


Crowdsourced Data Management Survey



Group 13
The University of Melbourne

Overview

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 - b. Our Project Goal
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 - i. Application & Platforms
 - ii. Techniques
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Team & Project Goal

Project Goal:

- 1. Compose a survey to help the general public understand Crowdsourced Data Management.
- 2. Illustrate the survey results through presentation.

Communication & Research Tools

Manage weekly meetings & tasks

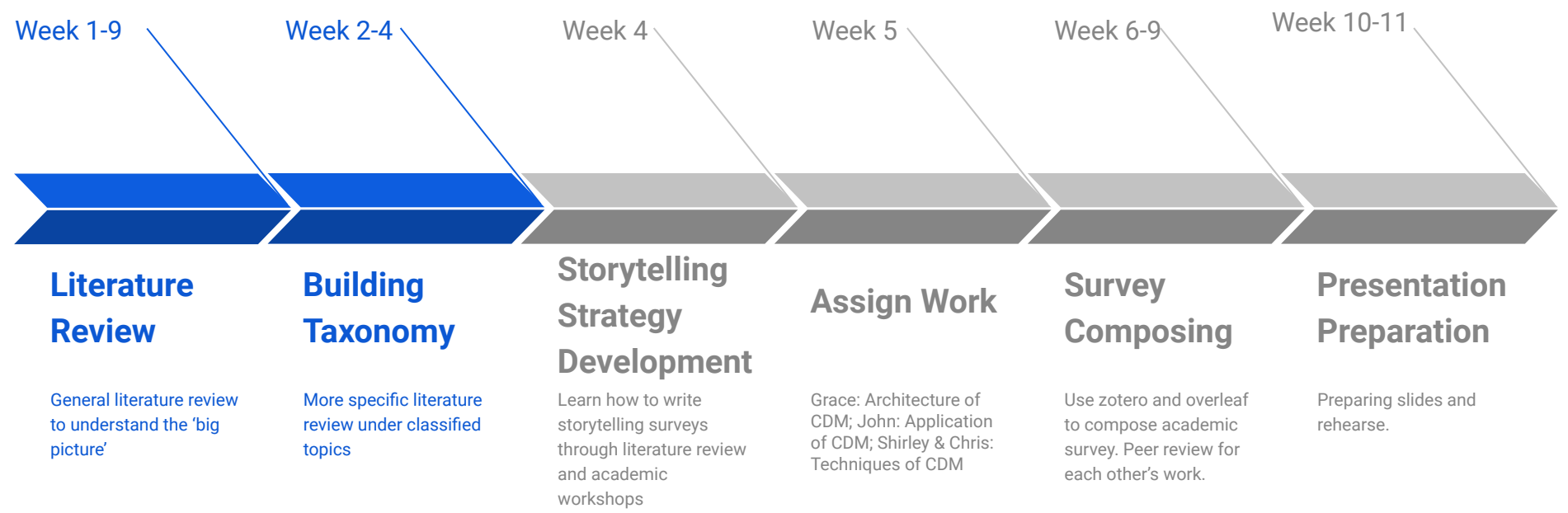


Manage report and reference



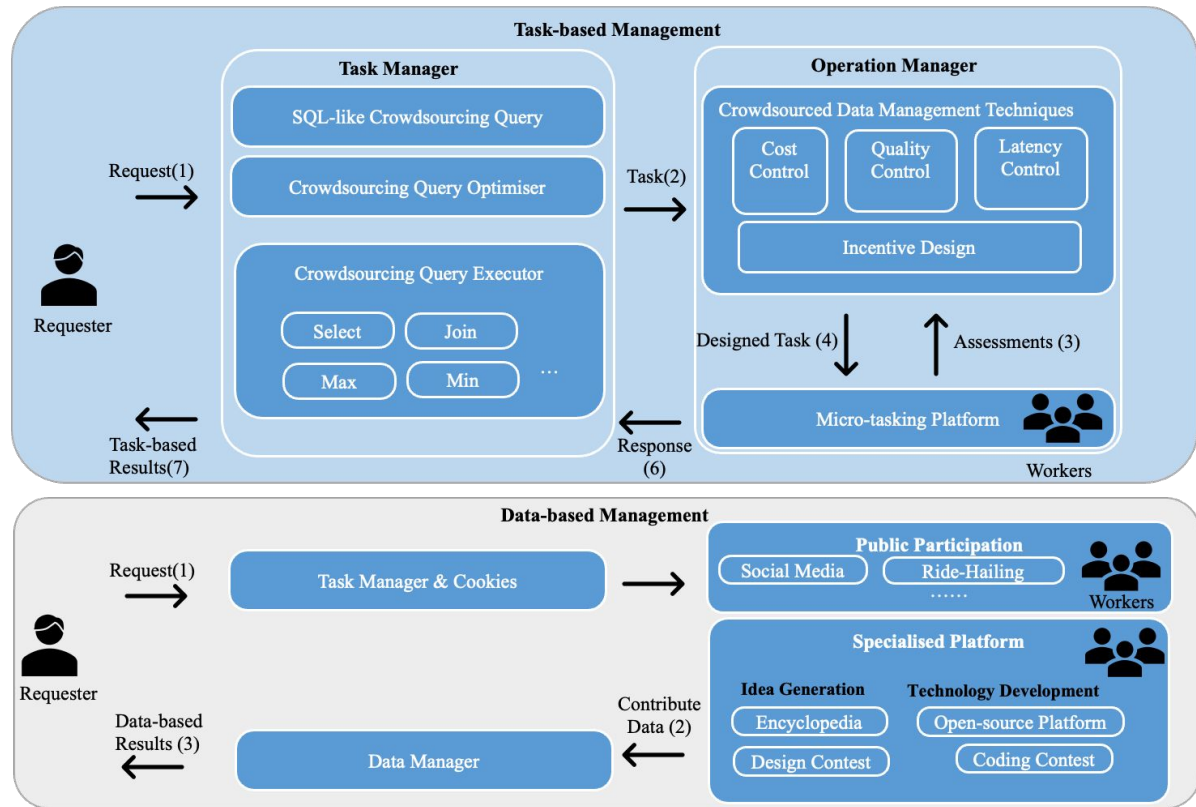
Project Timeline

Approx.10 hours per week

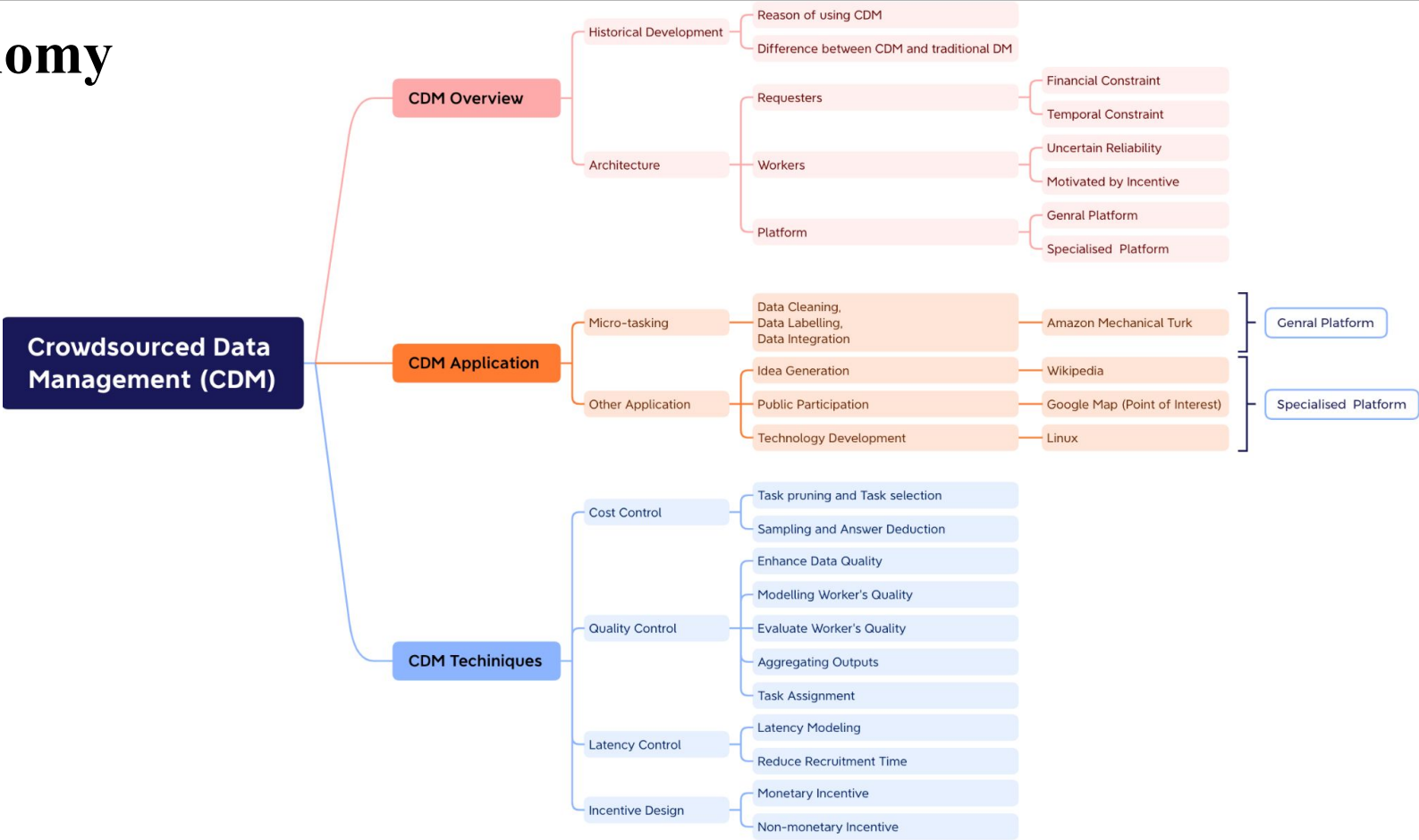


What is Crowdsourced Data Management? (Ellis, 2004; Howe, 2006)

- 1714 ● Longitude Prize
- 1936 ● Toyota Logo Design
- 2005 ● Launch of Amazon Mechanical Turk
- 2006 ● Conceptualisation of “Crowdsourcing”
- 2000s ● Crowdsourcing Applications & Technology Explosion
- 2010s ● Crowdsourced Data Management Applications & Techniques**
- 2020s ● Nowadays Application



Taxonomy



Platform (Chittilappilly et al., 2016)

