

# Introduction to Adaptive Surveys and Amazon Mechanical Turk

Computational Social Science Skills Workshop

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March 2, 2017

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# Motivation

- ▶ **Crowdsourcing:** employers connect with global pool of free-agent workers to complete a specialized or repetitive tasks
- ▶ **Traditionally:**
  - ▶ One way communication
  - ▶ Many tasks in-built on crowd-sourcing clients
  - ▶ Passive sampling
  - ▶ May have biases requester has not controlled for
- ▶ **Adaptive Surveys:**
  - ▶ Two-way communication (more efficient)
  - ▶ Usually not in-built on clients (harder to code)
  - ▶ Active sampling
  - ▶ Biased by design

# Roadmap for this Talk

- ▶ Introduction to Crowdsourcing
- ▶ Introduction to Adaptive Surveys
  - ▶ Quicksort
  - ▶ Multi-armed Bandits
  - ▶ Active Sampling
- ▶ Skill 1: Collect crowdsourced data via MTurk
- ▶ Skill 2: Run a simple quicksort on python
- ▶ GitHub repository for session material:  
[github.com/NandanaSengupta/MTurk\\_Adaptive\\_CSS](https://github.com/NandanaSengupta/MTurk_Adaptive_CSS)

## Introduction to Crowdsourcing

# Crowdsourcing and Mechanical Turk

- ▶ Employers connect with global pool of free-agent workers to complete a specialized or repetitive tasks
- ▶ **Workers:** mostly from US (then India), supplementary income
- ▶ **Employers:** mostly academics and non-profits + few big companies
- ▶ **Academic Impact:** In 2015, 800 published studies used crowdsourced data
- ▶ **Economic Impact:** In 2013, \$2 billion revenue 48 million registered workers (5 million active)
- ▶ Type of tasks
  - ▶ Annotation, Tagging, Classification
  - ▶ Ratings, Comparisons

# Crowdsourcing and Mechanical Turk

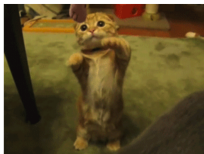
## ► Sentiment Analysis:

This was the best book I ever read!!! Thank you so much! :)

Sentiment expressed by the content:

Strongly Positive
Positive
Neutral
Negative
Strongly Negative

## ► Image Tagging:



Tag 1:

Tag 2:

Tag 3:

What emotion does this GIF invoke:

- ☐ Awesomeness
- ☐ Funniness
- ☐ Sweetness
- ☐ No emotion
- ☐ Sadness
- ☐ Creepiness

## Introduction to Adaptive Surveys



# Adaptive Surveys

- ▶ Two-way communication
- ▶ Fewer potential queries (more efficient)
- ▶ Biased by design
- ▶ Exploration vs Exploitation
- ▶ Where can it be used?
  - ▶ Information tasks where the space of options is very large
  - ▶ When some choices are very informative of whole space
  - ▶ Survey where options space is expanding

Which do you think is better for creating a greener, greater New York City?

Install bike racks on Metro North and LIRR trains.

Re-purpose underutilized retail areas as parks.

I can't decide

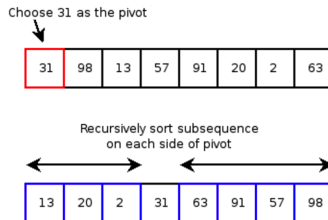
77604 votes on 268 ideas

Add your own idea here...

The image shows a web-based survey interface. At the top, a question asks which option is better for creating a greener, greater New York City. Below the question are two blue buttons with white text: 'Install bike racks on Metro North and LIRR trains.' and 'Re-purpose underutilized retail areas as parks.' Below these buttons is a grey button with the text 'I can't decide'. Underneath the 'I can't decide' button, it says '77604 votes on 268 ideas'. At the bottom of the interface is a text input field with the placeholder text 'Add your own idea here...'. The entire interface is enclosed in a light grey border.

