# Building Economic Models of Human-Computer Interaction Part IV

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#### **BUILDING A MODEL**

- 1. Get a precise definition of the problem, and all relevant data about it
  - Identify the factors and variable that may affect the system
    - Uncontrollable factors these are environmental and not under direct control
    - Controllable factors these can be controlled by the system and/or user
  - What factors are stochastic i.e. probabilistic?

 Assume there is some agent, who makes choices to advance their objectives.

- They make choices under constraints.

- So, who are these agents/people, and

   what are they trying to achieve
   (maximize/minimize)?
- What constraints are they under?
   Time, money, knowledge, skills?
- What interactions are available?

– And how can they interact with the system?

Adapted from Varian (1994)

- What benefit do they receive from the choices/interactions they make?
- What costs do they incur from the choices/interaction they make?
- Draw/sketch out what the process is that the person/agent undertakes.
- Consider how the variables/actions relate together.

- 2. Construct a mathematical model of the problem
  - i.e. define the objective function that needs to be minimized or maximized
  - Usually real world problems are very complicated
    - so make a simplified version by
      - Making assumptions
      - Using heuristics
      - And taking approximations.



#### • 3. Solve the model

- This could be done:
  - analytically i.e. mathematically
  - graphically i.e. plotting out the functions
  - via simulation especially if there are stochastic variables

#### • 4. Implement the model

- Put it into practice
- Draw hypotheses from the model

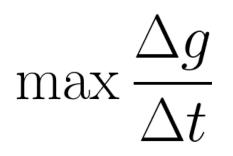
#### **COMPUTATIONAL EXAMPLE**

## **Computational Approach**

- Fix a number of parameters, and then varying one parameter over a range of values
- Plot how the search behavior (outputs) changes in response to the changes in the parameter.
- If some variables are stochastic, then a simulation can be performed, where the computations repeated for different roles of the dice.

• Under IFT, the forager wants to maximize the amount of gain per unit of time.

— i.e.



- Let's assume that we know:
  - The time a forager spends per query (say  $t_q$ )
  - The time a forager spends time per document (say  $t_d$ )

- We know that the total time spent after examining *i* documents is:
  - the query time  $(t_q)$  plus the number of documents examined (i) multiplied by the document time  $(t_d)$

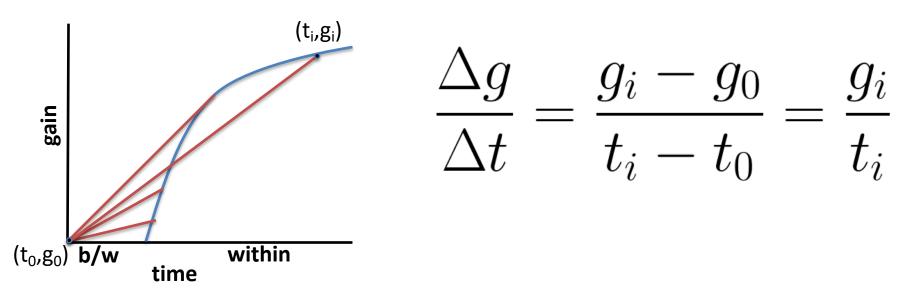
$$t_i = t_q + i.t_d$$

- Next we need to work out the gain received from each document assessed.
- Each document examined yields a certain amount of gain
  - Hmm... how much?

- No idea, so lets play and make something up!
  - Lets say that the 1<sup>st</sup> document gives 4 pieces of information, the 2<sup>nd</sup> document gives 3 pieces, then 2, 3, 1, 1, 0, 0 for the subsequent documents.
  - So the gain of document k is:

k	1	2	3	4	5	6	7	8
g(d <sub>k</sub> )	4	3	2	3	1	1	0	0

 Note that we could of ran a query and got the gain values, or used a function to model the gain.



- The slope of the line can be calculated using the formula above
- When the slope of the line is the greatest, then the gain per unit of time is maximized.

