



Building The Next Generation of Smart Classroom

Challenges and Opportunities in AI + Education

—— 好未来AI Lab负责人 刘子韬



好未来是一个以[智慧教育](#)和[开放平台](#)为主体，
以[素质教育](#)和[课外辅导](#)为载体，
在全球范围内服务[公办教育](#)，助力[民办教育](#)，
探索未来教育新模式的[科技教育公司](#)。





TAL AI LAB

好未来人工智能实验室

好未来用AI和科技赋能教育

更优质的
教学内容

更有效的
教学体验

更科学的
评测体系

更公平的
教育资源

Learning Environments

Classroom Environments

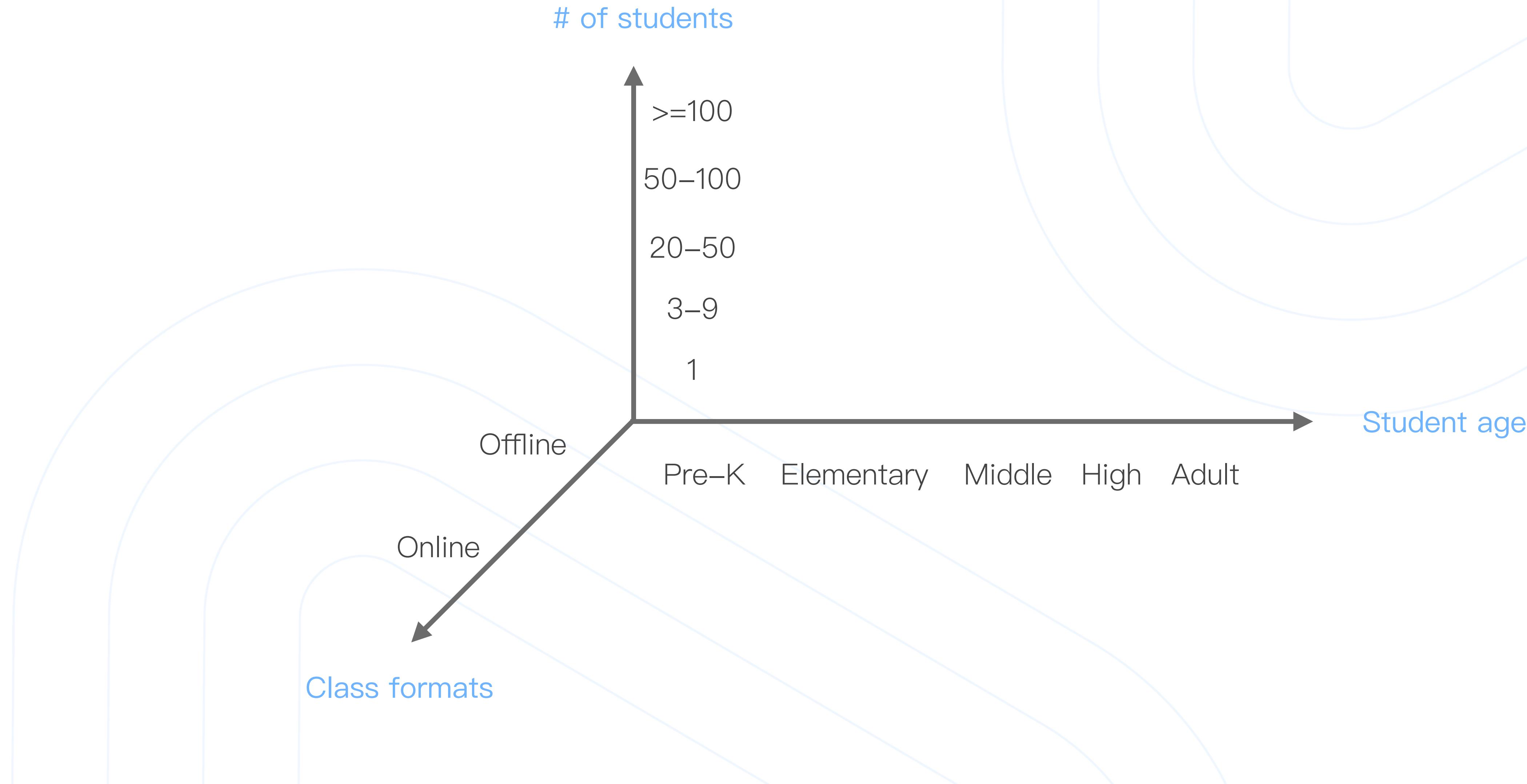
天空：
更有效？



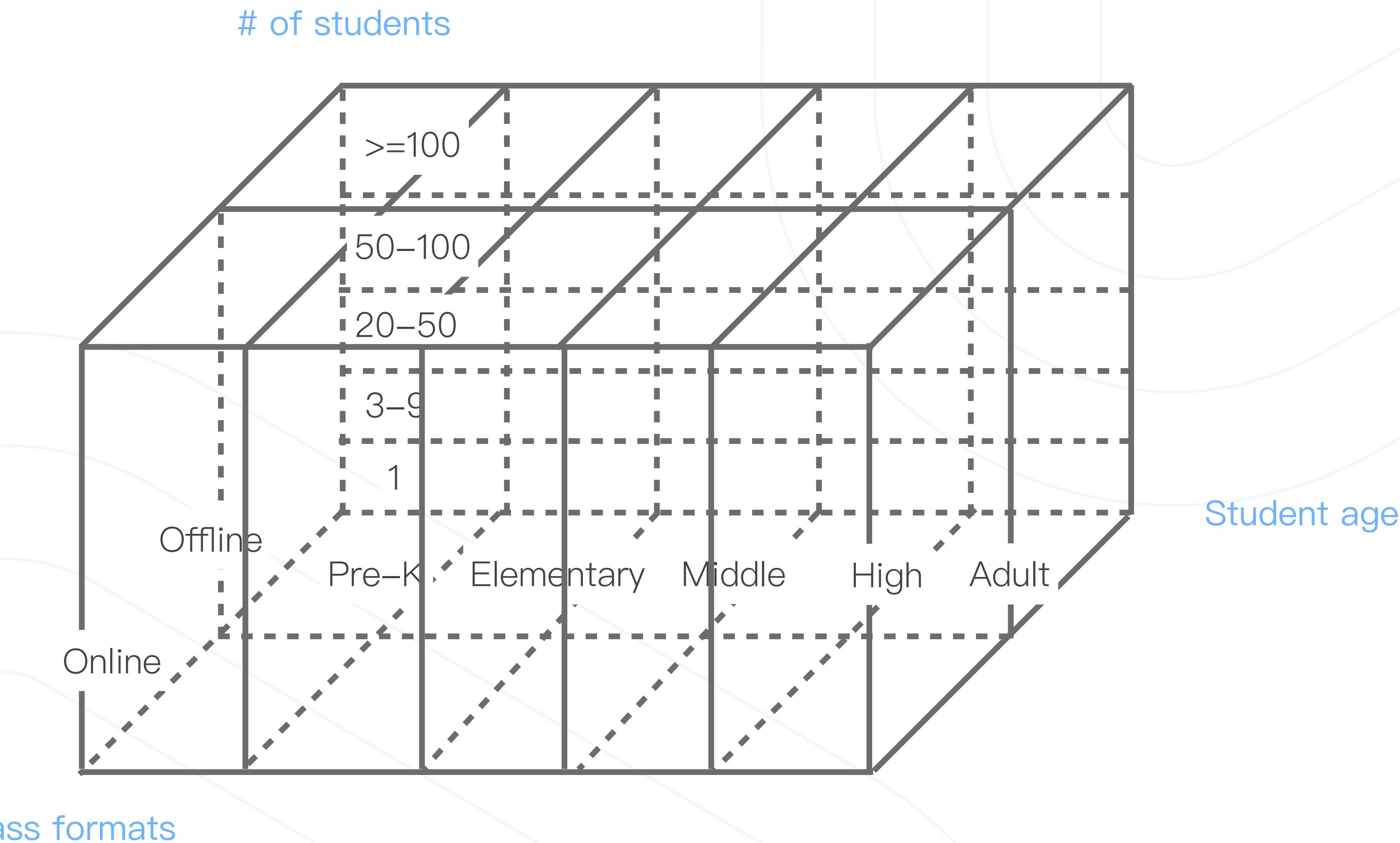
v.s.



Learning Environments



Learning Environments



Rich Educational Data

of students

Student age

Class formats

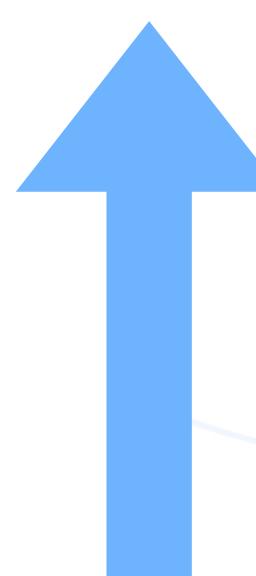
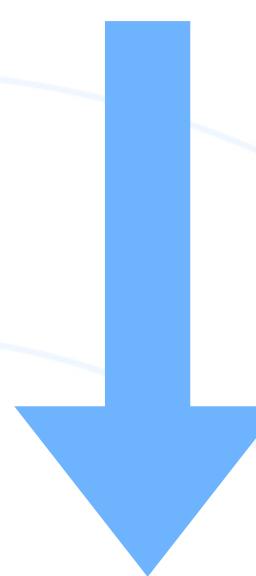


