

NGUARD: A Game Bot Detection Framework for NetEase MMORPGs

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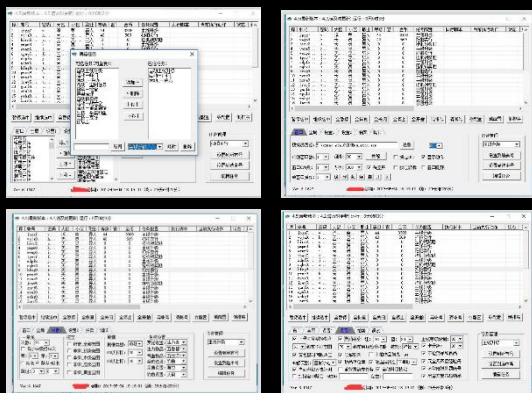
June 13th, 2018

MMORPG反外挂背景

MMORPG 大型多人在线角色扮演游戏



游戏外挂客户端截图



游戏外挂脚本界面

外挂严重破坏游戏的生态圈
降低游戏的公平性和可玩性

Main quest

创建
账号



Daily quest

登陆
游戏



登出
游戏

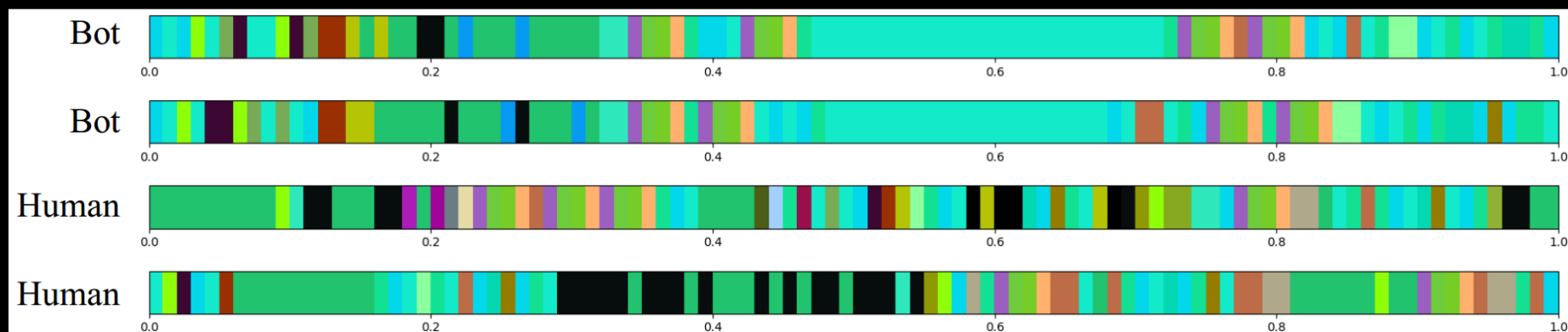
Instanced quest

进入
场景

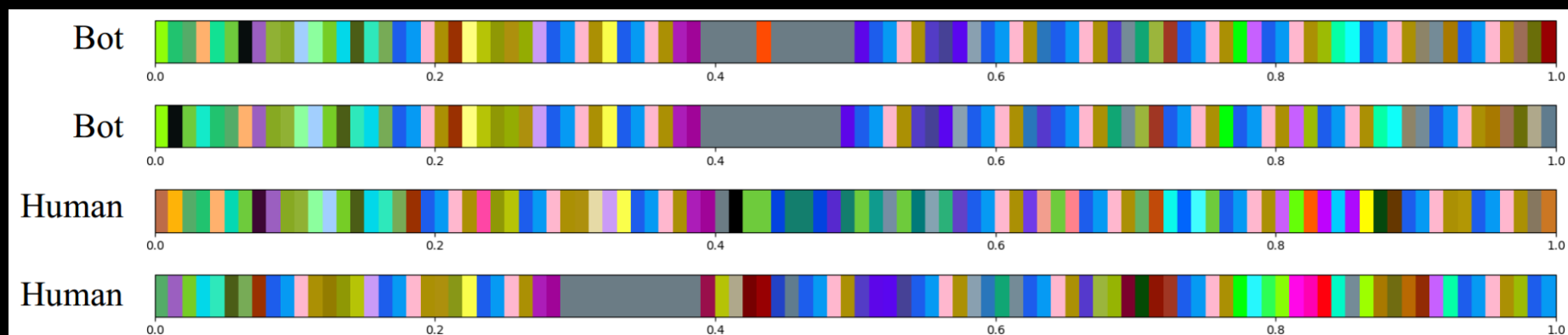


离开
场景