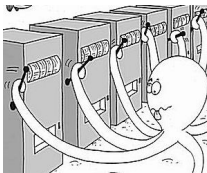
# QuickSort

- ▶ **Objective:** ranking a list of objects  $(k_1, k_2, \dots, k_n)$
- ▶ **Quicksort algorithm:**

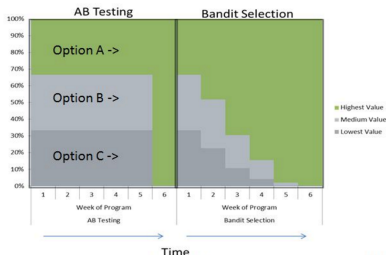


- ▶ **Benefits:** On average requires  $O(n \log(n))$  comparisons  $\ll$  random sampling  $O(n^2)$
- ▶ **In the social sciences?**
- ▶ Ranking of a) streetviews, b) policy options, c) skills

# Multi-Armed Bandits



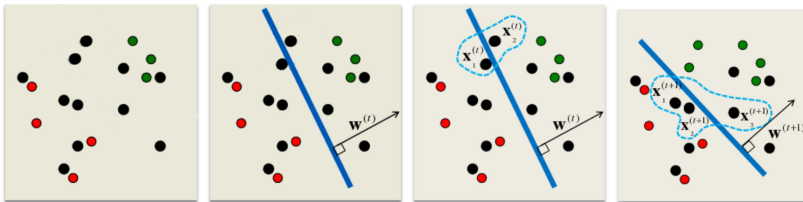
- ▶ Eg: Three options, 'A' 'B' and 'C',
- ▶ If 'A' performs best, probability of selecting 'A' higher than 'B', 'C'.
- ▶ Still random but depends on current best guess of option values.



- ▶ **Disadvantages:** Hard to implement, significant result takes time, can converge to suboptimal
- ▶ In the social sciences?
- ▶ Marketing online products, marketing policies

# Active Sampling

- ▶ With massive unlabeled pool, sometimes cannot afford exhaustive labelling
- ▶ Active sampling: At each time 't', pick 'most uncertain' or 'most informative' items to label



- ▶ **Disadvantages:** Hard to implement, can converge to suboptimal, some comparisons truly uncertain
- ▶ In the social sciences?
- ▶ Estimating individual preferences via pairwise comparisons

Skill 1: Collect crowdsourced data via MTurk

# Introduction to Amazon Mechanical Turk

- ▶ What is M(echanical) Turk?
  - ▶ MTurk is a crowdsourcing internet marketplace. *Requesters* get access to an on-demand workforce (*workers*) to perform surveys and tasks which computers are unable to do.
- ▶ Is this tutorial for Workers or Requesters?
  - ▶ This tutorial is for MTurk Requesters who want to collect and analyze crowdsourced data.
- ▶ What is a HIT?
  - ▶ HIT stands for Human Intelligence Task – these are the tasks requesters design and workers complete.

# Step by Step MTurk Tutorial

- ▶ Step 1: Setup a Requester Account
- ▶ Step 2: Purchase Prepaid HITS
- ▶ Step 3: Create a new survey
  - ▶ Step 3a: HIT properties
  - ▶ Step 3b: Design Layout
  - ▶ Step 3c: Preview and Finish
- ▶ Step 4: Publish HIT
- ▶ Step 5: Track Progress and Manage Results
- ▶ Step 6: Approve Workers and Download Results

Skill 2: Run a simple quicksort on python



## Skill 2: Quicksort demonstration

```
from random import randrange
from random import sample

# given list, sub-list (defined by start and end indices), random pivot index
def partition(lst, start, end, pivot):
    # place pivot at the end of the sub-list
    lst[pivot], lst[end] = lst[end], lst[pivot]
    lst_below_pivot = []
    lst_above_pivot = []
    query_counter = 0

    # dividing into two lists (above and below pivot)
    for i in range(start, end):
        query_counter += 1
        if lst[i] < lst[end]:
            lst_below_pivot.append(lst[i])
        if lst[i] >= lst[end]:
            lst_above_pivot.append(lst[i])

    next_pivot = start + len(lst_below_pivot)

    lst[start : next_pivot] = lst_below_pivot
    lst[next_pivot] = lst[end]
    lst[(next_pivot + 1): end+1] = lst_above_pivot

    return next_pivot, query_counter

def quick_sort(lst, start, end):
    if start >= end:
        return 0
    pivot = randrange(start, end + 1)
    new_pivot, nqueries = partition(lst, start, end, pivot)
    nqueries += quick_sort(lst, start, new_pivot - 1)
    nqueries += quick_sort(lst, new_pivot + 1, end)
    return nqueries

def sort(lst):
    nqueries = quick_sort(lst, 0, len(lst) - 1)
    return nqueries
```

Thanks!