General-purpose platform

- Amazon Mechanical Turk



Specialised platform

- Kaggle
- Lego Ideas



General-purpose Platform



Amazon Mechanical Turk

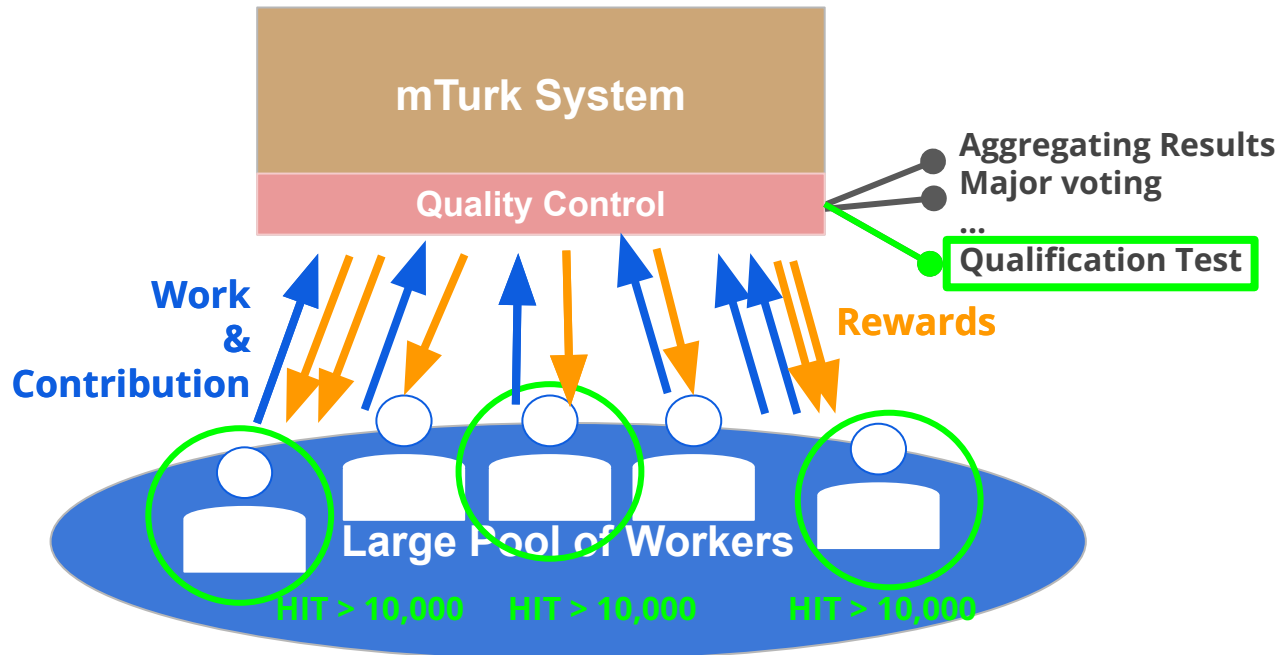
- Requester
- Worker
- Developer

Cost Control

- Multiple assignments
- Iterative Learning
- ...

Latency Control

- Monetary Rewards



General-purpose Platform



Microtasks in AMT

- Data Labelling
- Data Cleaning
- Data integration

Bounding Box

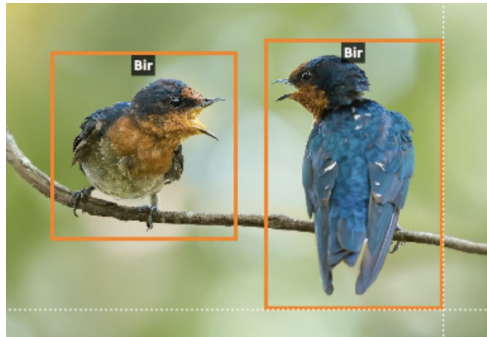


Image Labelling



Select an option

Cat	1
Dog	2
Bird	3
None of the Above	4

Survey

What is your favorite color for a bird?

example: pink

Check this box if you like birds



On a scale of 1-10, how much do you like birds?



