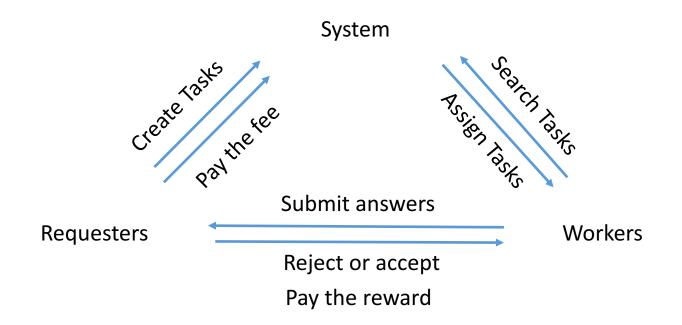
Task Recommendation in Crowdsourcing

You Ming Hsu 2018.10.16

Outline

- Purpose
- Crowdsourcing Recommendation Papers
- Conclusion and Future work
- References

Crowdsourcing Workflow



Amazon Mechanical Turk

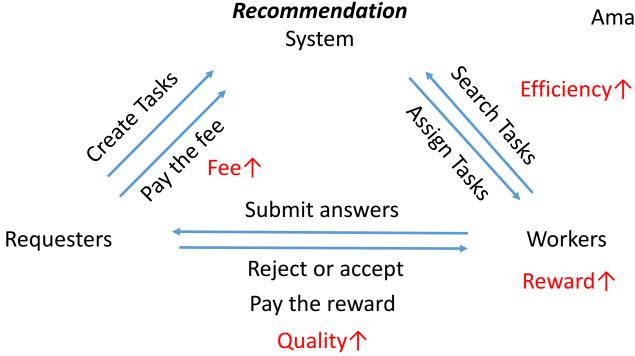
Worker ID: A1PAVCTFXQATXL (2) Hello, YouMing | Sign Out Search All HITs Qualifications amazonmturk HITs Dashboard Filter All HITs Your HITs Queue HIT Groups (1-20 of 691) Show Details Hide Details Items Per Page: Requester Title HITs ▼ Reward Created -Actions \$0.01 Ibotta Turk Are these receipts the same? 23s ago Preview Qualify Input specific values displayed in the image. \$0.02 26s ago amturk Preview Qualify 411Richmond Verify a single value from a receipt 3 \$0.01 20s ago Preview Accept & Work Job Spotter Collect store information of a hiring sign (WARNING: This HIT may contain adult c... \$0.06 Preview 3 27s ago Qualify Studio71 Tech VIDEO - Identify potentially offensive videos %s(between 17 & 30 minutes in length) \$0.75 39s ago Preview Qualify Transcribe a VIN Marina Martin \$0.03 28s ago Preview Accept & Work Ibotta Turk Are these receipts the same? \$0.01 32s ago Preview Qualify Time Allotted Qualifications Required Your Values Description Tell us if two receipts are the same 6 Min ✓ HIT approval rate (%) is greater than 95 100 **Expires** X Total approved HITs is greater than Requirement not in 18h 5000 met

Amazon Mechanical Turk

Worker ID: A1PAVCTFXQATXL 2 Hello, YouMing | Sign Out Search All HITs amazonmturk Dashboard Qualifications Filter Filter results: HITs that Reward Creation date: oldest first I'm qualified to work on Pays at least \$ 0.01 Creation date: newest first Require Masters Qualification Reward amount: lowest first Reward amount: highest first Clear | Reset Cancel HITs: least first HITs: most first 411Richmond 20s ago Preview Accept & Work Qualify VIDEO - Identify potentially offensive videos %s(between 17 & 30 minutes in length) Qualify 28s ago Preview Accept & Work Qualify ✓ HIT approval rate (%) is greater than 95 ★ Total approved HITs is greater than 2018/10/25

Purpose

Crowdsourcing Workflow



In 2011, a writer in San Francisco, claimed that he earned a measly USD\$ 4.38 for spending eight hours in a day crowdsourcing for Amazon Mechanical Turk.