Rich Educational Data

Pre-class

In-class

Post-class

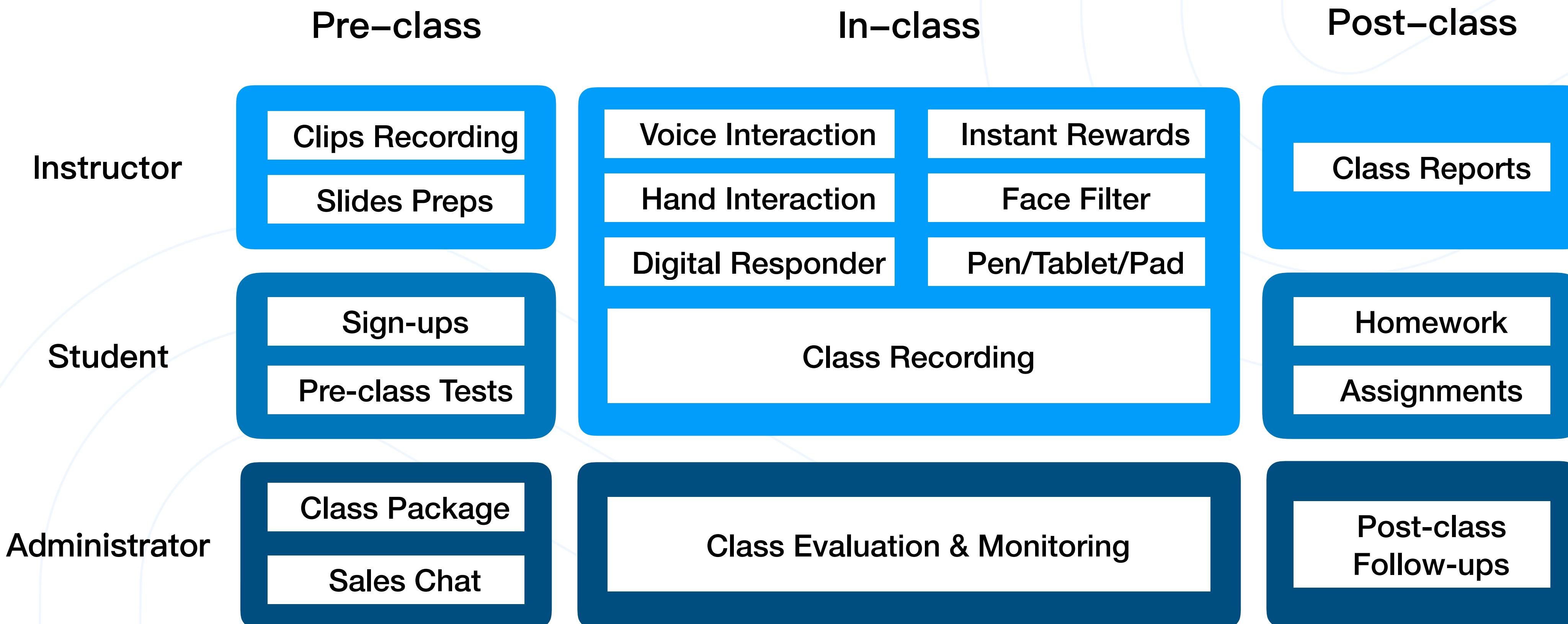


Students

Instructors

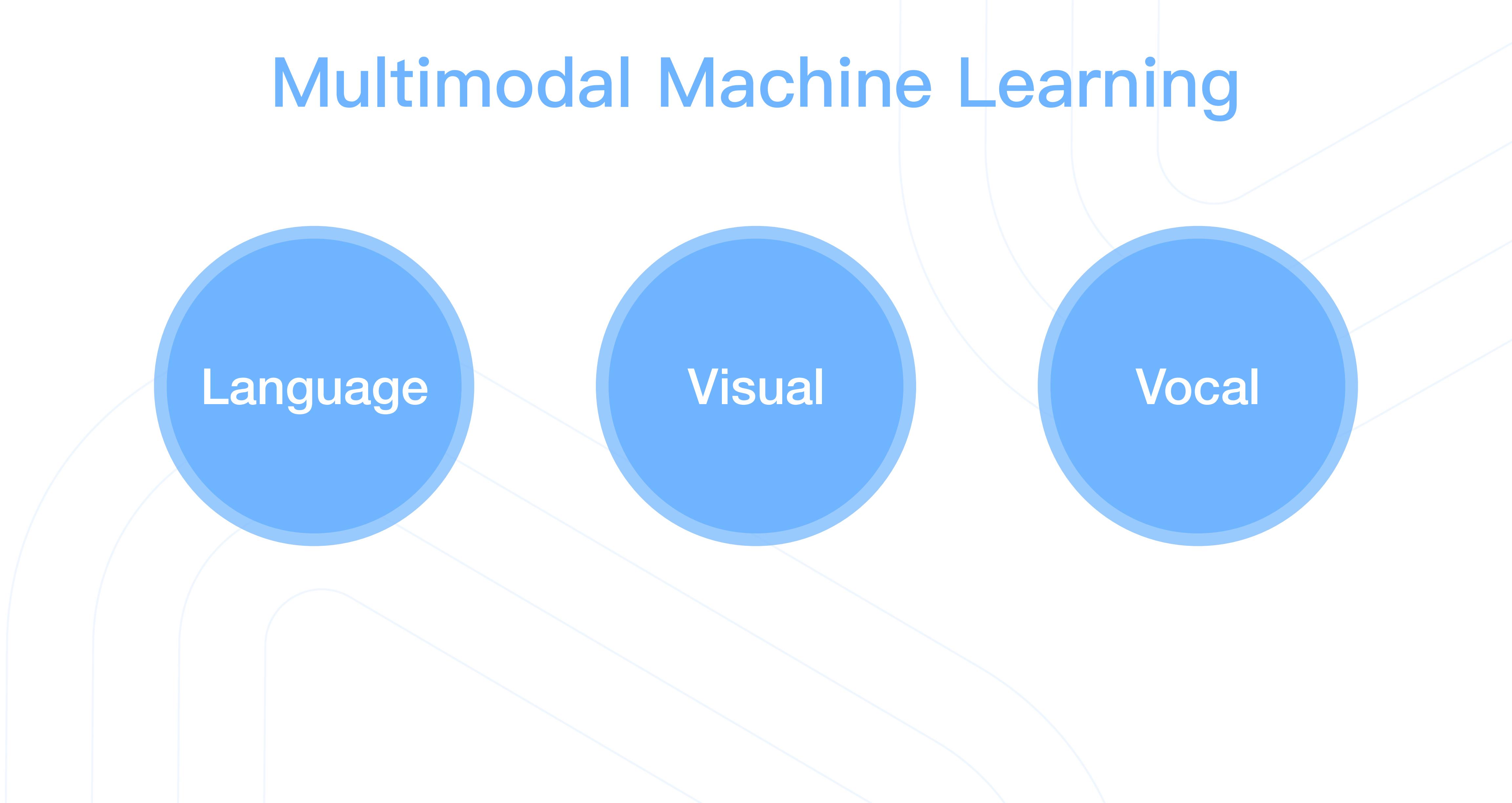
Administrators

Rich Educational Data



Educational data is multimodal

Multimodal Machine Learning



Language

Visual

Vocal

Multimodal Machine Learning

Challenges

Representation

Learn how to represent and summarize multimodal data

Translation

Map data from one modality to another

Alignment

Identify relations between (sub)elements from different modalities

Fusion

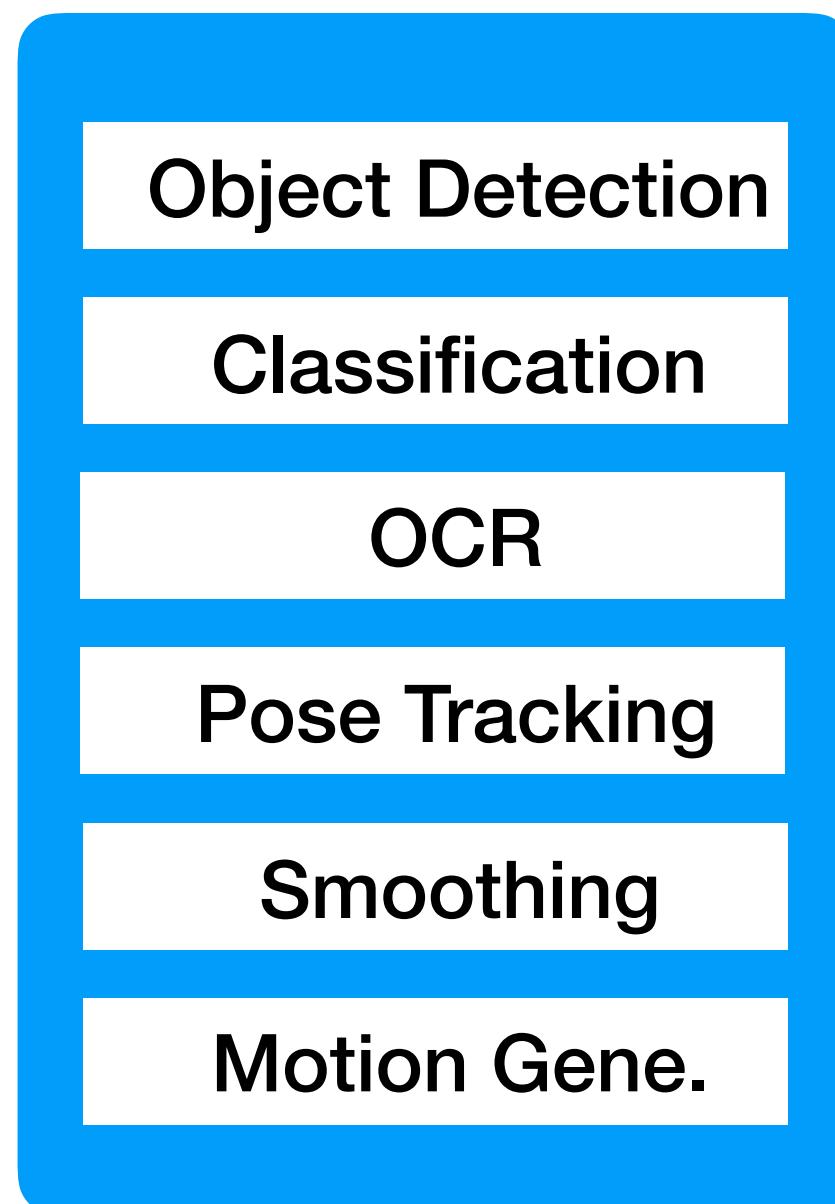
Join information from modalities to perform a prediction

Co-learning

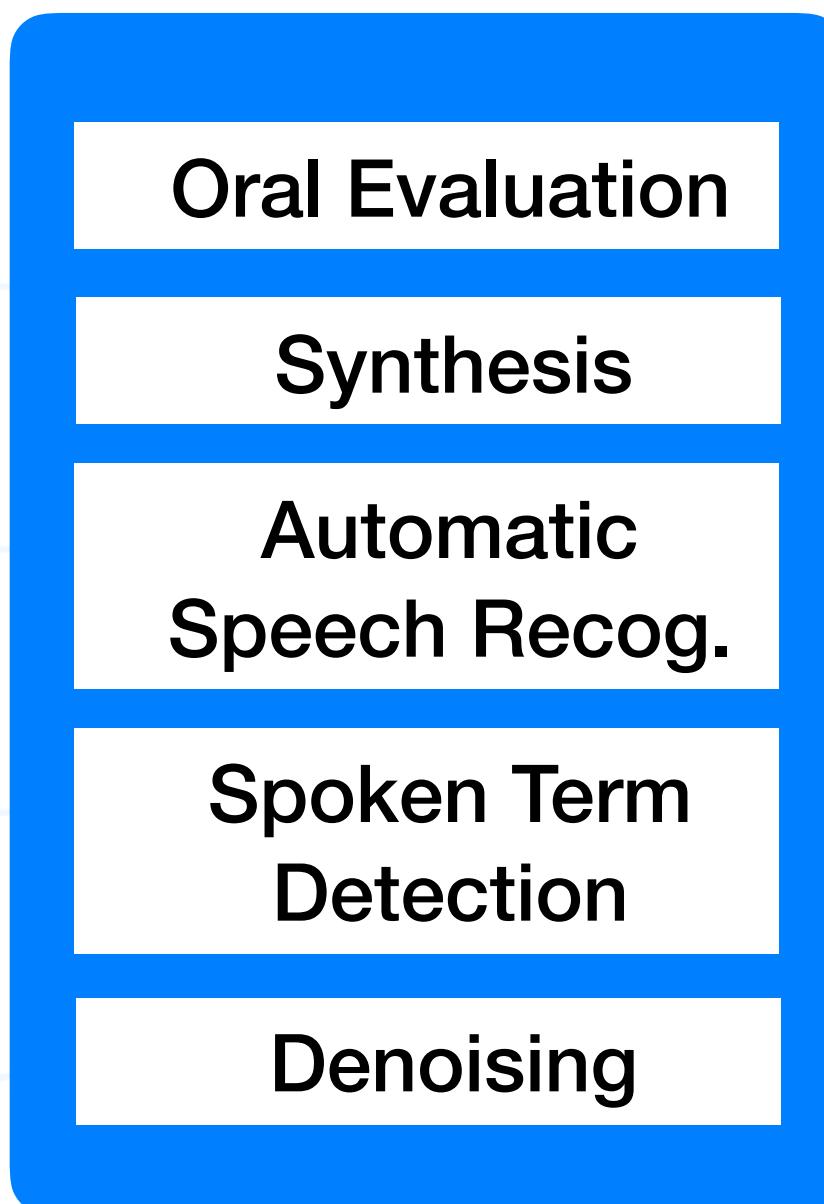
Transfer knowledge between modalities, their representation, and their predictive models