Specialised Platform



Kaggle

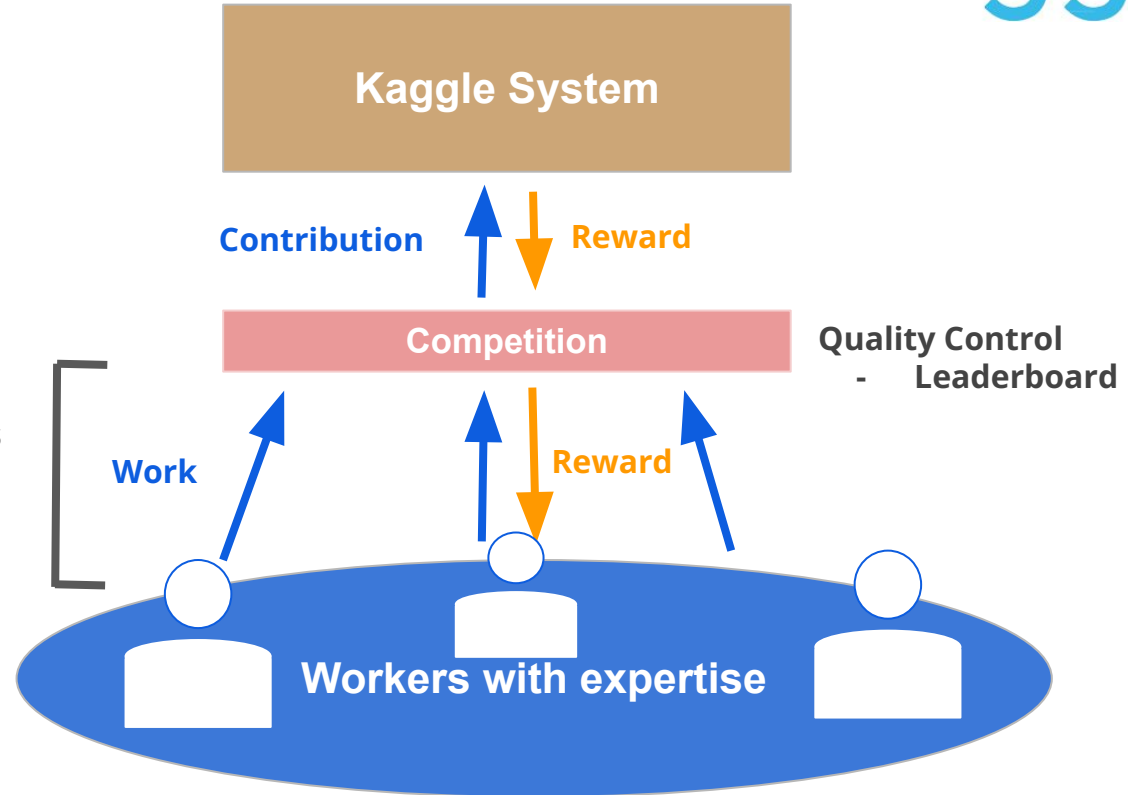
- Sponsors
- Workers

Cost Control

- Multiple assignments

Latency Control

- Monetary Rewards



Quality Control

- Data quality may be influenced by several factors
- Various approaches have been proposed to guarantee high quality results
 - ❖ Enhancing data quality
 - ❖ Selection of workers
 - ❖ Aggregating outputs of different workers
 - ❖ Assigning suitable tasks to high quality workers





Quality Control	Pros	Cons
Aggregating Result	<ul style="list-style-type: none">• Produce more accurate and reliable overall results• Assist identifying malicious or low-quality workers	<ul style="list-style-type: none">• Requires more workers than other methods• Higher cost
Majority Voting	<ul style="list-style-type: none">• Easy to implement• Robust to noise	The professional level of different workers and task difficulty is ignored
Confusion Matrix	Capture more information	<ul style="list-style-type: none">• Not applicable in all situations• Biases in ground truth.
Golden Task	Simplify the process of assessing worker quality.	<ul style="list-style-type: none">• Higher cost from hiring experts• Difficulty in deciding sufficient number of golden tasks• If the answer is leaked, or many requesters use the same golden task, the mechanism will fail
Qualification Task	Improves confidence of result's quality from reliable workers	<ul style="list-style-type: none">• Many workers are unwilling to answer “extra” tasks for free.• Poses potential risk that spammers could carefully label these golden tasks to increase their reputation.