Bag-of-Words Approach

 The bag-of-words approach uses the vocabulary of the task description and computes the similarity as the overlap in their vocabularies.

$$bow(t) = \frac{1}{|H|} * \sum_{i=1}^{|H|} c_i * |Voc(t_i) \cap Voc(t)|$$

- H = {(t, c)}, the history of a specific user.
- t is the task and its associated features.
- c indicate the scale of preference of the user for the task, here we only consider a binary preference, and therefore $c \in \{-1, 1\}$.

Classification Based Approach

- The bag-of-words approach can not incorporate the other features of a task like timestamp, reward etc.
- Using maximum entropy classifier, the classifier probability can be defined as:

$$Pr(c_i|t) = \frac{1}{Z(t)} exp\left(\sum_{j=1}^{n} \lambda_j f_{ij}(c_i, t)\right)$$

- t is a task and associated features.
- c_i is the class, where c = 1 indicates a user will be interested in the task and c = -1 otherwise.
- f_{ij} are feature functions and Z(t) is a normalizing factor.
- λ_i are the weights for the feature functions.

TaskRank

• Define a score of a task t_{ij} in category c_i for a worker v_v is given as follows:

$$T\mathcal{R}_{i,j}(v_y) = \mathcal{A}\mathcal{R}_i(v_y) * \mathcal{CPS}_i(v_y) *$$

$$\left(1 - \left| \frac{m_{i,j} - \mathcal{RPS}_i(v_y)}{\mathcal{RPS}_i(v_y)} \right| \right) *$$

$$\left(1 - \left| \frac{e_{i,j} - \mathcal{TPS}_i(v_y)}{\mathcal{TPS}_i(v_y)} \right| \right)$$

- $AR_i(v_y) = \frac{|T_i'(v_y)|}{|T_i(v_y)|}$, the acceptance rate of tasks in category c_i for worker v_y .
- $CPS_i(v_y) = \frac{|T_i(v_y)|}{\sum_{p=1}^{C_N} |T_p(v_y)|}$, task category preference score on category c_i for worker v_y .
- $RPS_i(v_y) = \frac{\sum_{j=1}^{T_N} m_{ij}}{|T_i(v_y)|}$, reward preference score on category c_i for worker v_y .
- $TPS_i(v_y) = \frac{\sum_{j=1}^{T_N} e_{ij}}{|T_i(v_y)|}$, time alloted preference score on category c_i for worker v_y .

Recover the worker-task preferring matrix

• To transform workers' behaviors into values as follows:

Worker Behavior		Value
Worker's work done is accepted by requester.	\longrightarrow	5
Worker's work done is rejected by requester.	\longrightarrow	4
Worker completes a task and submits the work done.	\longrightarrow	3
Worker selects a task to work on but not complete it.	\longrightarrow	2
Worker browses the detailed information of a task.	\longrightarrow	1
Worker does not browse the detailed information of a task.	\longrightarrow	0

Yuen, M. C., King, I., & Leung, K. S. (2012, August). Task recommendation in crowdsourcing systems. CrowdKDD Yuen, M. C., King, I., & Leung, K. S. (2014). TaskRec: A Task Recommendation Framework in Crowdsourcing Systems. Neural Processing Letters

Recover the worker-task preferring matrix

Example for worker-task matrix:

	v_1	v_2	v_3	v_4	v_5	v_6
w_1		5	2		3	
w_2	4			3		4
w_3			2			2
w_4	5			3		
w_5		5	5			3

- Using Probabilistic Matrix Factorization (PMF) to recover this matrix.
- Select the top few of the tasks for recommendation.

Yuen, M. C., King, I., & Leung, K. S. (2012, August). Task recommendation in crowdsourcing systems. CrowdKDD Yuen, M. C., King, I., & Leung, K. S. (2014). TaskRec: A Task Recommendation Framework in Crowdsourcing Systems. Neural Processing Letters

Conclusion and Future work

- Ranking of all tasks to recommend, based on the several factors such as worker history, interest, performance and requester's feedback.
- If a worker gets a list of related recommended tasks:
 - Improving the quality of the task.
 - Increasing the solution efficiency.
 - Worker and system may get more reward.
- These kind of worker history data is difficult to obtain.
 - —> Synthetic recommendation system
 - Which can evaluate the performance between different methods.

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