AI Stacks

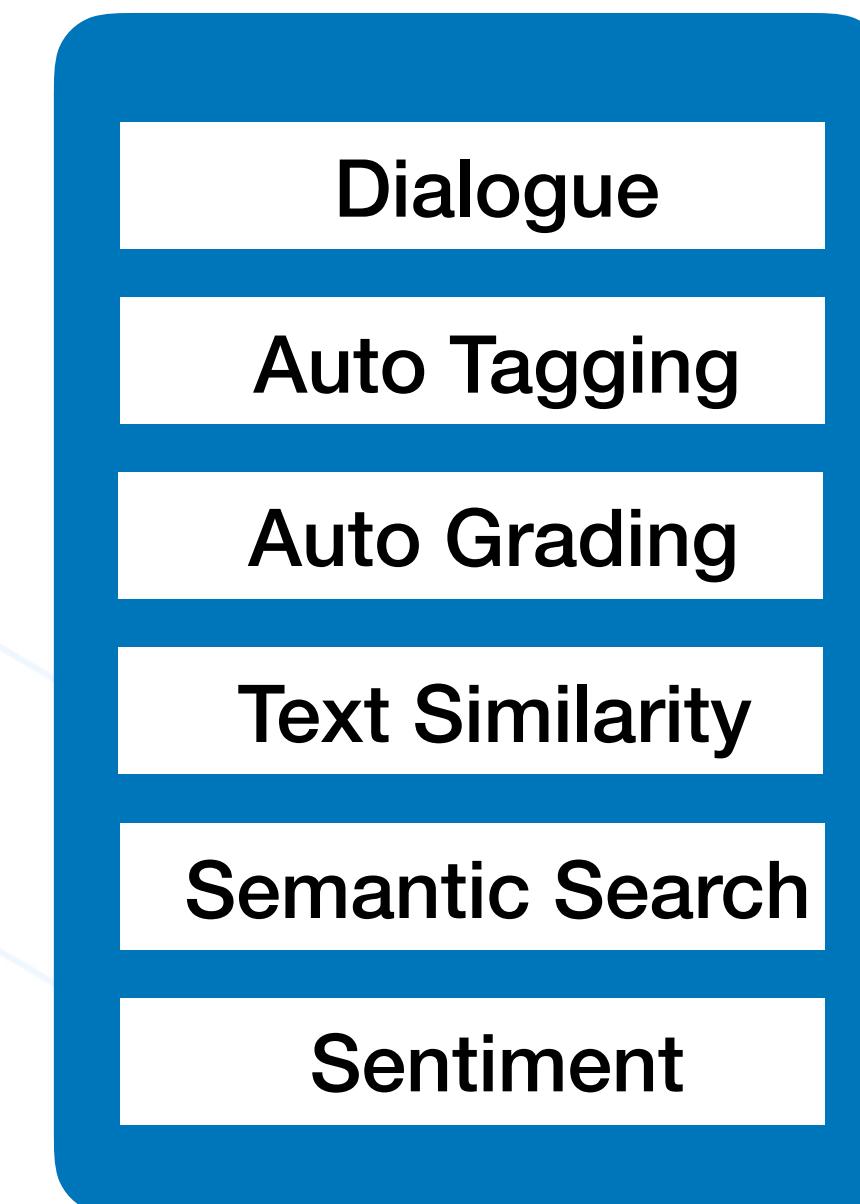
Computer Vision



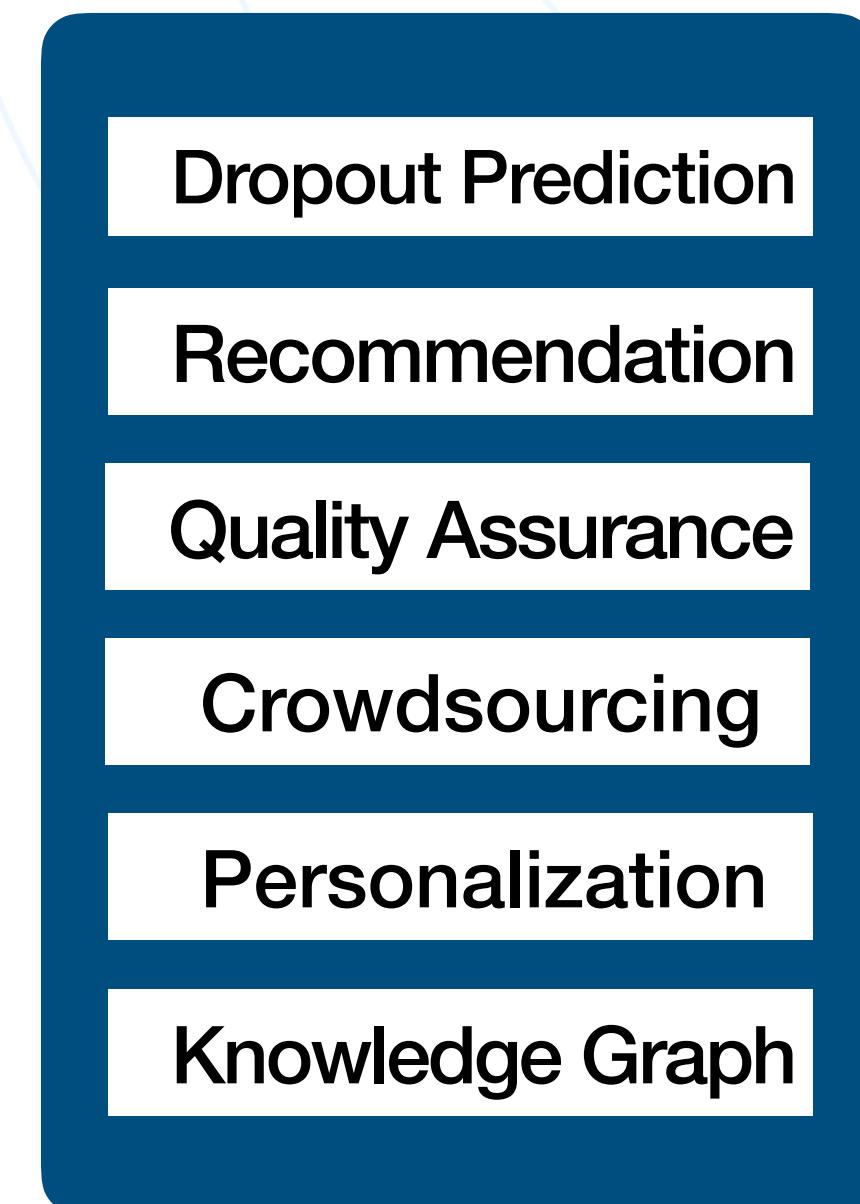
Speech



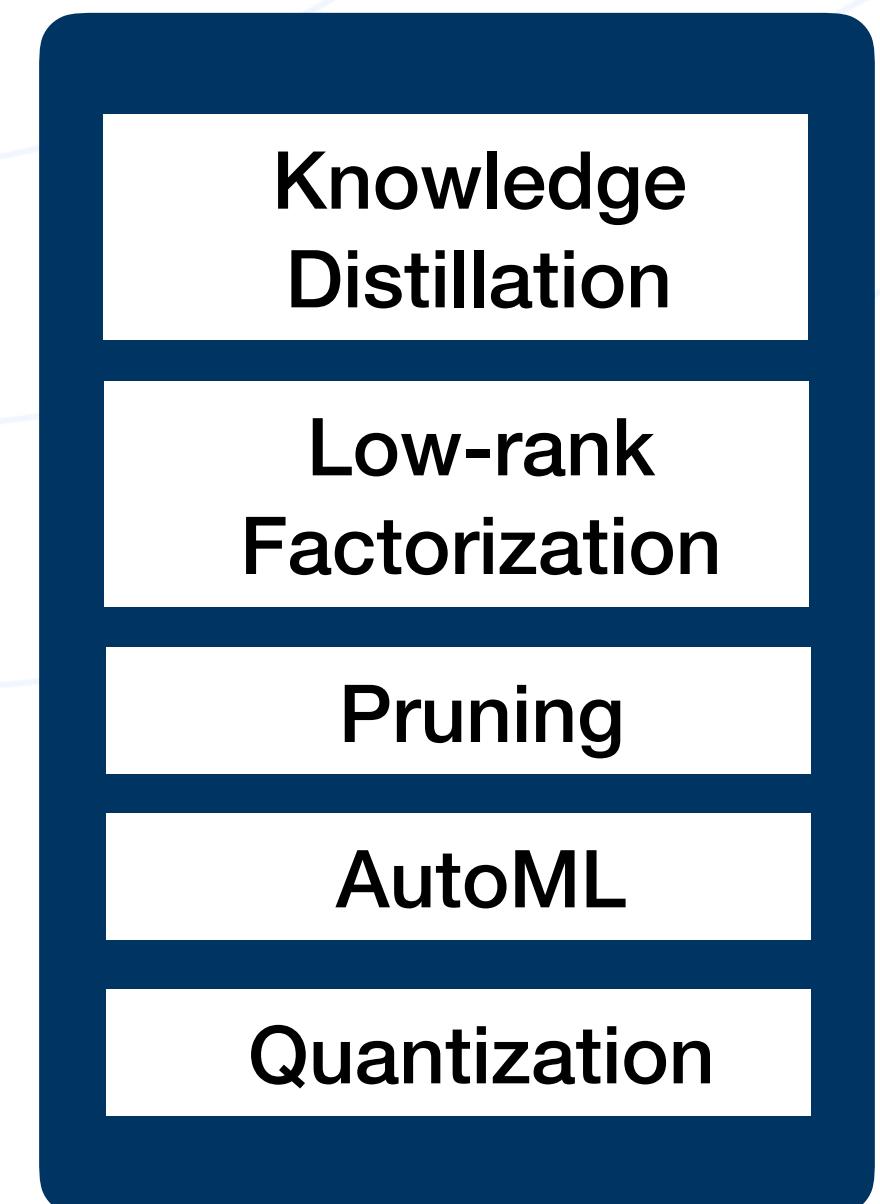
Natural Language Understanding



Data Mining



Compression & Acceleration



Challenges

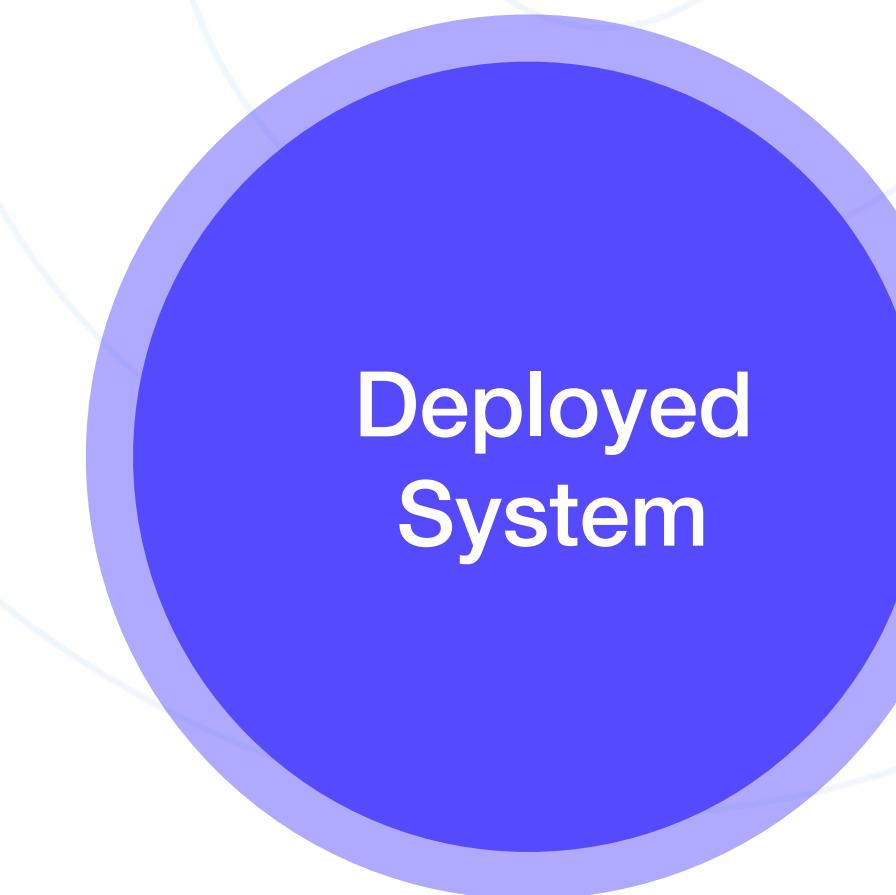
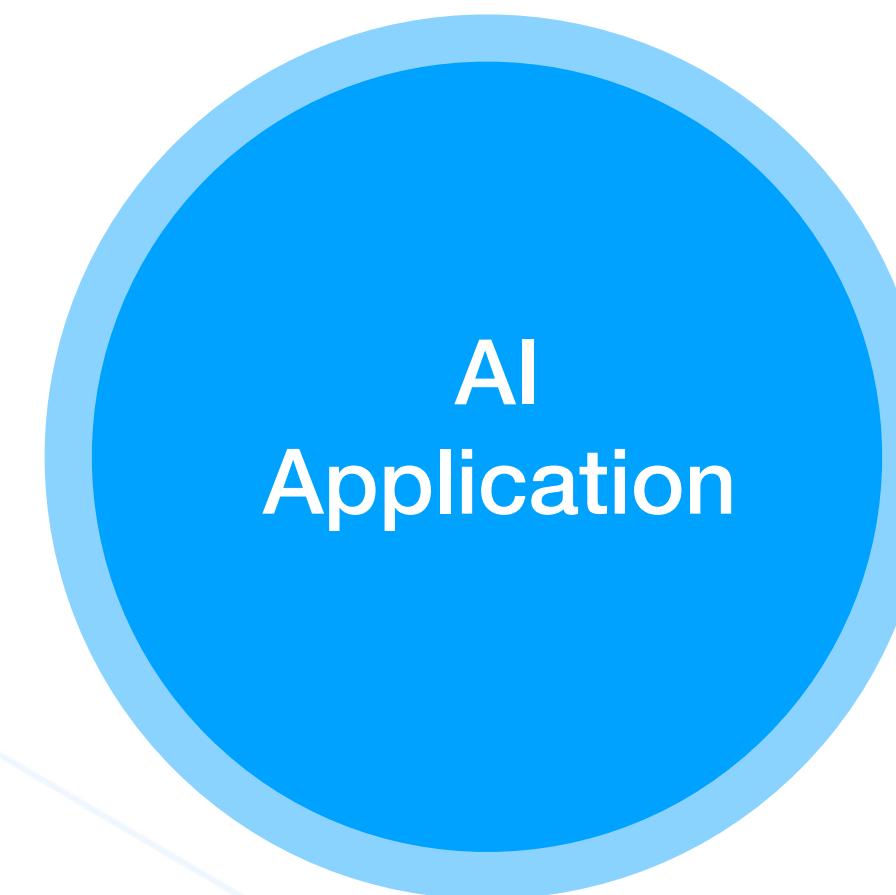
Small Data

Heterogeneity

Data Quality Issues

Evaluation

Our Work



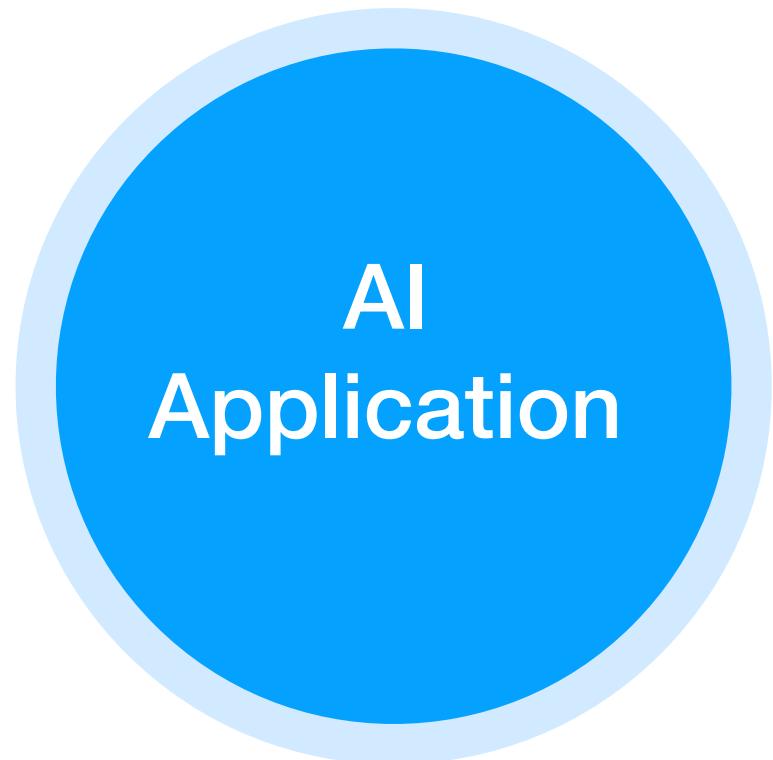
Content



Methodology &
Framework

- **Recommender Systems with Heterogeneous Side Information.** T. Liu, Z. Wang, J. Tang, S. Yang, Z. Liu. WWW'19, International Conference on the World Wide Web, 2019.
- **Learning Effective Embeddings From Crowdsourced Labels: An Educational Case Study.** G. Xu, W. Ding, J. Tang, S. Yang, Z. Liu. ICDE'19: The 35th IEEE International Conference on Data Engineering, 2019.
