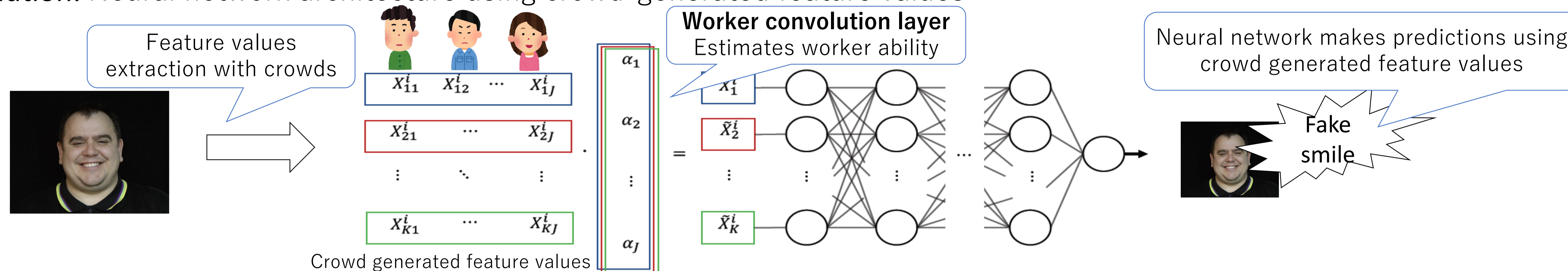




Summary

Goal: Classification problem using feature values given by crowdsourcing workers with different capabilities

