# Heuristic Choice Models

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- Options are not based on dimensional representations
- Instead options are represented as sets of valued aspects
- Frank Restle book 1962
- Tversky 1972 Psych Rev



Pick an aspect

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- If option A contains this aspect and option B does not, then choose option A
- otherwise repeat by picking another aspect and continue until dominance obtained

$$\Pr\left[A \mid \{A, B\}\right] = \frac{v\left(A \cap \bar{B}\right)}{v\left(A \cap \bar{B}\right) + v\left(\bar{A} \cap B\right)}$$

- For multiple alternatives e.g., A, B, C, use round robin tournement
  - pick one pair, e.g. A,B choose one from this set
  - carry first choice over and compare with another option
  - e.g. if A beats B, then compare A with C
  - repeat until reaching final remaining option
  - final choice is the winner of all the binary competitions

### Similarity effects: Paris, Rome, Paris plus a dollar

$$\Pr[P|\{P,R\}] = \frac{v(P \cap \bar{R})}{v(P \cap \bar{R}) + v(\bar{P} \cap R)}$$
$$= \frac{1000}{1000 + 1000} = .50$$
$$\Pr[P\$|\{P,P\$\}] = \frac{v(P\$ \cap \bar{P})}{v(P\$ \cap \bar{P}) + v(\bar{P}\$ \cap P)}$$
$$= \frac{1}{1+0} = 1$$
$$\Pr[P\$|\{P\$,R\}] = \frac{v(P\$ \cap \bar{R})}{v(P\$ \cap \bar{R}) + v(\bar{P}\$ \cap R)}$$
$$= \frac{1001}{1001 + 1000} = .499$$

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- Sach option is represented by a set of valued aspects.
- An aspect is probabilistically selected (with a probability that depends on strength)
- In the selected aspect is eliminated.
- Process continues until only one option remains, which is chosen.

#### EBA Model for N=3 options



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# EBA Model for triadic choices from a set {A,B,C}

• Let K be the sum of the utilities of all of the aspects,

$$K = v(a) + v(b) + v(c) + v(ab) + v(ac) + v(bc).$$

• The probability of selecting A from the set, T, is:

$$\Pr[A|T] = \frac{v(a) + v(ab) \cdot \Pr[A|\{A, B\}] + v(ac) \cdot \Pr[A|\{A, C\}]}{K}$$

- First, aspect a could be selected from the set of aspects with probability v(a)/K. If so, A would be chosen with probability 1.
- Second, aspect *ab* could be selected from the set of aspects with probability v(*ab*)/K. IF so, A would be chosen with probability Pr [A|{A, B}]
- Third, aspect *ac* could be selected from the set of aspects with probability v(ac)/K. IF so, *A* would be chosen with probability  $\Pr[A|\{A, C\}]$

• Pr [A|{A, B}] determined by the ratio of the strength of the aspects in A and not in B relative to strength of aspects in B and not in A

$$\Pr[A|\{A, B\}] = \frac{v(a) + v(ac)}{v(a) + v(ac) + v(b) + v(bc)}$$

• Pr [A|{A, C}] determined by the ratio of the strength of the aspects in A and not in C relative to strength of aspects in C and not in A

$$\Pr\left[A|\{A, C\}\right] = \frac{v\left(a\right) + v\left(ab\right)}{v\left(a\right) + v\left(ab\right) + v\left(c\right) + v\left(bc\right)}$$

### EBA Model for Similarity Effect

- $T = \{$ Lexus, Infiniti, Focus $\}$
- Lexus and Infiniti both share a large number of quality features that are not share by Focus
- Focus has a number of unique economic features that are not shared by Lexus and Infiniti
- Quality features
  - $v_Q(L \cap I \cap F) = 1$
  - $v_Q(L \cap I \cap \overline{F}) = 10$
- Economic features
  - *v<sub>E</sub>* (*L* ∩ *I* ∩ *F*) = 2
    *v<sub>F</sub>* (*L* ∩ *Ī* ∩ *F*) = 12
- Style
  - $v_S(L \cap I \cap \overline{F}) = 3$
  - $v_S(\bar{L} \cap I \cap \bar{F}) = 2$ •  $v_S(L \cap \bar{I} \cap \bar{F}) = 1$

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### EBA Model for Similarity Effect

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$$\Pr[L|\{L,F\}] = \frac{10+3+1}{10+3+1+12} = .54$$
  
•  $\Pr[L|\{L,I\}] = \frac{1}{1+2}$ 

- $\Pr[L|\{L,I\}] = \frac{1}{1+2} = .333$
- $\Pr[I|\{I,F\}] = \frac{10+3+2}{10+3+2+12} = .56$
- $\Pr[L|\{L, I, F\}] = \frac{1+(10+3)\cdot.333}{28} = .19$

• 
$$\Pr[I|\{L, I, F\}] = \frac{2+(10+3)\cdot.666}{28} = .38$$

• 
$$\Pr[F|\{L, I, F\}] = \frac{12}{28} = .43$$

$$\begin{array}{l} v_Q \left( L \cap I \cap \bar{F} \right) = 10 \\ v_E \left( \bar{L} \cap \bar{I} \cap F \right) = 12 \\ v_S \left( L \cap I \cap \bar{F} \right) = 3 \\ v_S \left( \bar{L} \cap I \cap \bar{F} \right) = 2 \\ v_S \left( L \cap \bar{I} \cap \bar{F} \right) = 1 \end{array}$$

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- Payne Bettman Johnson 1993
- Gigerenzer and Todd 1999
- Choose tool adaptively for appropriate environment
- Tools include
  - non-compensatory rules
  - compensatory rules
- Trade off effort for accuracy
  - heuristic rules fast and frugal but may be less accurate
  - optimal rules more accurate but more effortful
- Sometimes simple rules are both less effort and more accurate

## Tools in the box

- Weighted sum rule
  - $U(option) = \sum w_i \cdot v_i$
- Equal weight rule
  - normalize values on common scale, sum values for each option with unit wgt
- Lexico graphic rule
  - Start with most important dimension
  - pick best on most important
  - if there are ties on most important , move to second most important
  - pick best of second most important
  - if ties remain on second most important, move to third most important
  - ect
- Majority of confirming dimensions
  - pick option that is best on the most attributes

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### Tool comparison

- Weighted sum rule
  - compensatory
  - within alternative across dimension processing
  - requires trade offs
  - requires common scale
  - requires process all dimensions
  - more effort but maybe more accurate
  - needed in environments with each dimension equally important

- Lexico graphic rule
  - non compensatory
  - across alternative within dimension processing
  - does not require trade offs
  - does not require common scale
  - processes limited number of dimensions
  - less effort but less accurate
  - works well in environments with one very important dimension