- **Representation Learning from Limited Educational Data via Adaptive Hard Example Mining.** G. Xu, W. Fu, W. Ding, G. Huang, Z. Liu. Submitted.

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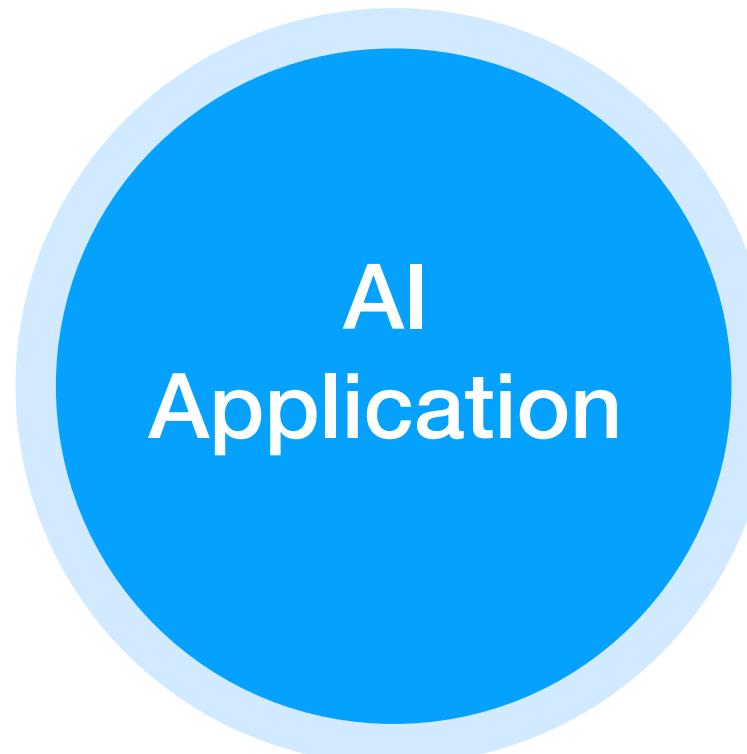


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- **Deep Knowledge Tracing with Side Information.** Z. Wang, X. Feng, J. Tang, G. Huang, Z. Liu. AIED'19: The 20th International Conference on Artificial Intelligence in Education, 2019.
 - **Automatic Short Answer Grading via Multiway Attention Networks.** T. Liu, W. Ding, Z. Wang, J. Tang, G. Huang, Z. Liu. AIED'19: The 20th International Conference on Artificial Intelligence in Education, 2019.

