

Steering User Behavior With Badges

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Badges

- Have long history
 - Military medals, scouting, primary school, loyalty programs
- Not explicitly competitive
 - Contrast with ranking-based prizes
- Increasingly common online
 - Encouraging participation/contribution
 - Recognizing skills/achievement
 - Some backlash, ‘badge measles’



Social Psychology of Badges

- Broad range of possible individual value to earning badges [Antin and Churchill, 2011]
 - Goal setting
 - Instruction
 - Reputation
 - Status/Affirmation
 - Group identification



Badges and User Behavior

- Do badges affect user engagement and can we better characterize how?
 - Overall level of participation
 - Engagement in certain types of activities
- Can we help site designers define useful badges and systems if badges?
- In this work
 - Develop a model
 - Investigate in context of StackOverflow
 - Consider badge placement problem



Connected components in a graph with 100 million nodes

Move apps to the cloud
without rewriting code.
Once you get it, you'll get it.



Windows Azure

FREE 90-DAY TRIAL



I am trying to get the list of connected components in a graph with 100 million nodes. For smaller graphs, I usually use the `connected_components` function of the Networkx module in Python which does exactly that. However, loading a graph with 100 million nodes (and their edges) into memory with this module would require ca. 110GB of memory, which I don't have. An alternative would be to use a graph database which has a connected components function but I haven't found any in Python. It would seem that Dex (API: Java, .NET, C++) has this functionality but I'm not 100% sure. Ideally I'm looking for a solution in Python. Many thanks.

`python` `graph`

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asked Jun 13 '12 at 13:48



user1453508

27 • 4

1 Answer

active

oldest

votes



SciPy has a `connected components algorithm`. It expects as input the adjacency matrix of your graph in one of its `sparse matrix formats` and handles both the directed and undirected cases.

Building a sparse adjacency matrix from a sequence of `(i, j)` pairs `adj_list` where `i` and `j` are (zero-based) indices of nodes can be done with

Assumptions and Goals of Model

- Assume badges have value to users
- Assume each user has a preferred mix of actions, with a cost to deviate from that mix
- A user trades off between the preferred mix and the goal of winning the badge
- Want to understand effects of badges on overall engagement level and in “steering” user actions

The Model

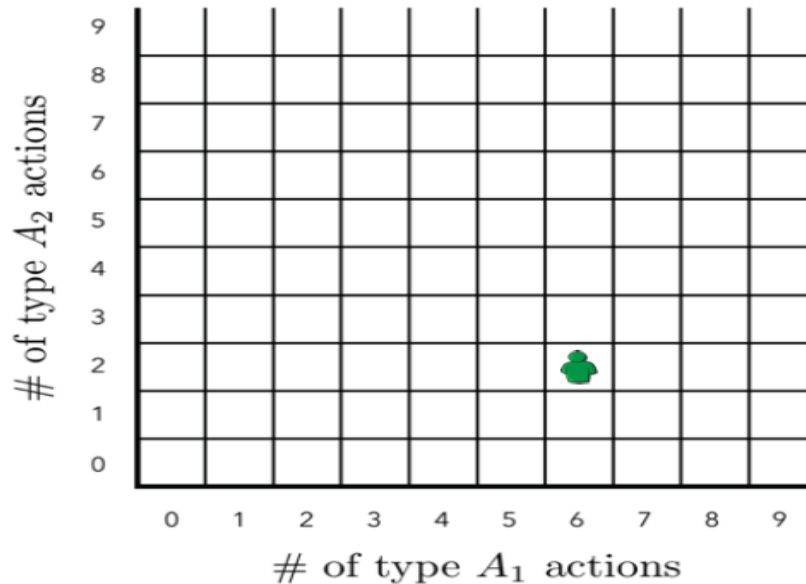
- A population of users and a site designer, with a fixed set of user actions

Ask Q
Answer Q
Vote on Q
⋮



The Action Space

- Action types A_1, A_2, \dots form space denoting number of actions of each type



User Model

- User has a preferred distribution, \mathbf{p} , over action types
- At each time step user picks a probability distribution, \mathbf{p}' , and samples an action from it
- User incurs utility penalty for deviating from preferred distribution