Table 3: Pros and Cons for Task Assignment Techniques

Techniques	Pros	Cons
Multiple assignment	<ul style="list-style-type: none">• No need to know or infer worker’s reliability.• Easy to implement.	<ul style="list-style-type: none">• Not efficient: workers didn’t get potentially acquainted tasks.
Iterative Learning	<ul style="list-style-type: none">• Worker’s reliability estimated based on comparing with other’s answers.	<ul style="list-style-type: none">• Increasing cost.• High computational cost.
Dual Task Assigner	<ul style="list-style-type: none">• Leveraged resource usage.• Worker’s reliability can be estimated from their previous performance.	<ul style="list-style-type: none">• Difficult to infer worker’s reliability when there’s no enough historical data.
Real-time Task As- signer	<ul style="list-style-type: none">• Leveraged resource usage.• Reduced cost on multiple assignments.	<ul style="list-style-type: none">• Only near-optimal results can be achieved.
Collaborative Task As- signment	<ul style="list-style-type: none">• Flexible with dynamic task assignments.• Worker’s reliability is known.• More flexible task assignment with worker’s domain knowledge.• Can be applied in Knowledge-Intensive and collaborative crowdsourcing settings.• Application is narrow in real context, as most times worker’s skill set is unknown.	
Task Assignment with AI Planning	<ul style="list-style-type: none">• Leveraged task standardisation process• Enabled testing of task allocation strategies with different scenario variables	<ul style="list-style-type: none">• Did not perform optimisation under consider budget.

Cost Control

- Keeping costs under control without compromising the quality of the results
- Techniques:
 - ❖ Fix the number of workers per task based on the requester budget
 - ❖ Reduce the number of tasks performed by workers
 - Task Pruning
 - Task Selection
 - Sampling
 - Answer Deduction



Comparison

Cost Control	Pros	Cons
Fixing number of workers	Easy to implement	Wasted expenses
Task Pruning	Significantly saves labour costs while maintaining high quality	<ul style="list-style-type: none">• Cost cannot be reduced on a per-task level• Limited to certain types of tasks
Task Selection	Sufficiently reduce the number of tasks	<ul style="list-style-type: none">• Sacrifice some quality• Introduce some delay
Answer Deduction and Sampling	Avoid crowds doing a lot of unnecessary work	<ul style="list-style-type: none">• Introduce human error• Sampling fail under certain situation

Latency Control

- If the requester has a time limit, controlling latency is important
- Several strategies to address this issue:
 - ❖ Adjusting price
 - ❖ Latency modelling
 - ❖ Reducing Recruitment Time

Comparison:

- ❖ Increasing task price will greatly increase the cost
- ❖ Dynamic budget allocation may cause confused or dissatisfied among workers
- ❖ Latency modelling provides more objective decisions but
- ❖ The retainer model can efficiently reduce latency, but may introduce low quality results and more costs



Incentive

- **Monetary incentives**
 - ❖ Financial Benefits
 - ❖ Straightforward & Easy Implementation
- **Non-monetary incentives**
 - ❖ Individual Development
 - ❖ Public Good
 - ❖ Reputation
 - ❖ Gamification



Comparison

Incentive types	Pros	Cons
Monetary	<ul style="list-style-type: none"> • Straightforward way to motivate individuals to participate. • Compensation for contributions • Can attract a consistent and steady involvement of participants. • Better accuracy and speed compared to voluntary work 	<ul style="list-style-type: none"> • Pay rate should be carefully considered to match time and effort required. • Overpaying or underpaying can lead to issues. • Participants may attempt the project multiple times, leading to poor data quality. • Not feasible for project starters with no or low budget
Non-Monetary	<ul style="list-style-type: none"> • Cost-effective • Increase motivation. • Provide better data quality. • Can offer opportunities for personal skill development. • Can contribute to public good. • Enhance an individual's reputation 	<ul style="list-style-type: none"> • Fewer people may be willing to participate. • Not applicable for all workers • More time and effort required for task design. • Different people value different types of incentive

Discussion & Future Direction

- Uncertainty
- Bias
- Ethical issue

Reference

Chittilappilly, A. I., Chen, L., & Amer-Yahia, S. (2016). A Survey of General-Purpose Crowdsourcing Techniques. *IEEE Transactions on Knowledge and Data Engineering*, 28(9), 2246–2266.

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