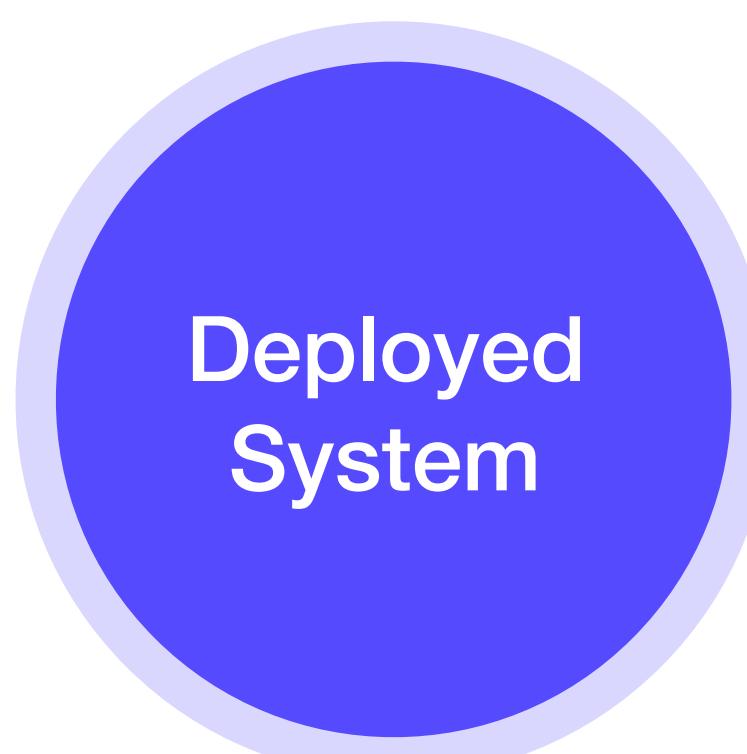
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- **A Multimodal Alerting System for Online Class Quality Assurance.** J. Chen, H. Li, W. Wang, W. Ding, G. Huang, Z. Liu. AIED'19: The 20th International Conference on Artificial Intelligence in Education, 2019.
- **Dolphin: A Verbal Fluency Evaluation System for Elementary Education.** G. Xu, T. Liu, W. Fu, Y. Song, C. Guo, W. Ding, G. Huang, Z. Liu. Submitted.

Learning From Crowdsourced Labels

Learning Effective Embeddings From Crowdsourced Labels: An Educational Case Study. G. Xu, W. Ding, J. Tang, S. Yang, Z. Liu. ICDE'19: The 35th IEEE International Conference on Data Engineering, 2019.

Introduction

Learning representation is important!

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Existing approaches rely on:

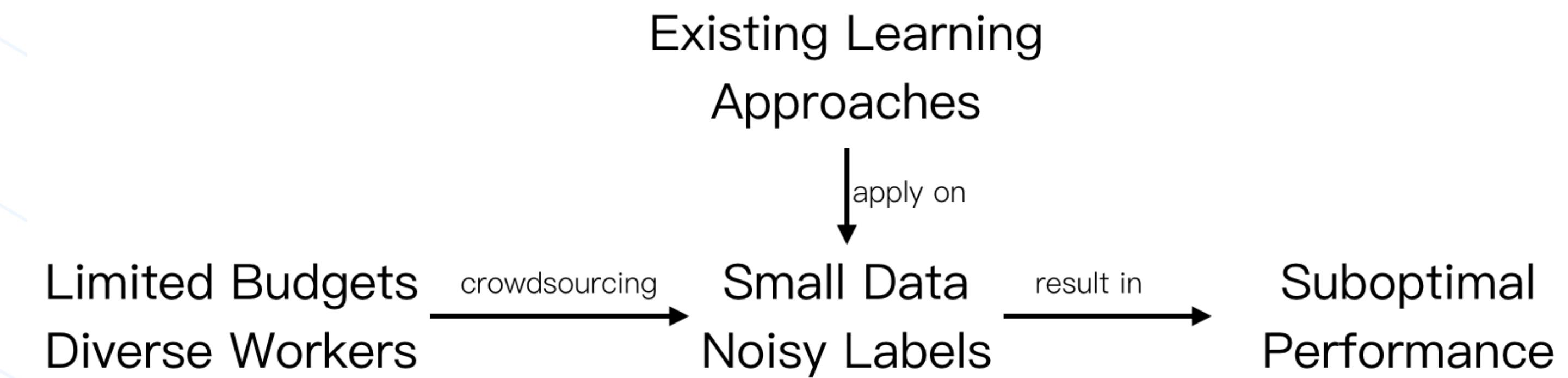
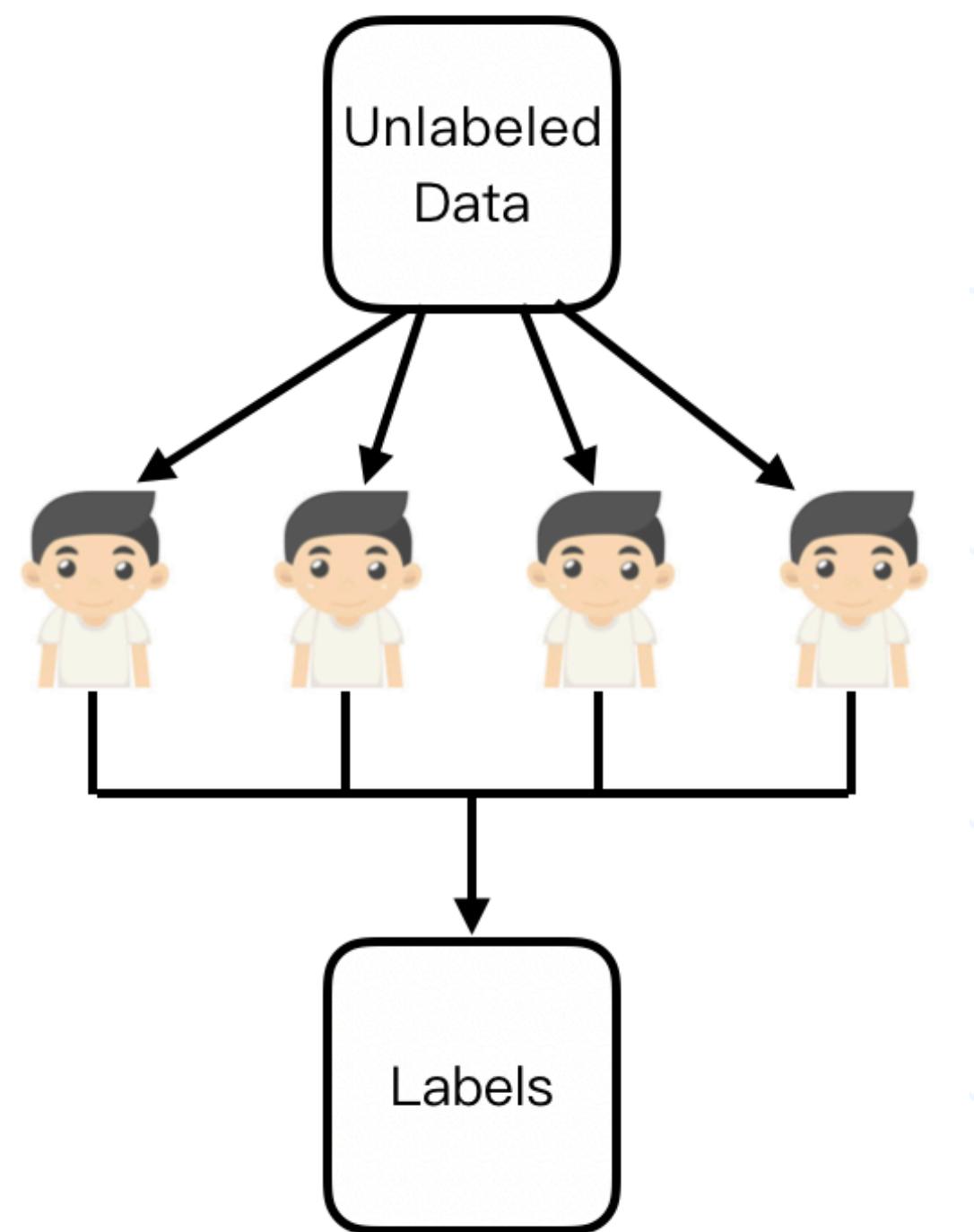
- **Massive** data
- **Accurate** labels

Real-world reality:

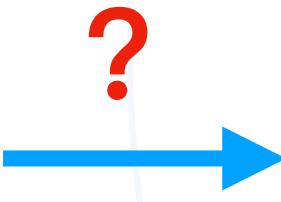
- **Limited** data
- **Noisy** labels

Introduction

How to obtain labeled data? **Crowdsourcing**



Solution

Limited data + Noisy crowdsourced labels  Robust representation

RLL: An unified framework to learn robust representation from small data and inconsistent labels

Grouping Based Architecture

Original data

$$\mathcal{D}^+ = \langle \mathbf{x}_1^+, \mathbf{x}_2^+, \dots, \mathbf{x}_m^+ \rangle$$

$$\mathcal{D}^- = \langle \mathbf{x}_1^-, \mathbf{x}_2^-, \dots, \mathbf{x}_n^- \rangle$$

\mathbf{x}_1^+ \mathbf{x}_2^+ ... \mathbf{x}_m^+

\mathbf{x}_1^- \mathbf{x}_2^- ... \mathbf{x}_n^-

Positive



Negative



Data size:

$$O(m + n)$$

Grouping data

\mathbf{x}_1^+ \mathbf{x}_2^+ \mathbf{x}_1^- \mathbf{x}_2^- ... \mathbf{x}_k^- g_1

\mathbf{x}_1^+ \mathbf{x}_3^+ \mathbf{x}_1^- \mathbf{x}_2^- ... \mathbf{x}_k^- g_2

\mathbf{x}_1^+ \mathbf{x}_4^+ \mathbf{x}_1^- \mathbf{x}_2^- ... \mathbf{x}_k^- g_3

\vdots \vdots \vdots \vdots \vdots g_{N-1}

\mathbf{x}_2^+ \mathbf{x}_1^+ \mathbf{x}_i^- \mathbf{x}_j^- ... \mathbf{x}_l^- g_N

\mathbf{x}_i^+ \mathbf{x}_j^+ \mathbf{x}_p^- \mathbf{x}_q^- ... \mathbf{x}_r^- g_N

Pairing

$$\mathbf{g}_i = \langle \mathbf{x}_i^+, \mathbf{x}_j^+, \mathbf{x}_1^-, \mathbf{x}_2^-, \dots, \mathbf{x}_k^- \rangle$$

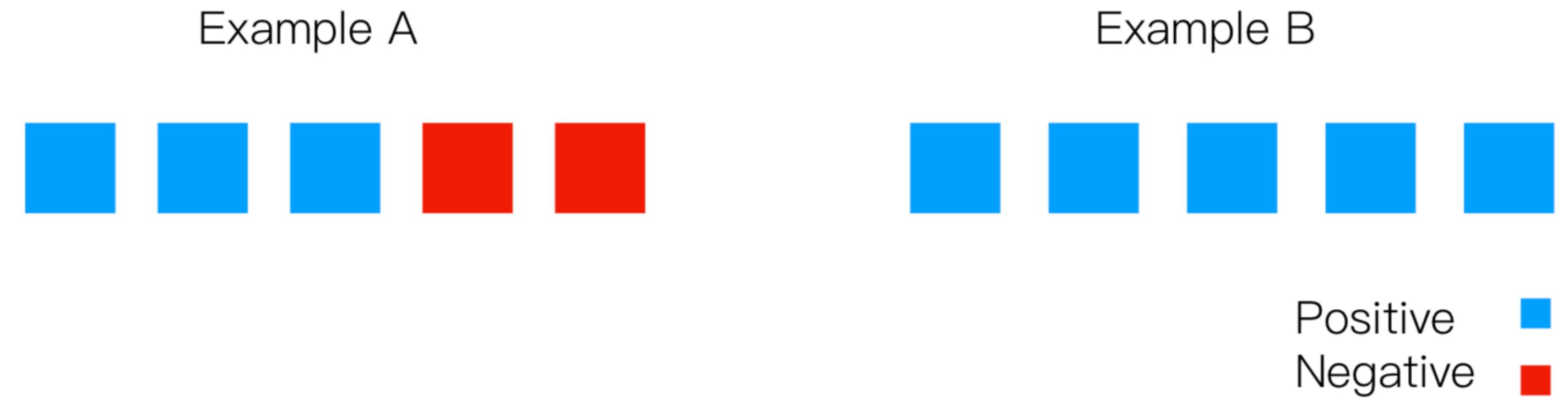
$$\mathcal{G} = \{\mathbf{g}_1, \mathbf{g}_2, \dots, \mathbf{g}_N\} \quad O(|m|^2 \cdot |n|^k)$$

Substantially
more data!