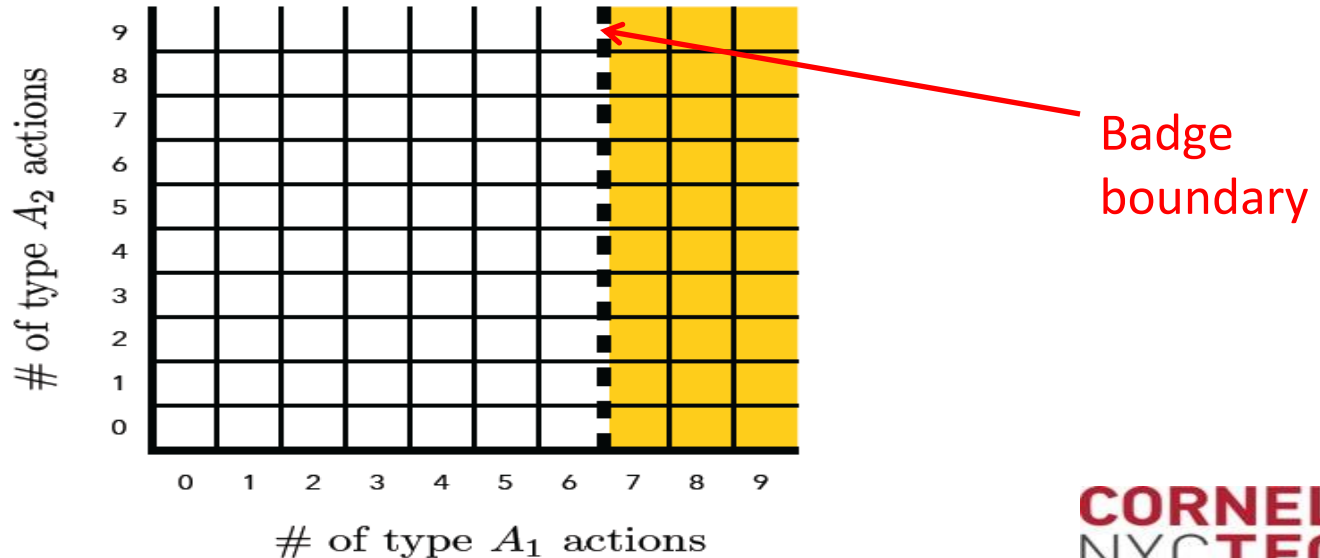
$$g(\mathbf{p}, \mathbf{p}') = \|\mathbf{p} - \mathbf{p}'\|_2^2$$

- User survives to next step with probability Θ (generally $\Theta=.99$)

Badges



- Set of badges, B , each b is subset of cells in action space and has value (utility) V_b



Utility Function

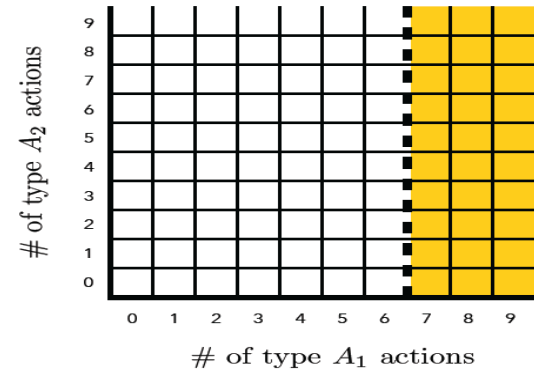
- User's utility composed of 3 parts
 - Value from badges won
 - Cost of deviating from p

Utility from won badges

$$f(\mathbf{a}) = \underbrace{\sum_{b \text{ won}} V_b}_{\text{Value from badges won}} + \underbrace{\theta[\mathbf{p}_a^1 \cdot f(a_1 + 1, a_2) + \mathbf{p}_a^2 \cdot f(a_1, a_2 + 1)]}_{\text{Expected utility of next state}} - \underbrace{g(\mathbf{p}, \mathbf{p}_a)}_{\text{Cost}}$$