The time at *i* is the query time (*t<sub>q</sub>*) plus the number of documents examined (*i*) times the average time to examine a document (*t<sub>d</sub>*)

$$t_i = t_q + i.t_d$$

 The total gain at *i* is the sum of the gain of all the docs:

$$g_i = \sum_{k=1}^{i} g(d_k)$$

• Let 
$$t_q = 3$$
 and  $t_d = 1$ 

k	1	2	3	4	5	6	7	8
g(d <sub>k</sub> )	4	3	2	3	1	1	0	0
g@i	4	7	9	12	13	15	15	15
t@i	4	5	6	7	8	9	10	11
g/t	1	1.4	1.5	1.7	1.6	1.5	1.4	1.3

- The optimal stopping point is at *i*=4
  - Recall that this assumes that all the patches have a similar distribution of gain.

#### ANALYTICAL EXAMPLE

#### **An Example Gain Function**

t > c $g = k.(t-c)^{\beta}$  $0 \le \beta \le 1$ 

- *t* time spent looking at results.
- g gain received from the results.
- c cost of the query, cost per document is 1.
- *beta* and *k* free parameters controlling the how much and how fast gain is encountered.
- The graphs in the previous slides used k=1,c=2,b=0.5

- To compute the stopping point, we need to construct a tangent line from the origin to the gain curve.
  - Take the first derivative of g(t) to get the slope of the line, and let that equal g over t.

$$\frac{dg}{dt} = k.\beta.(t-c)^{(b-1)} = \frac{g}{t}$$

- Recall that the slope of a line is  $m = (y_1-y_0)/(x_1-x_0)$ , where m is the gradient,  $y_0=0$  and  $x_0=0$ 

• This results in the following criteria:

– The optimal time per patch is:

$$t = \frac{-c}{\beta - 1}$$

– And the gain received is:

$$g = k \cdot \left(\frac{-\beta \cdot c}{\beta - 1}\right)^{\beta}$$

#### **Static Comparatives**

- Fix all variables but one, and see how the outcome is affected.
- What happens to the time in patch:
  - if the cost *c* of querying goes up/down?
  - If the performance beta goes up/down?

$$t = \frac{-c}{\beta - 1}$$

# Insights from IFT's Patch Model

- If the query cost *c* increases, then users will spend more time in the patch (i.e. examine more documents).
- If the rate of gain *beta* increases, then users will spend less time in the patch (i.e. examine less documents).
- If the magnitude of gain k increases, then user behavior does not change, but they receive more gain.

## Summary

- What benefit do they receive from the choices/interactions they make?
- What costs do they incur from the choices the choices/interaction they make?
- Now, work through a simple (the simplest) example possible, so you can see what is going on.
  - i.e. One user, one query...
  - How does it generalize to one user, n queries.
- Always remember to make it as simple as possible (KISS)

Theory is not like a pair of glasses; it is rather like a pair of guns; it does not enable one to see better, but to fight better - Merquior

## SCENARIOS

## **Result Page Exercise**

- How many result snippets should we show the user per page?
  - Assume the user wants to examine *m* snippets, where *m* is likely to be a number greater than 10
  - i.e. we want to set the number of results per page such that the user's costs are minimized.
- Hints:
  - **Draw** up a screen to represent the problem
  - What are the variables of interest/importance?
- More Hints:
  - What are the constraints? What are the main interactions and interaction costs?
  - What if we only showed 1 results per page? Compare that to 2 results per page? Which one is better?

# **App Search**

- On a mobile phone, what is better: to search for the app or to browse through the apps?
- **Goal**: Find app x on a phone in the minimum amount of time.
- What is the optimal number of apps to show per screen?
  - Consider what interactions are associated with searching and browsing.
  - Consider how the time to locate an app on a screen changes with the number and size of app icons.

#### **Extensions to App Search**

- Let's say we swap to a tablet, where the screen size is larger.
  - What is the optimal number of apps to show per screen, now?
- Let's say that that we wanted to evaluate a hierarchy based menu approach for app search?
  - Would this be more efficient?

### **Collaborative Search**

 A student and a supervisor are working on a particular research topic and they need to find around 30-40 references.

– How should they divide their effort?

- Assume that the student's time is cheap, and the supervisor's time is expensive.
- However, the supervisor's search prowess is better than the student's.
- Who should spend more time searching
   And under what conditions?

## **Mobile Search**

- You need to search for some information while walking around the mean streets of Melbourne.
  - Should you type your query?
  - Or use voice and tell your mobile what you want?
  - Consider how long it takes to type/talk, and how easily one can type/talk, and whether the input is correct or not.