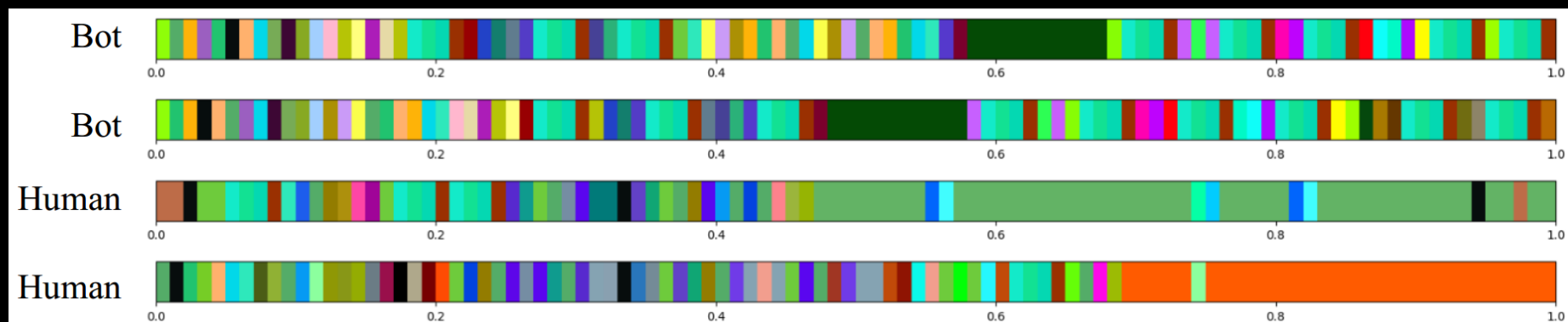
Main quest



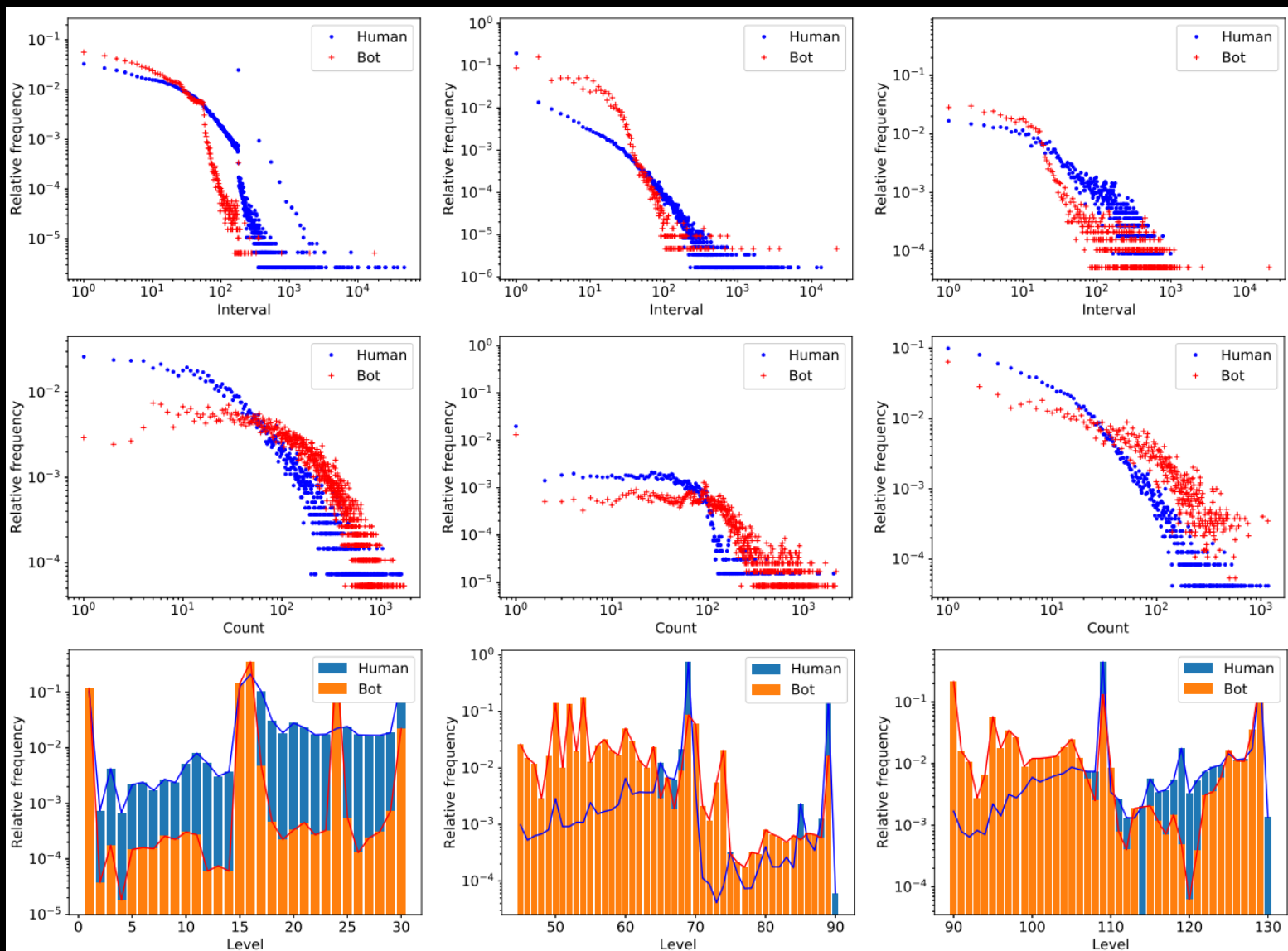
Daily quest



Instanced quest



Interval



Count

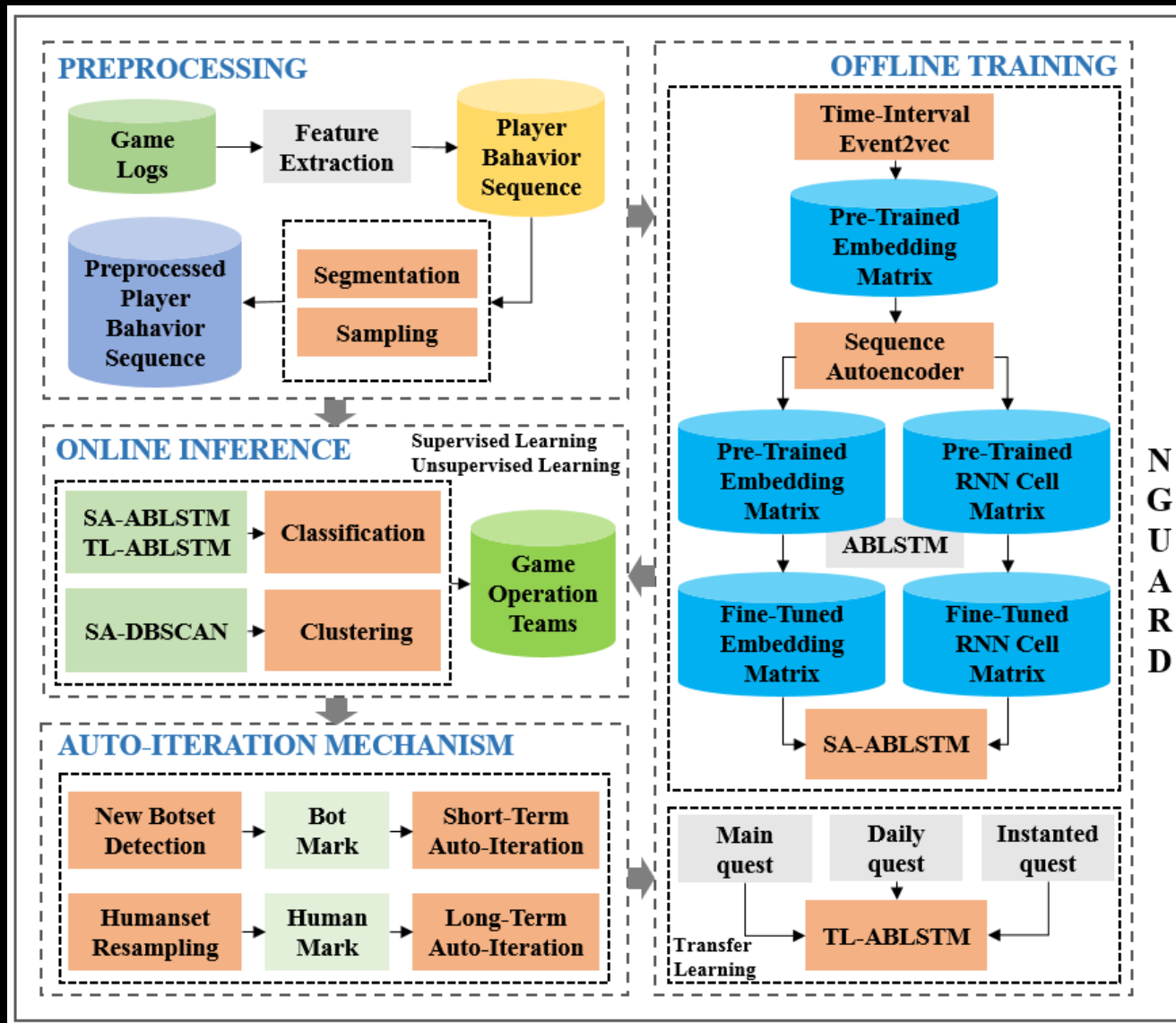
Level

Main quest

Daily quest