Optimization: One Targeted Dimension

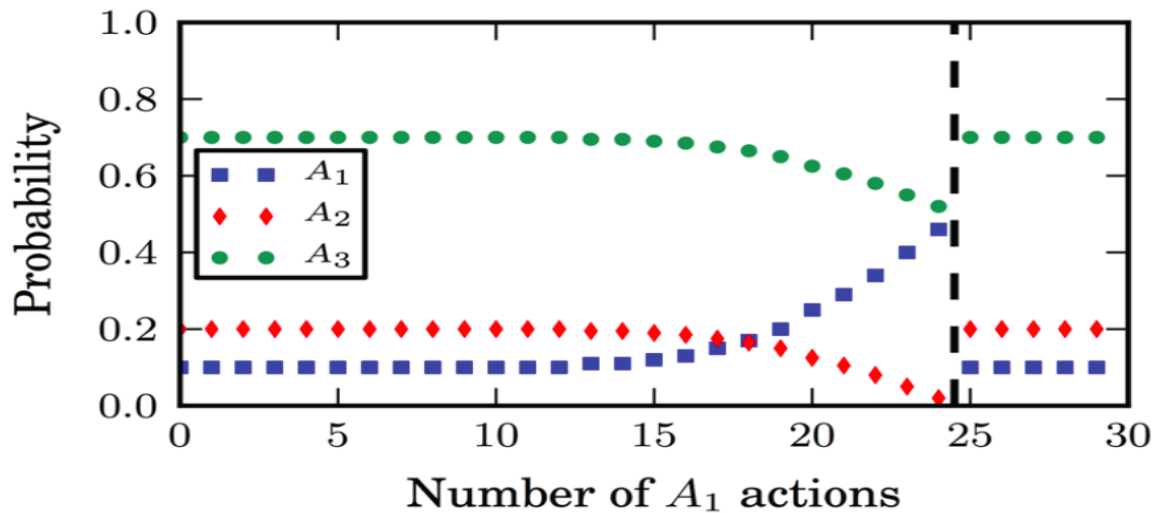
- Use dynamic programming to solve
- No reward for deviating from \mathbf{p} past boundary
- Before boundary select \mathbf{p}_a to maximize expected utility
- Collapses along A_2 dimension
 - 1D problem, solve from boundary back to origin