Solution: Neural network architecture using crowd-generated feature values

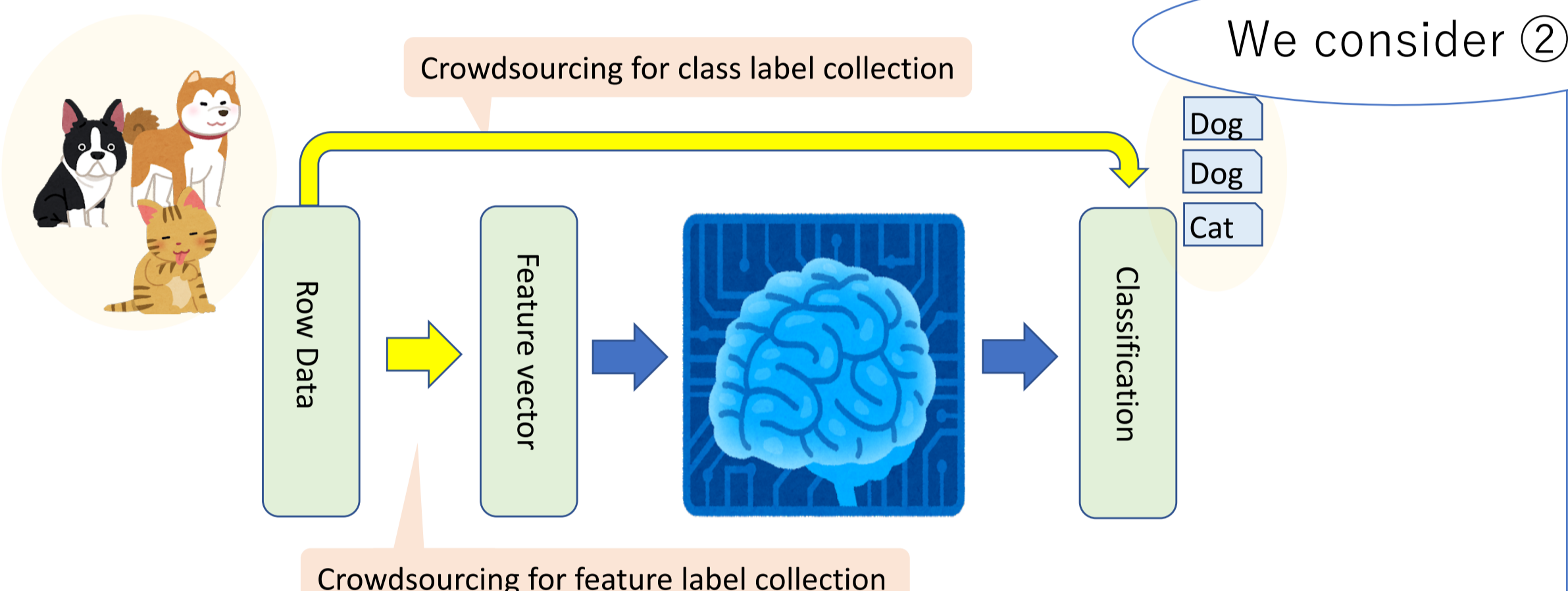


Background | Human-in-the-loop machine learning

Crowdsourcing is a system for outsourcing work to an unspecified number of workers via the Internet