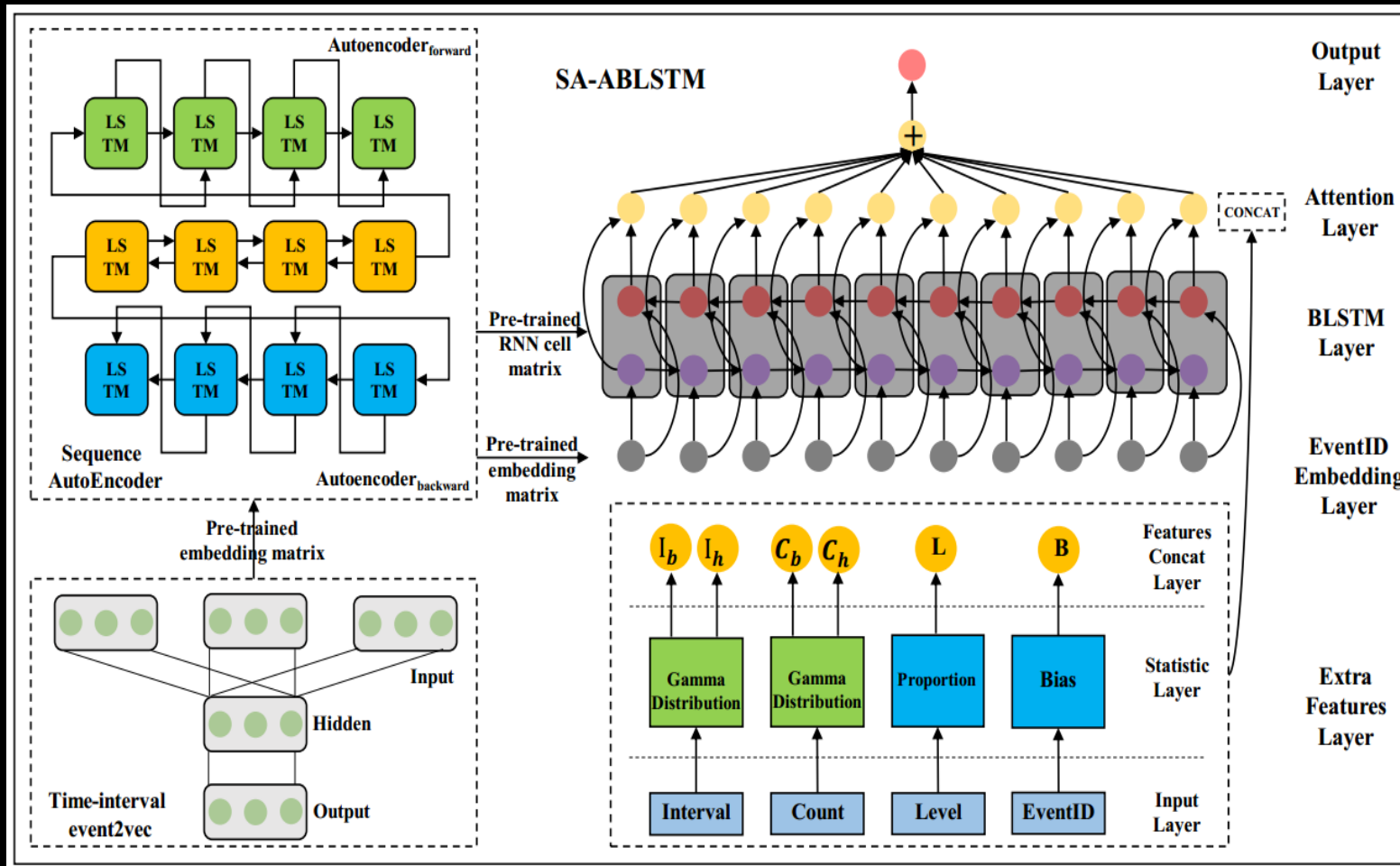
Instanced quest

Framework



- **Preprocessing**
segment
sampling
- **Offline Training**
training the models offline
- **Online Inference**
online service