Confidence Score Estimator

5 workers label examples

A case where labels are inconsistent



Treat confidence score as a random variable:

- Using maximum likelihood estimation (MLE):

$$MLE(\delta_i) = \sum_{j=1}^d y_{i,j}/d$$

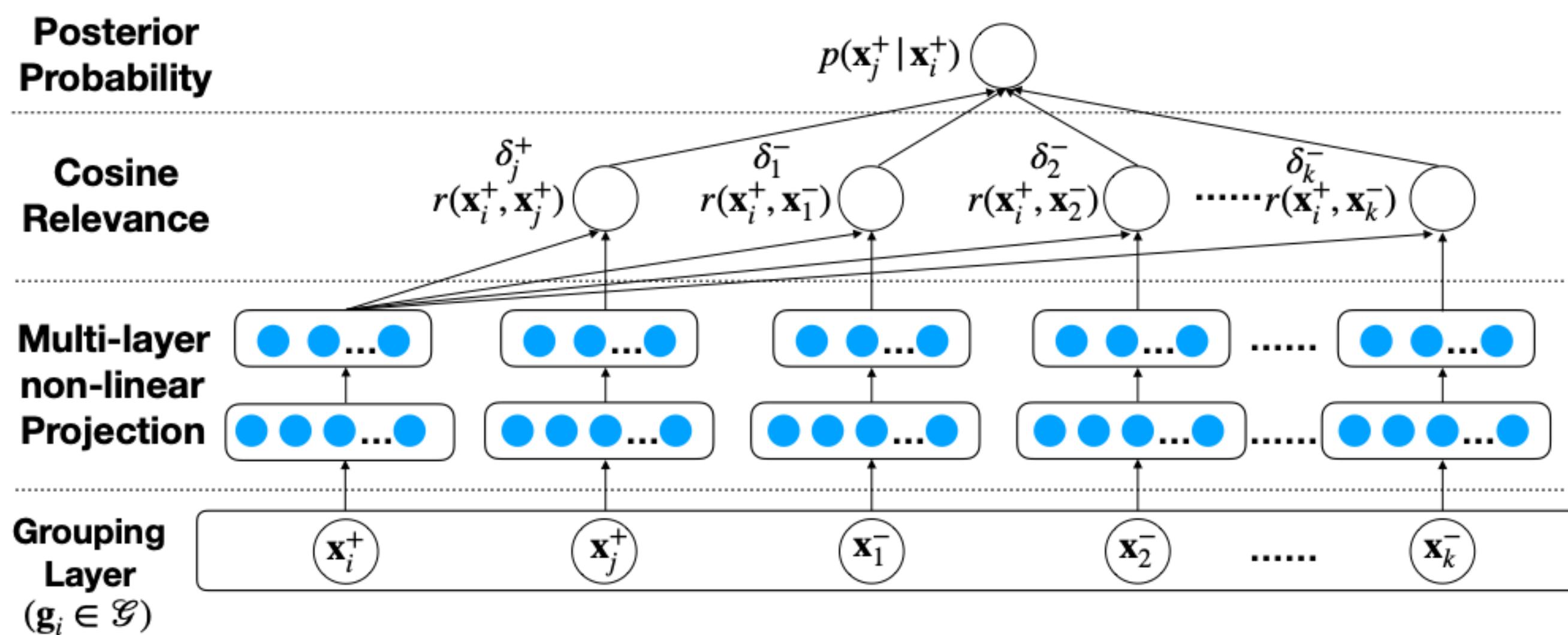
- Using Bayesian inference:

$$Bayesian(\delta_i) = \frac{\alpha + \sum_{j=1}^d y_{i,j}}{\alpha + \beta + d}$$

α and β are label class priors

RLL Framework

$$\text{RLL} = \text{Grouping based strategy} + \text{Confidence estimator} + \text{Model learning}$$



Integrate confidence into posterior probability:

$$\hat{p}(\mathbf{x}_j^+ | \mathbf{x}_i^+) = \frac{\exp(\eta \cdot \delta_j \cdot r(\mathbf{x}_i^+, \mathbf{x}_j^+))}{\sum_{\mathbf{x}_* \in \mathbf{g}_i, \mathbf{x}_* \neq \mathbf{x}_i^+} \exp(\eta \cdot \delta_* \cdot r(\mathbf{x}_i^+, \mathbf{x}_*))}$$

Loss function:

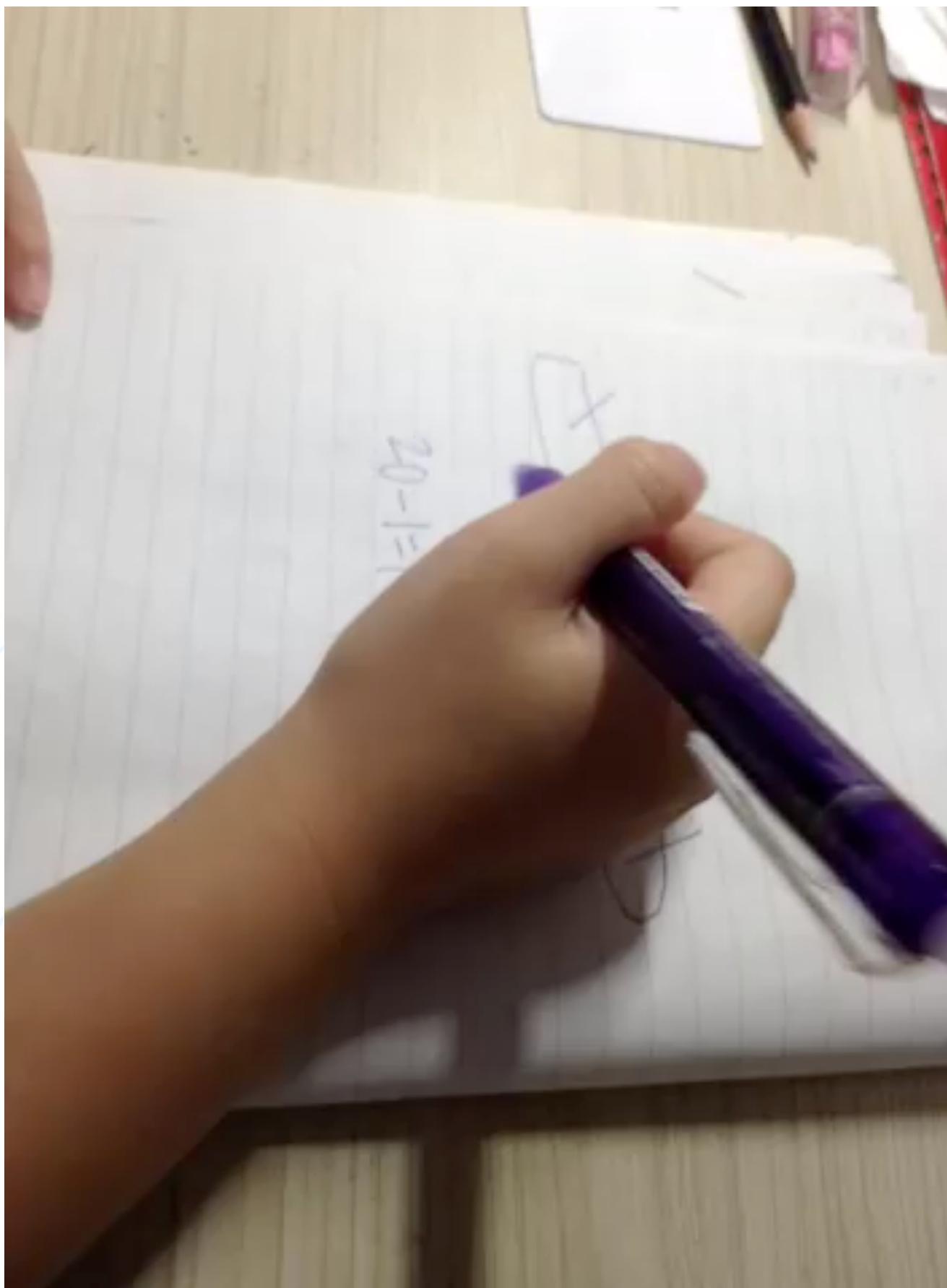
$$\mathcal{L}(\Omega) = - \sum_{i=1}^n \log p(\mathbf{x}_j^+ | \mathbf{x}_i^+)$$

The overview of RLL framework

Automatic Short Answer Grading via Multiway Attention Networks

Automatic Short Answer Grading via Multiway Attention Networks. T. Liu, W. Ding, Z. Wang, J. Tang, G. Huang, Z. Liu.
AIED'19: The 20th International Conference on Artificial Intelligence in Education, 2019.