Crowdsourcing is actively used in machine learning, especially for

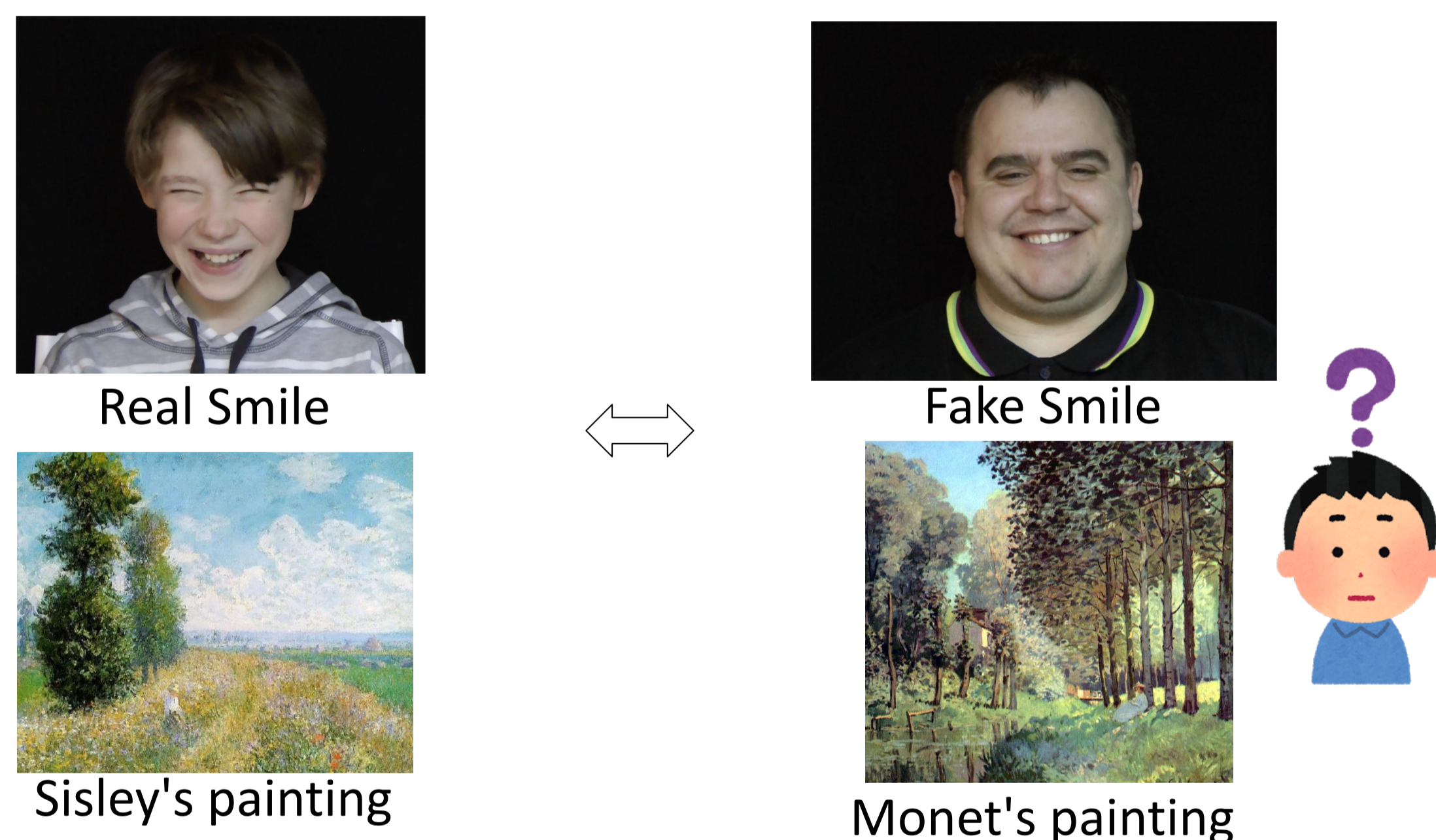
- class label collection for supervised learning, and
- feature label extraction for data representation