- **Auto-iteration mechanism**
short-term
long-term

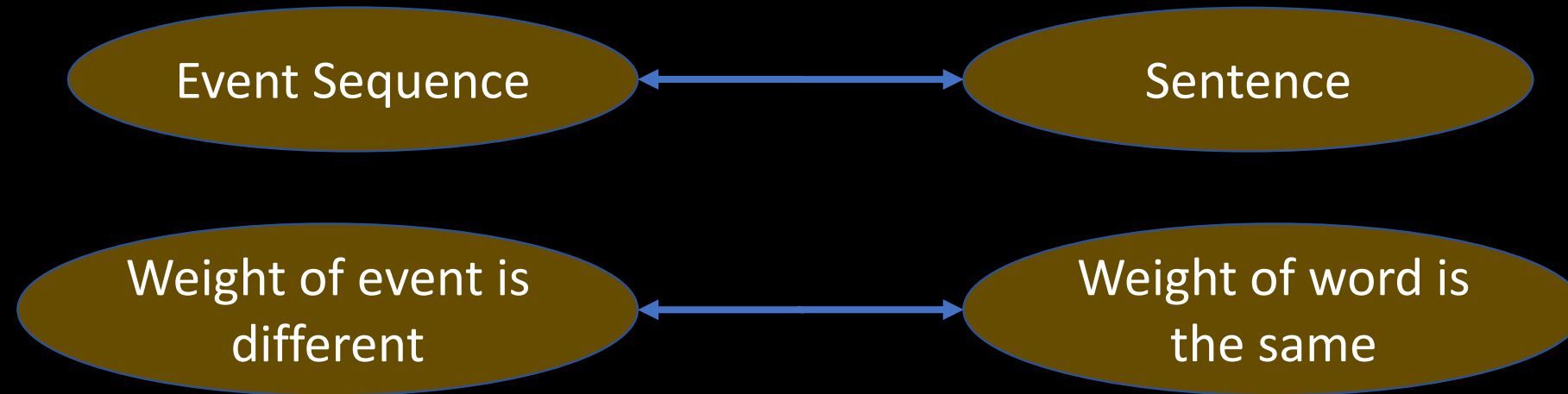
Offline Training



- **Pre-training**
Time-interval Event2vec
Sequence Autoencoder
- **Modeling**
SA-ABLSTM
- **Transfer-learning**