Introduction



→
★ + 评价

第15讲作业 提交

王婧瑄6 家庭作业 - 第4... 上一题 下一题

已批改

题

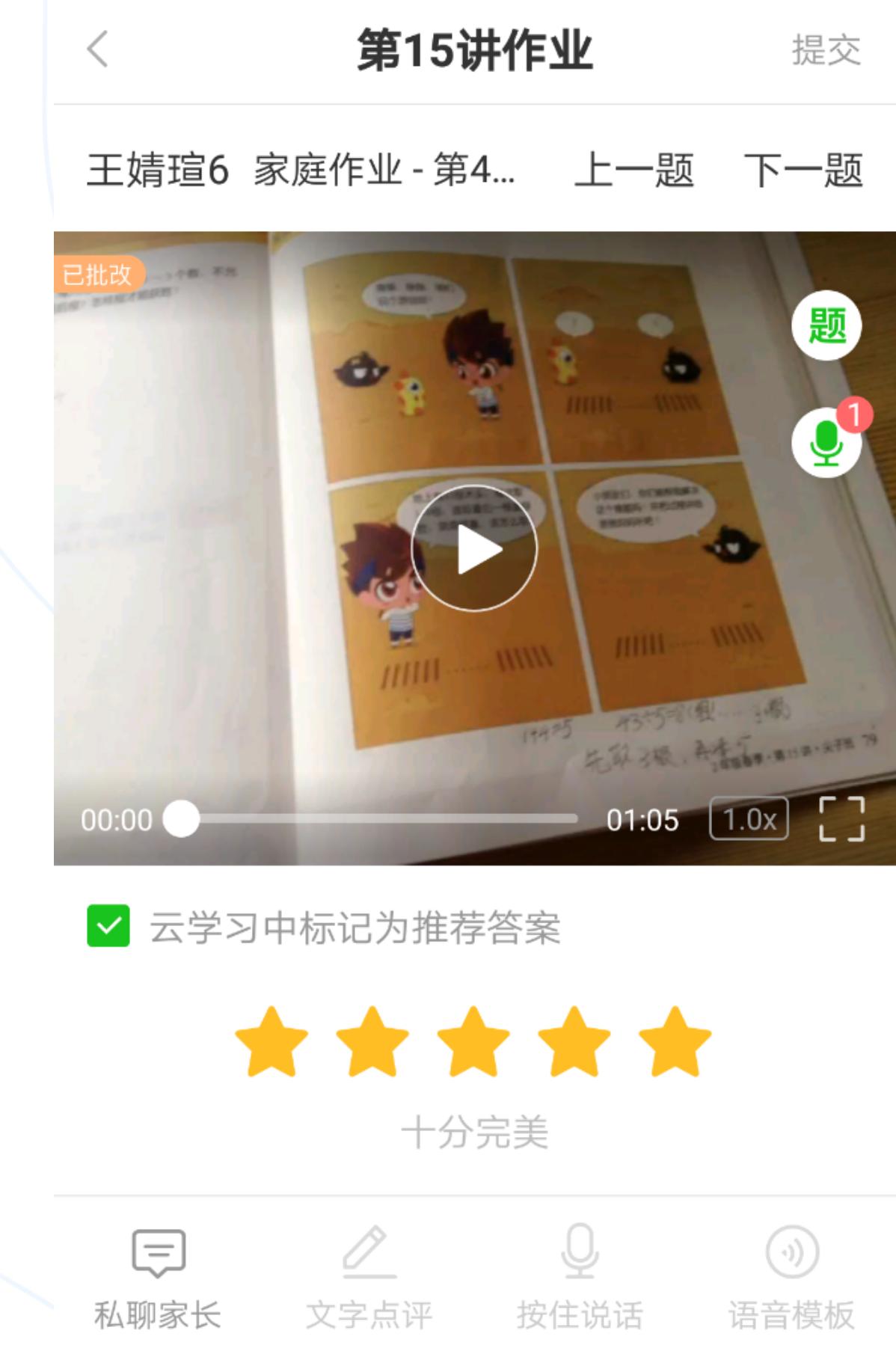
1

00:00 01:05 1.0x

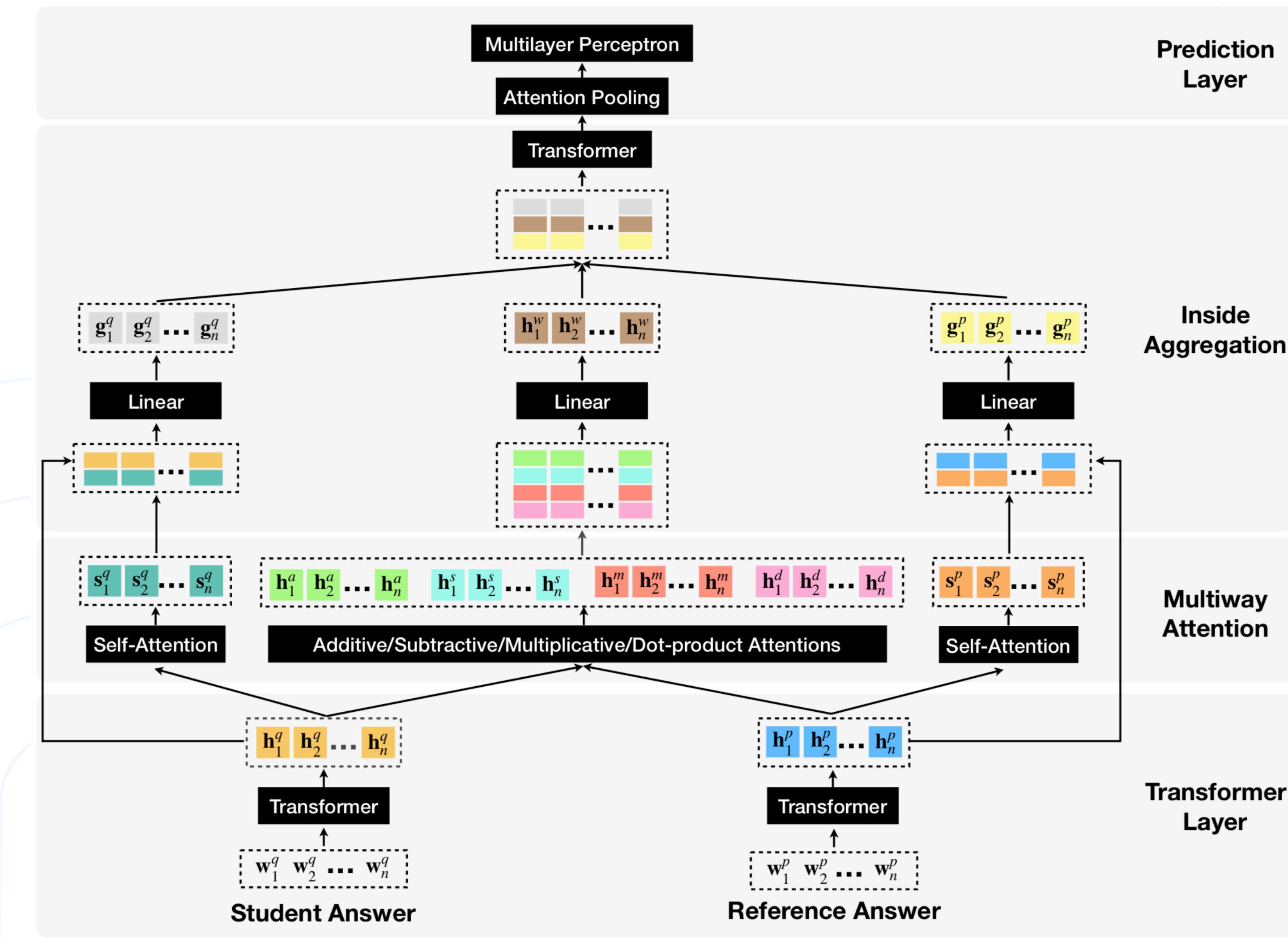
✓ 云学习中标记为推荐答案

★★★★★ 十分完美

私聊家长 文字点评 按住说话 语音模板



Our Solution



Dataset

We collect 120,000 pairs of student answers and questions. Each labeled with binary value indicating whether the student has the right answer.

Results

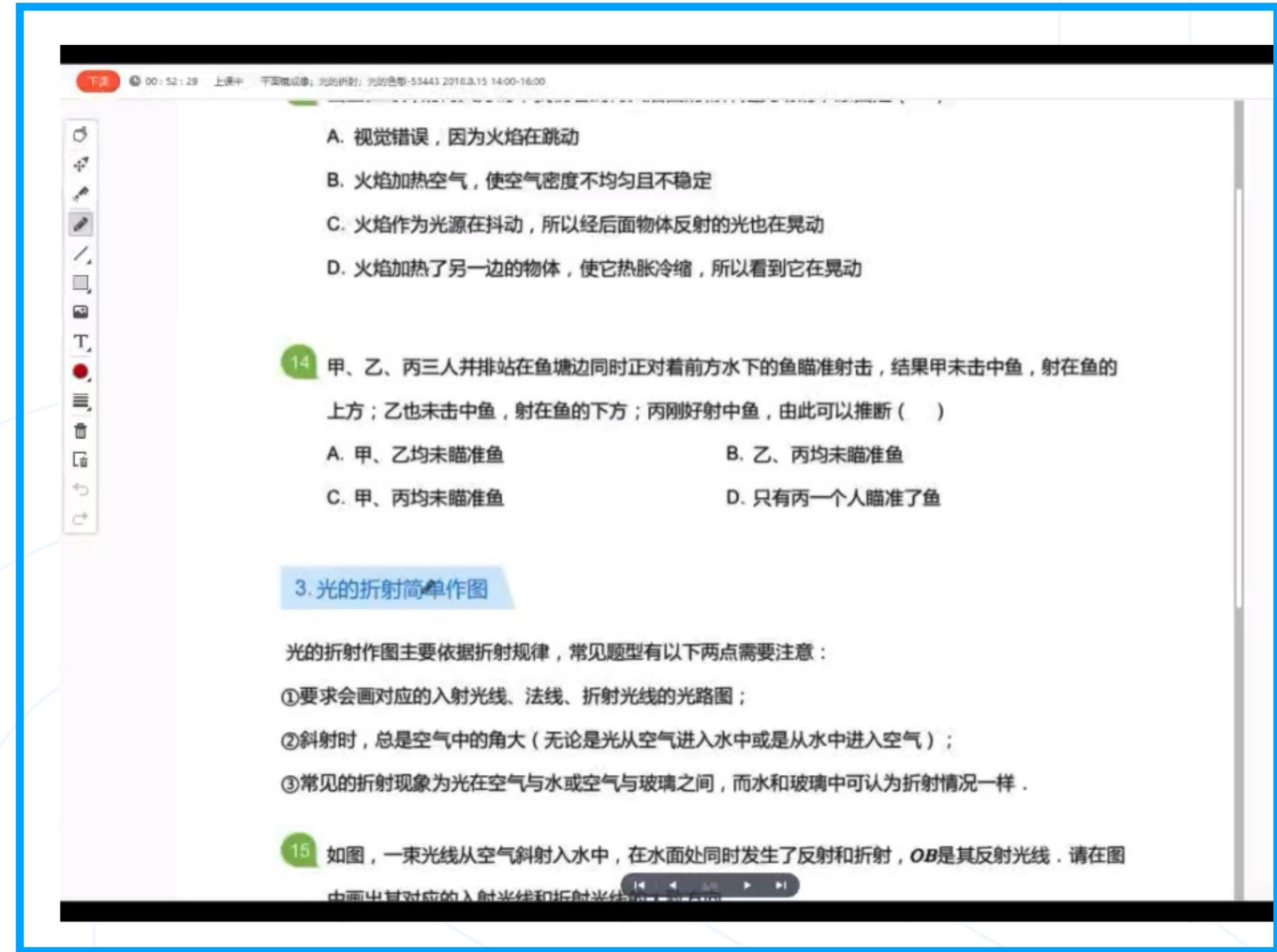
Table 1: ASAG performance comparison on a real-world K-12 dataset.