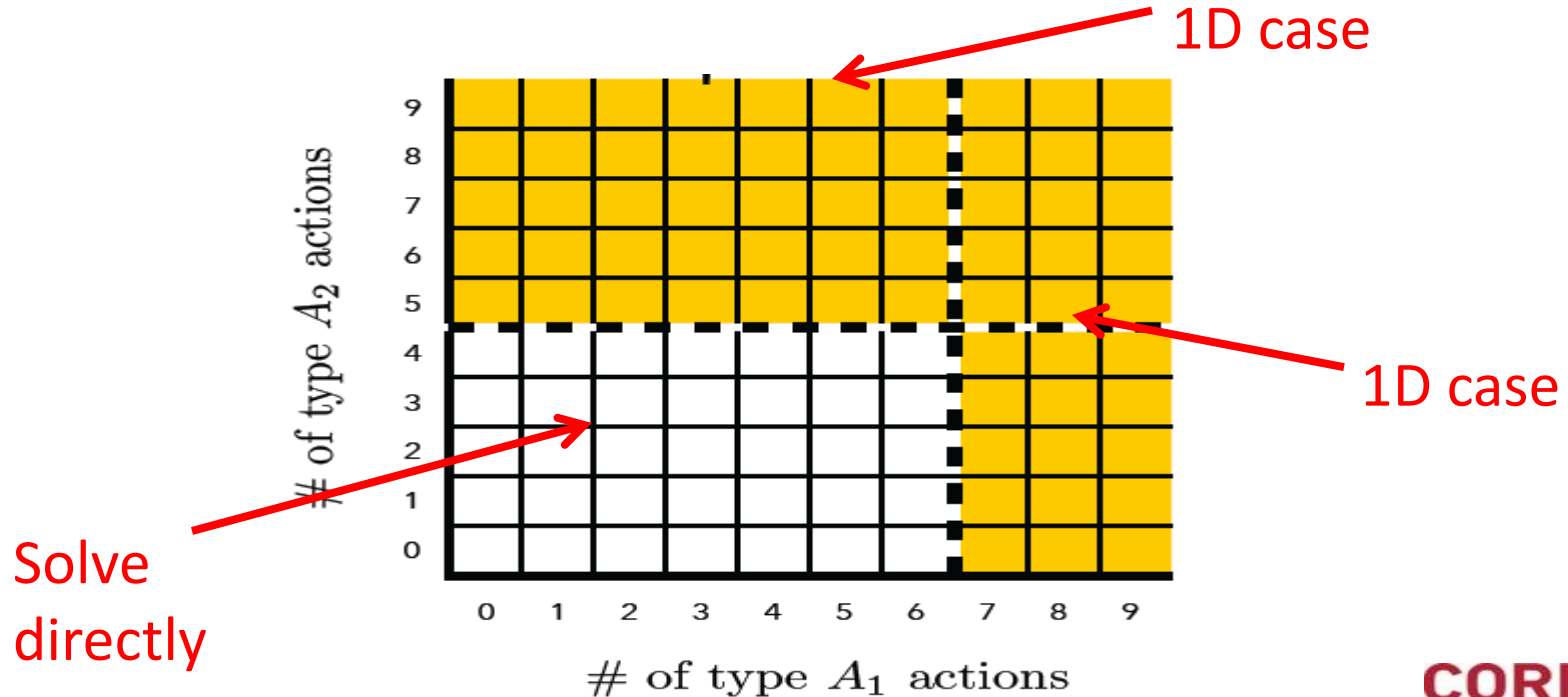
1D Example

- Level of targeted activity accelerates towards boundary

Example: badge at 25 type A_1 actions

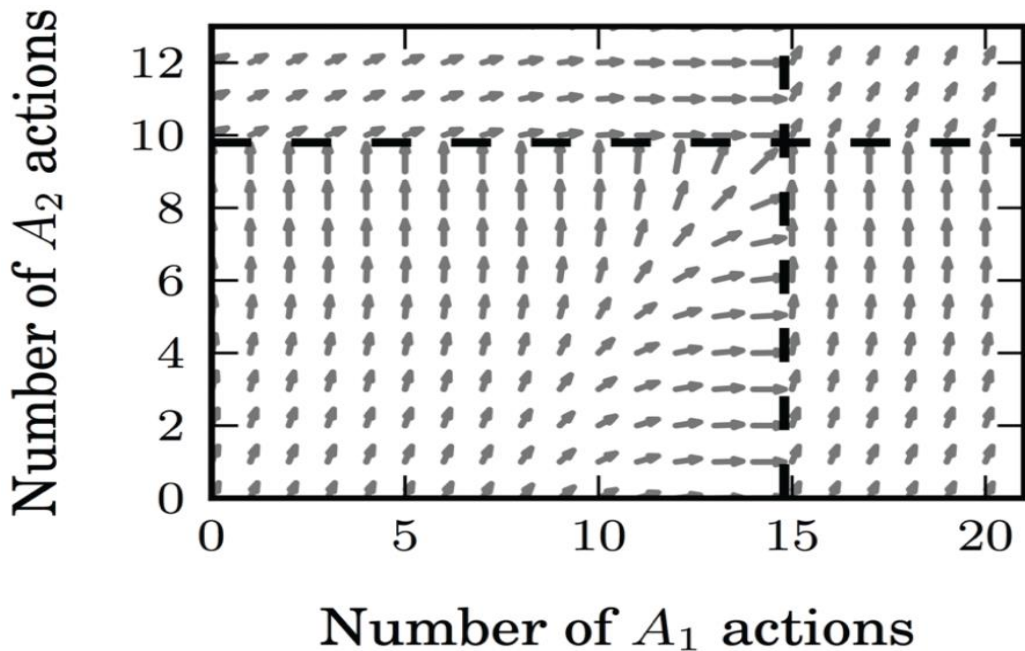


Optimization: Two Targeted Dimensions



2D Example

- Acceleration toward badge boundary



A Limitation of the Model

- Return to baseline, p , after achieving badges
- Does not allow for possible de-motivating effect of achieving a badge
- In practice an external incentive can lower a person's intrinsic incentive/preference
 - E.g., paying for blood donations can reduce the number of donors
- Possible extension, but not seen in our data

