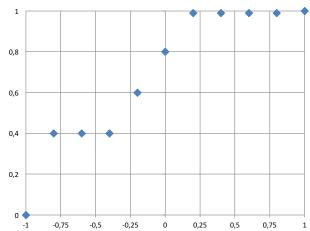
Determining a continuous distribution

- You wish to know the opinion of an expert on the probabilities related to the realization of a random continuous variable Z.
 - Determine the minimal & maximal value of the variable.
 - Discretize the interval [min. value, max. value]
 - 3 For each value $z_1, z_2, ...$, use the wheel of probabilities to determine the probabilities $P(Z \le z_1)$, $P(Z \le z_2)$, ...
 - Trace a continuous and et non-decreasing curve passing through the points found.
 - **6** Confirm the result with the expert.

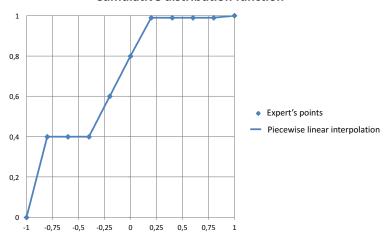
- Do a linear interpolation between known points
 - Advantage: this distribution respects the principle of simplicity (i.e «the most simple hypothesis is the most likely») since it is the one that maximizes the «disorganization» of Z while respecting the known percentiles.
- Fit a parameterized functional form
 - Advantage: the expert can correct the points to be coherent with the distribution they propose.
 - Advantage: allows for the calibration of abstract distribution parameters through the use of interpretable probabilities.

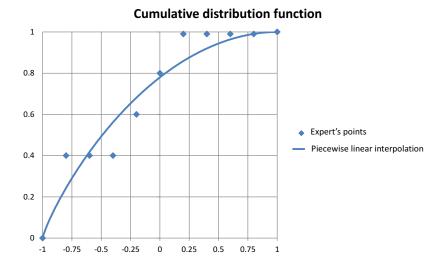


Expert's points



Cumulative distribution function





Mystery bag exercise (see Excel file)

- We present to an «expert» a bag containing a quantity of coins (1¢, 5¢, 25¢, 1\$)
- Objective: Quantify the knowledge of this expert relative to the monetary value of the bag by following this procedure:
 - Establish the min & max of this value
 - Discretize the interval to obtain the values $\{z_1, z_2, \dots, z_n\}$
 - Determine with the help of the wheel of fortune the amounts for which :

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p_1 = P(\text{value of bag} \le z_1), \ p_2 = P(\text{value of bag} \le z_2), \dots, \\ p_n = P(\text{value of bag} \le z_n).
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- Trace a curve passing through the points (z_i, p_i) .
- What is the expected value of the bag?

Survey from session 1

Are you good «experts»???

Quality of a subjective distribution

Two criteria for assessing the quality:

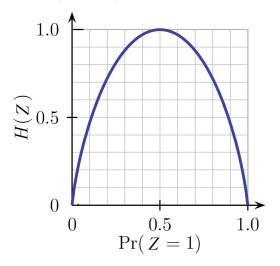
- The distribution should be «informative»
 - The prob. of an event should differ from 0.5, otherwise the expert doesn't know whether the event is more likely to happen or not.
 - The distribution should be concentrated and non uniform.
 - The amount of information is often considered inversely proportional to entropy :

$$H(Z) = \sum_{i} P(Z = z_i) \log_2(1/P(Z = z_i))$$

- The distribution should represent uncertainty authentically
 - In the long term, the events that the expert claimed to have a probability p should occur a proportion of the time that is close to (100 * p)%.
 - This property is also called «good calibration»

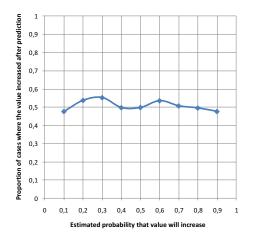
Example of Informativeness

Entropy of a 0-1 (Bernoulli) random variable :



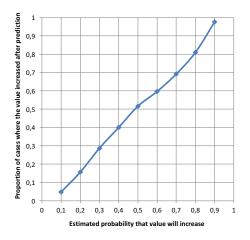
Example of Calibration I

Subjective probability that Microsoft's stock rises next day :



Example of Calibration II

Subjective probability that Microsoft's stock rises next day :



Determining a distribution over many variables

- You wish to know the opinion of an expert regarding the probabilities associated to the realization of many random variables W, X, Y, Z
 - Approach 1 : decompose using conditional probabilities

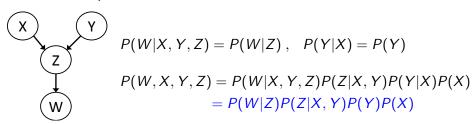
$$P(W,X,Y,Z) = P(W|X,Y,Z) \cdot P(X|Y,Z) \cdot P(Y|Z) \cdot P(Z)$$

 Difficulty: A distribution must be determined for each potential realization of the conditioned variables.

Χ	Υ	Z	P(W=1 X,Y,Z)	X	Υ	Z	P(W=1 X,Y,Z)
0	0	0	?	1	0	0	?
0	0	1	?	1	0	1	?
0	1	0	?	1	1	0	?
0	1	1	?	1	1	1	?

Determining a distribution over many variables

- Approach 2: exploit the «local Markov property» of an influence diagram:
 - If we know <u>neither more nor less than</u> the value of all the variables that influences a variable X directly, then X is independent of all sets of variables that aren't downstream from it.
- Advantage: we can limit the amount of conditioned distribution that need to be estimated.
- Example :



Reduction of the estimation task

Consider that $W, X, Y, Z \in \{0, 1\}$, then the number of probabilities to estimate reduces from 15 to 8 when we exploit the independences described in the precedent diagram.

- With independence :
 - For P(X), P(Y): estimate P(X = 1) & P(Y = 1)
 - For P(Z|X, Y): estimate P(Z = 1|X = 0, Y = 0), P(Z = 1|X = 0, Y = 1), P(Z = 1|X = 1, Y = 0) & P(Z = 1|X = 1, Y = 1)
 - For P(W|Z) : estimate P(W = 1|Z = 0) & P(W = 1|Z = 1)
- Without independence :
 - For P(X): estimate a probability
 - For P(Y|X): estimate 2 probabilities
 - For P(Z|X,Y): estimate 4 probabilities
 - For P(W|X, Y, Z): estimate 8 probabilities