Challenge: Quality control of feature labels

- Quality of the provided feature labels is uneven because of different capability and diligence of crowd workers (Sometimes there are spam or malicious workers)
- We need to integrate feature labels from different workers to improve label quality

Motivation | Class labels are hard to give by non-experts



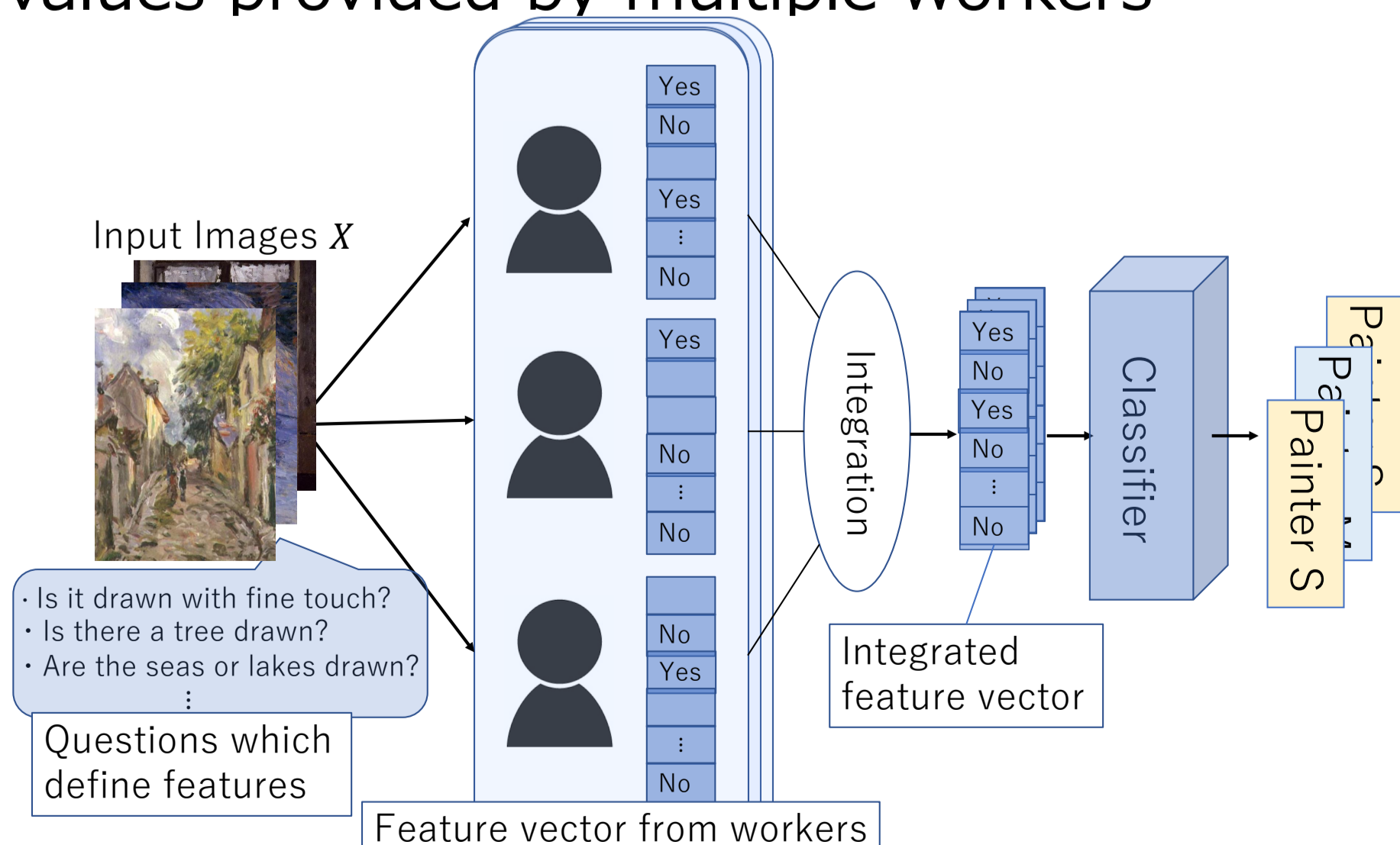
Hard for non-experts to give correct class labels.