	LR	GBDT	TextCNN	Bi-Transformer	MaLSTM	MAN	Our
Accuracy	0.8297	0.8628	0.8772	0.8813	0.8825	0.8808	0.8899
AUC	0.8808	0.9287	0.9312	0.9335	0.9375	0.9365	0.9444

A Multimodal Alerting System for Online Class Quality Assurance

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Slides



00:52:29 上课中 平面视像：光的折射：光的色散-53443 2018.8.15 14:00-16:00

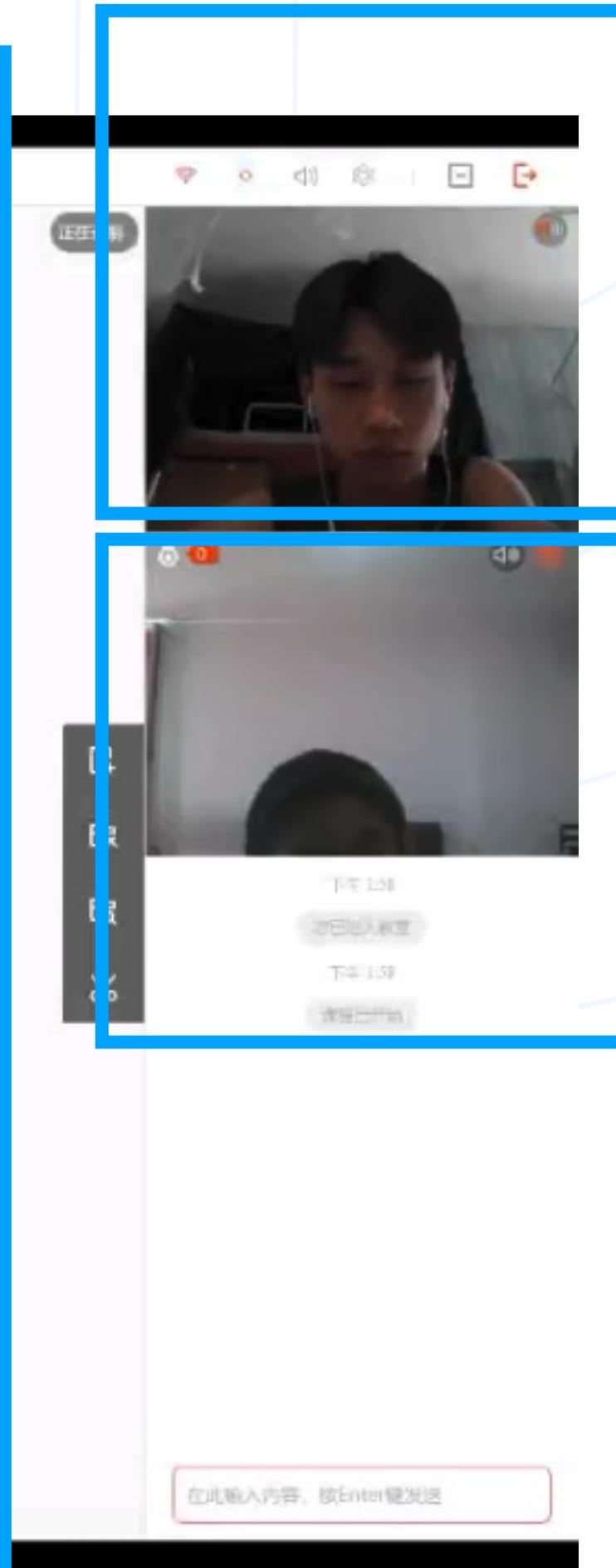
A. 视觉错误，因为火焰在跳动
 B. 火焰加热空气，使空气密度不均匀且不稳定
 C. 火焰作为光源在抖动，所以经后面物体反射的光也在晃动
 D. 火焰加热了另一边的物体，使它热胀冷缩，所以看到它在晃动

14 甲、乙、丙三人并排站在鱼塘边同时正对着前方水下的鱼瞄准射击，结果甲未击中鱼，射在鱼的上方；乙也未击中鱼，射在鱼的下方；丙刚好射中鱼，由此可以推断（ ）
 A. 甲、乙均未瞄准鱼 B. 乙、丙均未瞄准鱼
 C. 甲、丙均未瞄准鱼 D. 只有丙一个人瞄准了鱼

3. 光的折射简单作图

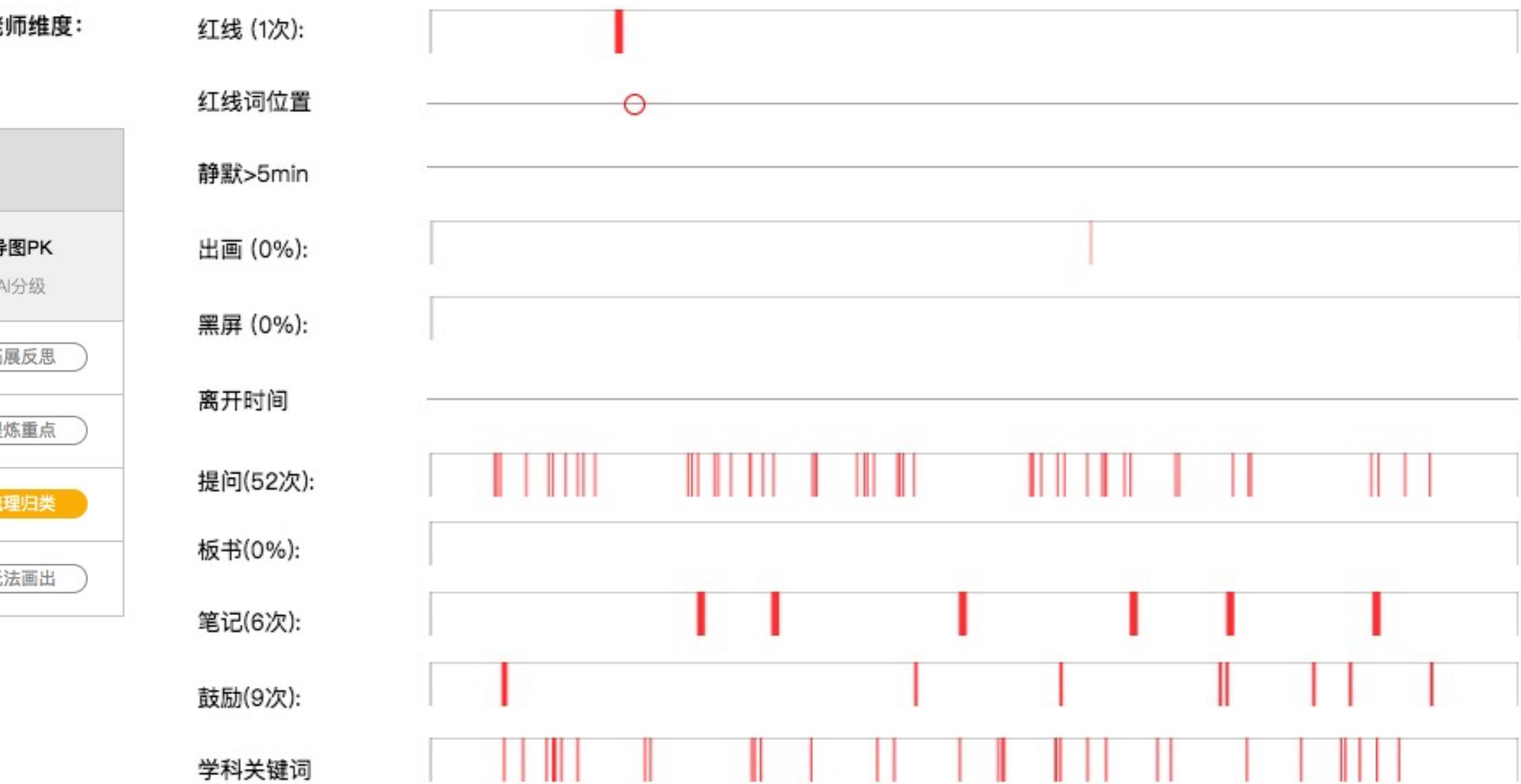
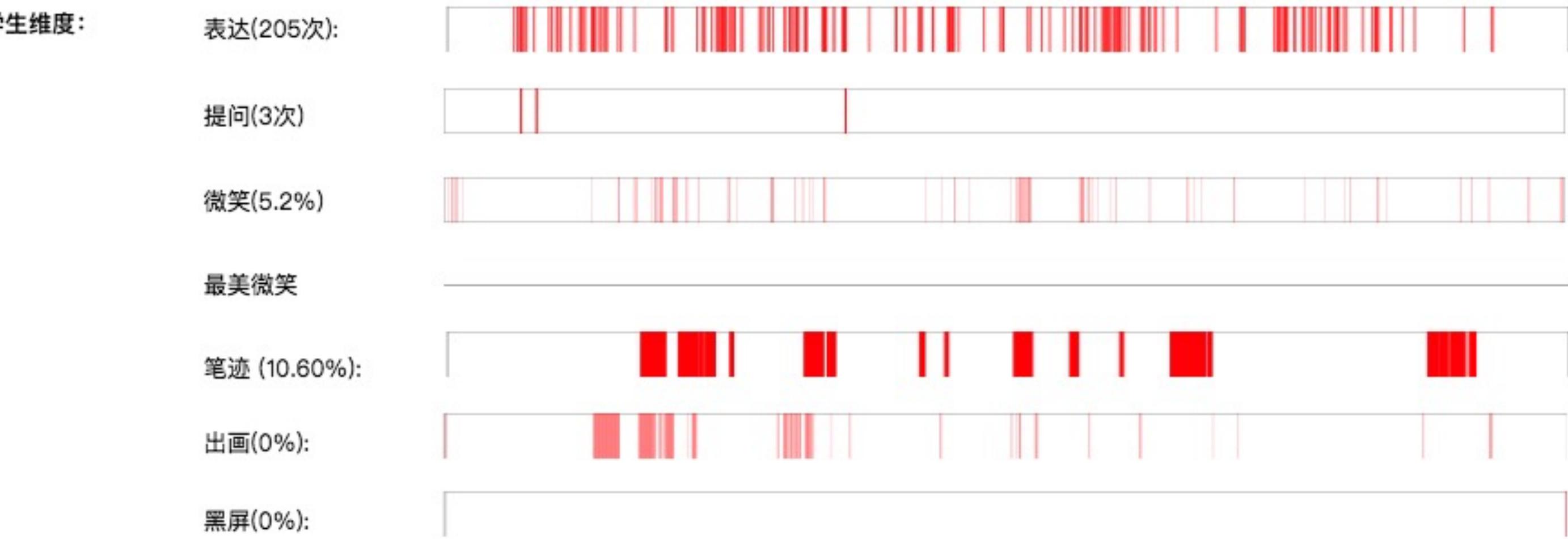
光的折射作图主要依据折射规律，常见题型有以下两点需要注意：
 ①要求会画对应的入射光线、法线、折射光线的光路图；
 ②斜射时，总是空气中的角大（无论是光从空气进入水中或是从水中进入空气）；
 ③常见的折射现象为光在空气与水或空气与玻璃之间，而水和玻璃中可认为折射情况一样。

15 如图，一束光线从空气斜射入水中，在水面处同时发生了反射和折射，OB是其反射光线。请在图中画出其对应的入射光线和折射光线。



Instructor
Student

Solution



	愿表达		勤动笔		善总结	
评价项	问答 AI分级	讲题 AI分级	笔记 AI分级	做题 老师分级	错误纠因 AI分级	导图PK AI分级
1	提问	新方法见解	关联其他	独到简单方法	设法改进	拓展反思
	回答完整	独立完成	重点突出	完整做出	完成阐述	提炼重点
	回答不完整	辅助完成	完成记录	部分完成	指出原因	梳理归类
	不回答	不能完成	无笔记	无从下手	错因不明	无法画出

AI + Education

Challenges and **MORE** opportunities!

Thanks!