StackOverflow Badges

- Extensive use of badges



- Consider two cumulative badges



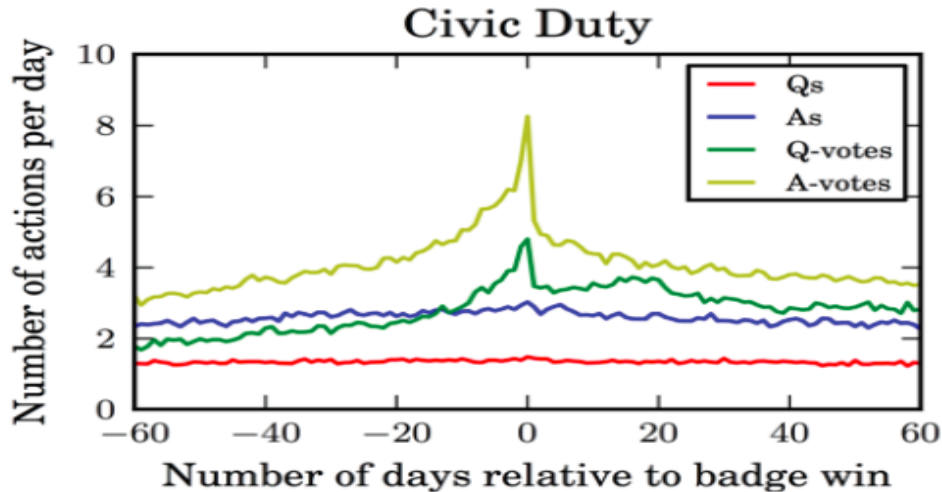
User votes on 600 questions



User votes 300 times

Civic Duty, 300 votes (Silver)

- Qualitatively consistent with model
 - Acceleration towards boundary
 - Increased targeted activity level and overall

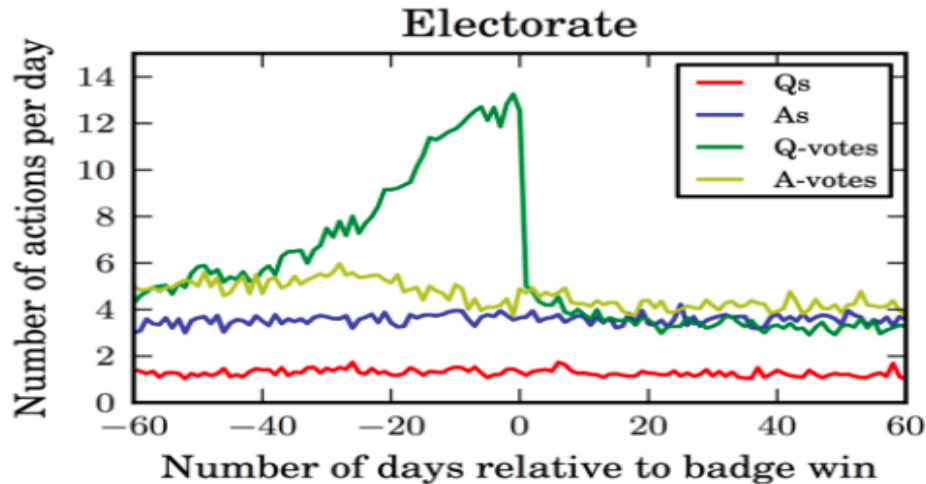


Note: aligned
by day earned

For people
active +/- 60
days from
earning

Electorate, 600 Q Votes (Gold)

- Again qualitatively consistent
 - Single targeted dimension shows not only increased overall level but also tradeoff



Badge Placement Problem

- How should designers “place” cumulative badges to achieve desired effects?
 - E.g., frequent flyer mile status levels, votes on StackOverflow
- Define yield to be fraction of actions over lifetime on targeted dimension
 - Placement to maximize yield