Easier for non-experts to give feature labels

Ex) "Are the trees in this painting clearly drawn to the branches?"

Feature extraction using human beings is effective

Problem setting | Binary classification problem based on features values provided by multiple workers



- Perform binary classification from the feature values given by multiple workers.
- Feature labels are collected in the form of binary questions ("Yes" or "No").
- 3 workers are assigned to each feature

Proposed method | Neural network simultaneously estimate workers' ability and the classifier

Workers with higher feature extraction ability contribute more to predictions

- Estimate the worker's ability based on the prediction result
- Integrate opinions based on the estimated ability

We propose **worker convolution layer**

- Express worker's ability as weights of one-dimensional filter α .
- α_j corresponds to the ability of the j -th worker.
- Convolute with filters for each feature and generate integrated labels.

Experiments

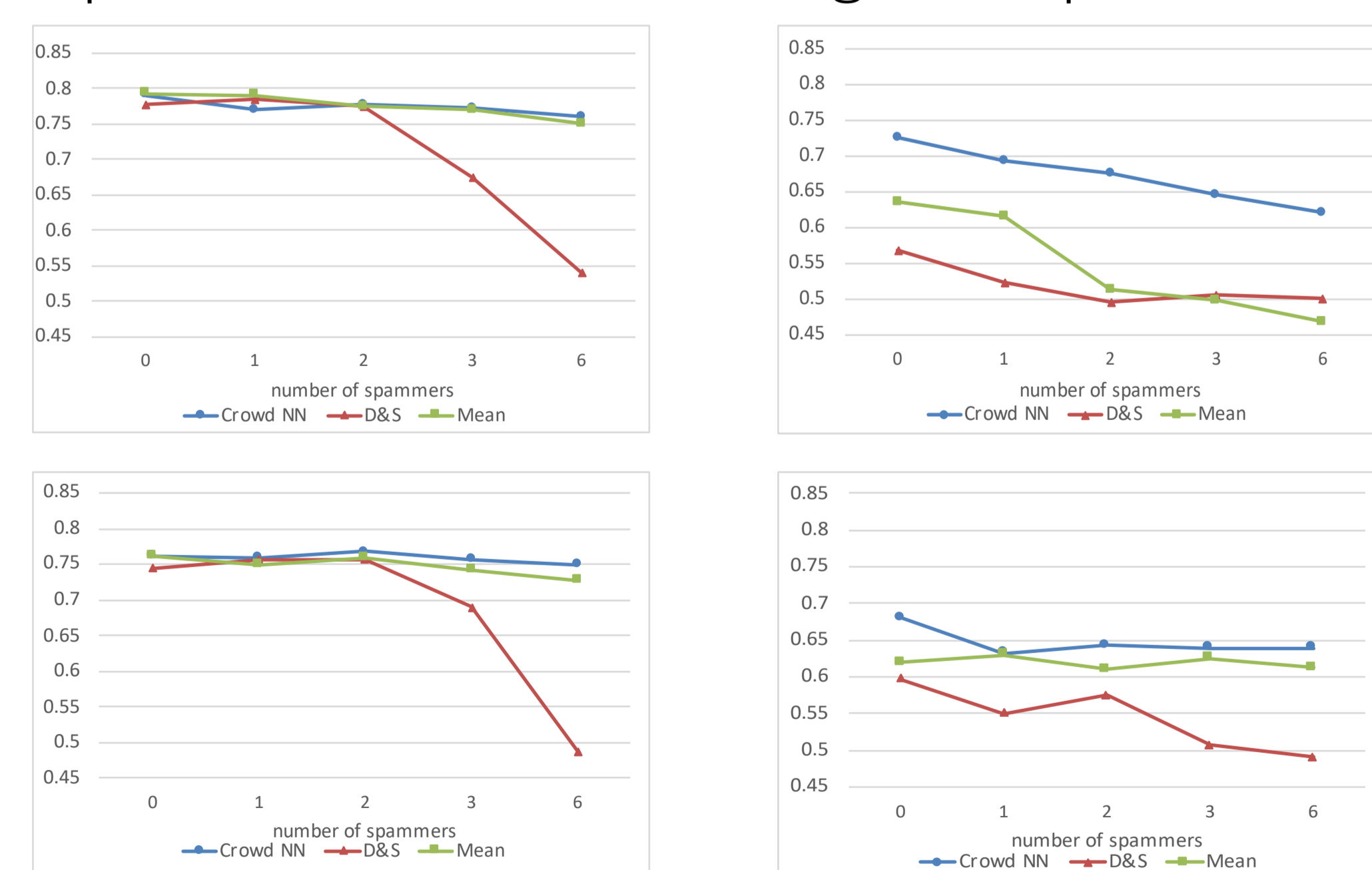
Three experiments with four datasets

① Model performance with original datasets

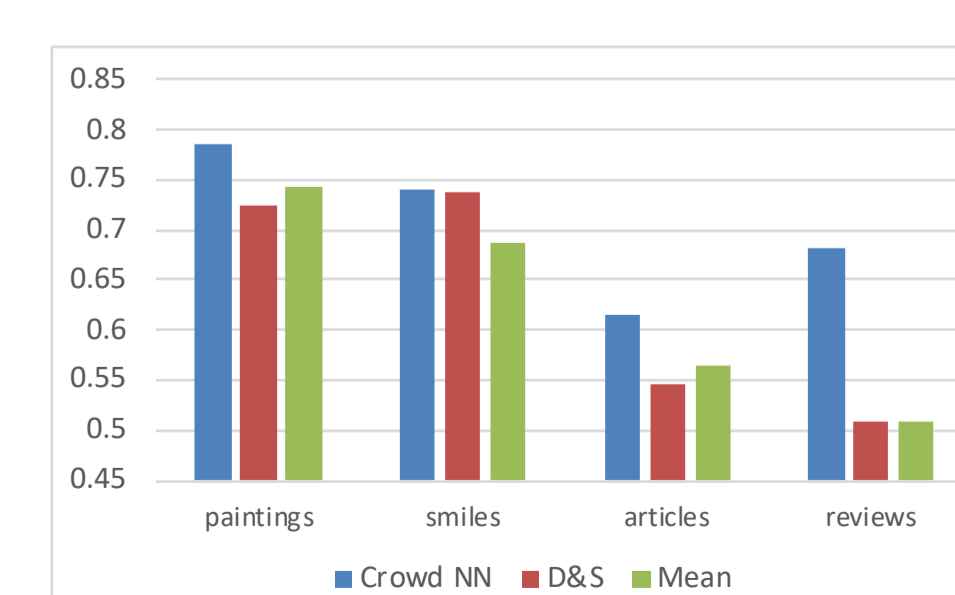
Dataset	Proposed method	Mean	Existing method
Paintings	0.790	<u>0.793</u>	0.778
Fake smiles	<u>0.763</u>	<u>0.763</u>	0.745
Fake reviews	<u>0.680</u>	0.620	0.598
Top news	<u>0.725</u>	0.635	0.568

② Experiments with simulated spam workers

Proposed method is robust against spam workers



③ Robustness against malicious workers



- Experiments with simulated malicious (giving reversed feature labels) workers.
- Proposed method exploits malicious workers to improve predictions.