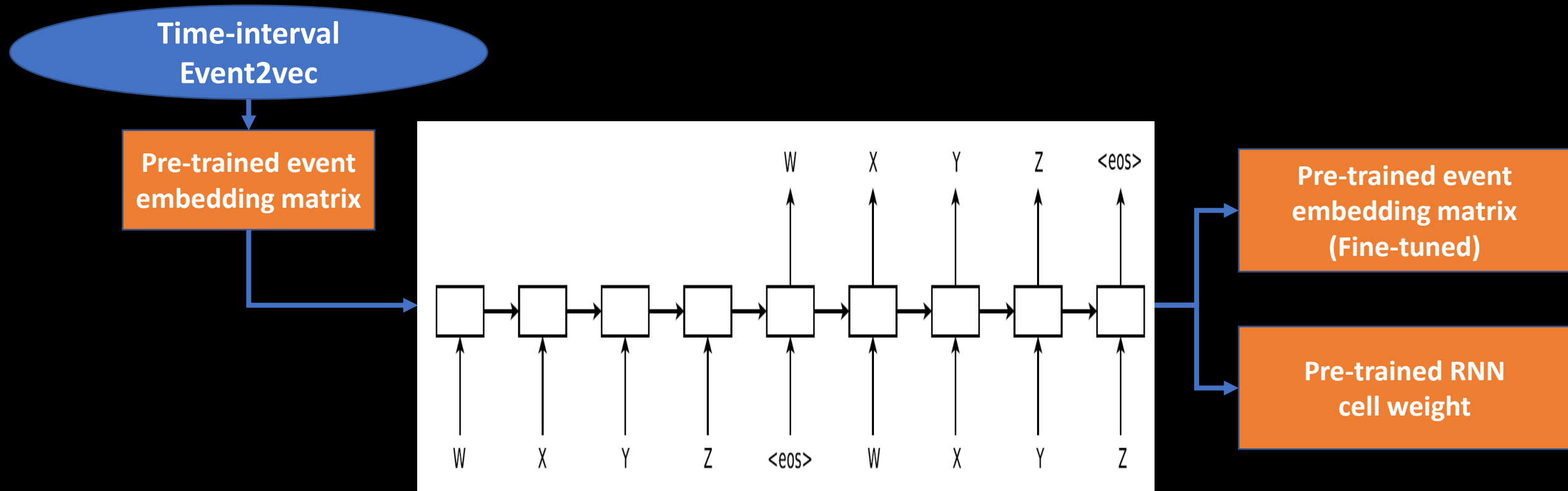
Time-interval Event2vec

Word2vec

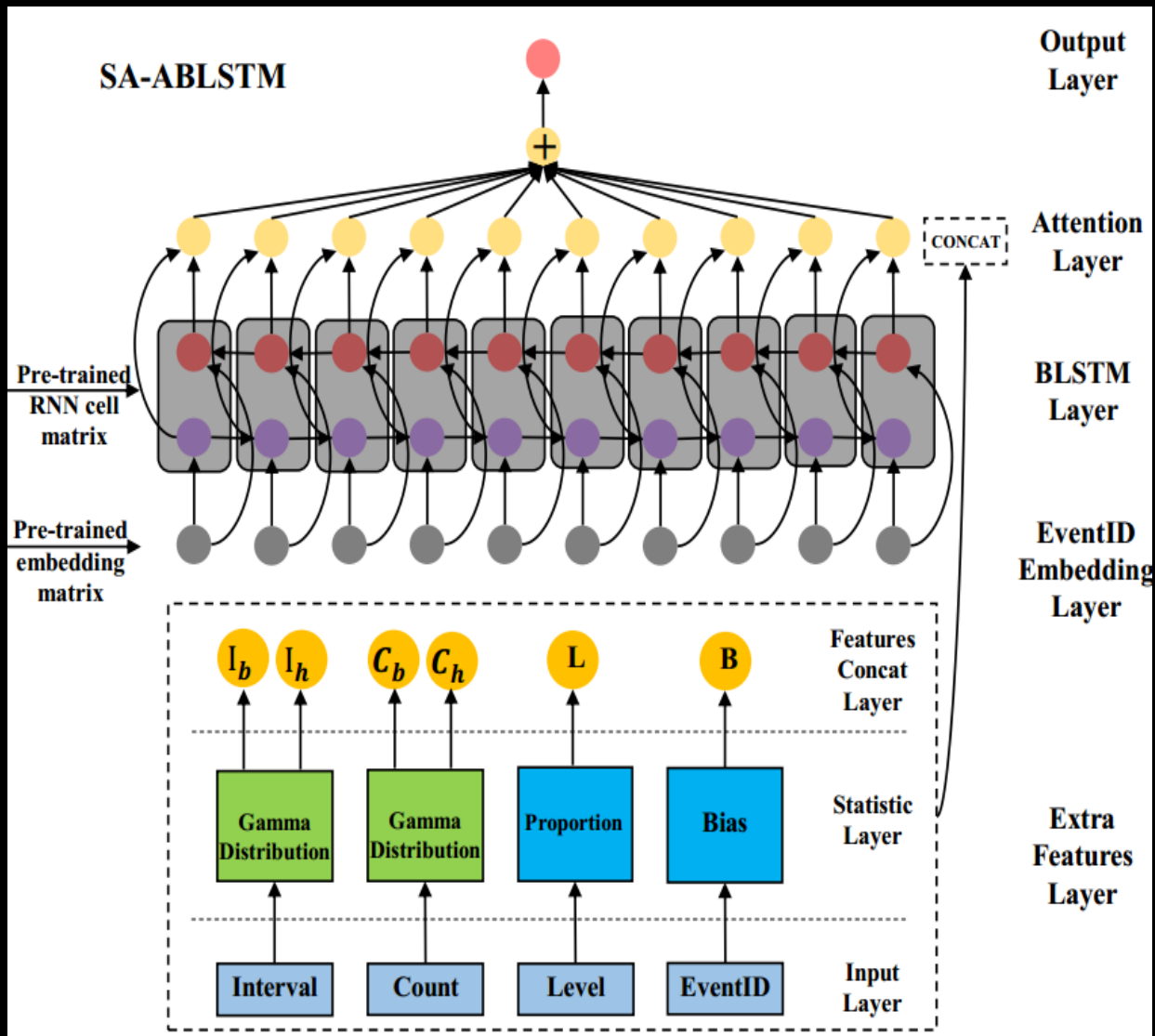


- Time-interval Event2vec considers the time elapsed between two events.
- The event within shorter time gaps to the target event should contribute more in predicting the target event.

Sequence Autoencoder



SA-ABLSTM



- **Input**
Event sequence $\{E_1, \dots, E_n\}$
Extra features: Interval, Count, Level
- **Bi-LSTM for Event sequence**
- **Knowledge-based method for extra features**

Knowledge-based method for extra features

- Interval

Gamma distribution : $\text{Interval}_h \sim \Gamma(\alpha_{i,h}, \beta_{i,h})$ $\text{Interval}_b \sim \Gamma(\alpha_{i,b}, \beta_{i,b})$