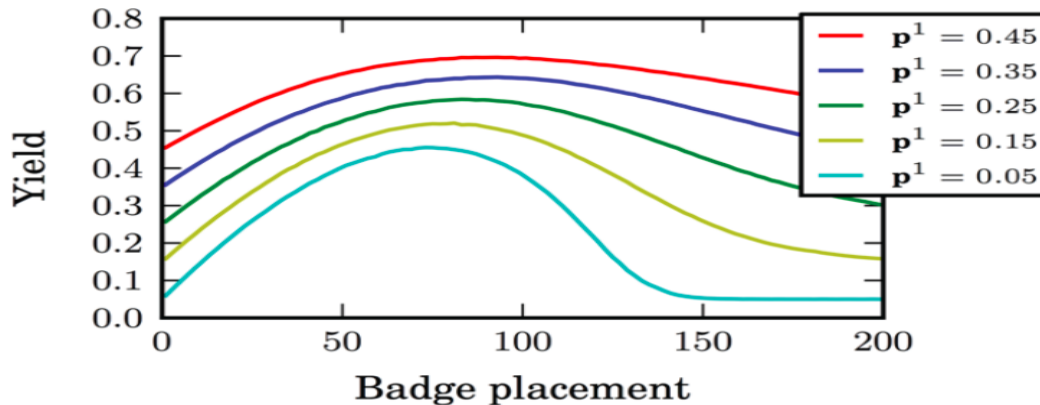


Electorate

User votes on ??? questions

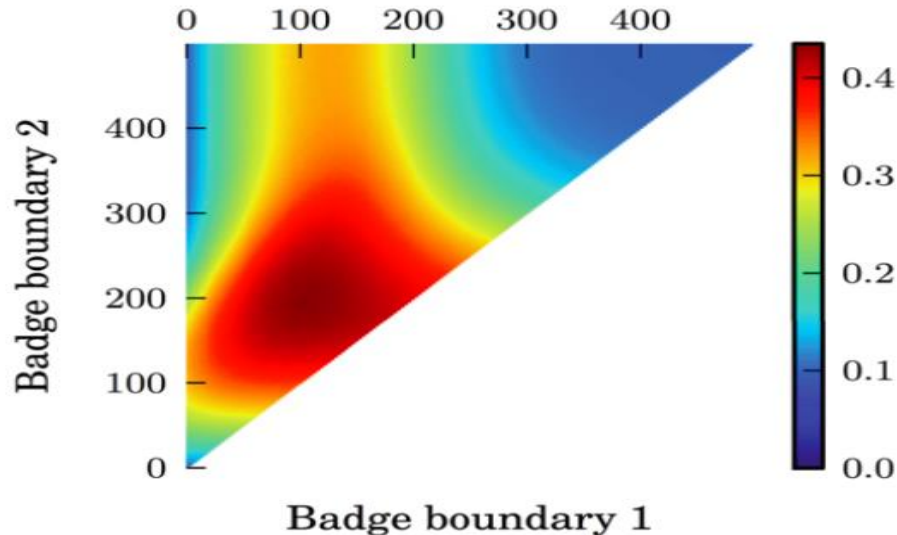
Placing a Single Badge

- Best yielding placements are those which are quite challenging for users to achieve
 - For $\Theta=.99$ expect only 5 targeted actions for $p^1=.05$ yet optimal badge at 75



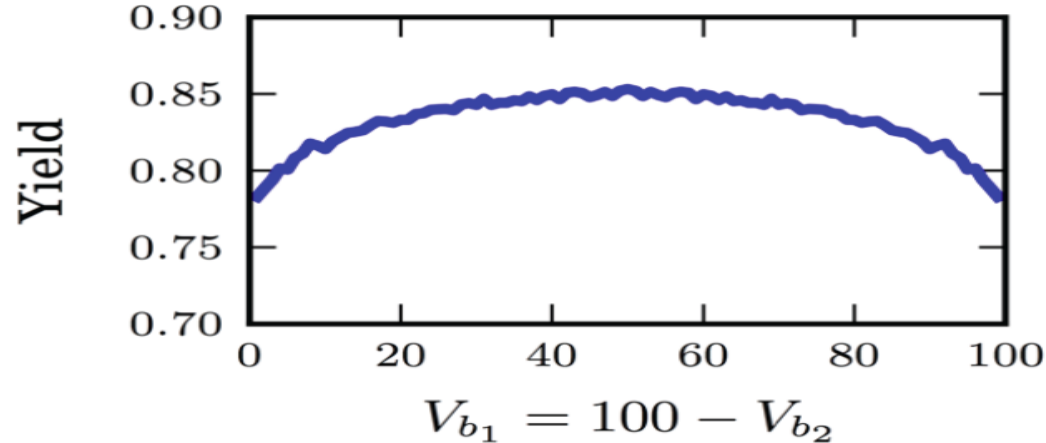
Two Badges on Single Dimension

- Highest yield when badges placed relatively equally apart (illustration for $\Theta=.99$)



Relative Badge Values

- For two badges with fixed total value, best to split value equally



Conclusions

- Introduced model of user behavior
 - Predicts users increase overall engagement and steer distribution actions to achieve badges
- Observe qualitative predictions in StackOverflow data
- Introduced badge placement problem
- Many questions
 - Where value in badges comes from
 - Competition and scarcity in badges
 - Analogies with offline domains
 - Badge system design