$$I_{i,h} = f(t_i; \alpha_{i,h}, \beta_{i,h}) , I_{i,b} = f(t_i; \alpha_{i,b}, \beta_{i,b})$$

- Count

Gamma distribution : $\text{Count}_h \sim \Gamma(\alpha_{c,h}, \beta_{c,h})$ $\text{Count}_b \sim \Gamma(\alpha_{c,b}, \beta_{c,b})$

$$C_{i,h} = f(c_i; \alpha_{c,h}, \beta_{c,h}) , C_{i,b} = f(c_i; \alpha_{c,b}, \beta_{c,b})$$

- Level

$$L_i = P(l_i) = \frac{N_{bot}^{l,e}}{N_{bot}^{l,e} + N_{human}^{l,e}}$$

- bias

$$B_i = P(e_i) = \frac{N_{bot}^e}{N_{bot}^e + N_{human}^e}$$

- Output of Bi-LSTM: $H=[h_1, ..., h_n]$

- Concat H with the statistical values:

$$h'_i = h_i \circ I_{i,h} \circ I_{i,b} \circ C_{i,h} \circ C_{i,b} \circ L_i \circ B_i$$

- Self-attention:

$$a_i = \text{softmax}(w_i^T H' + b_i) , r_i = a_i \odot h'_i$$

- **Supervised solution**

Classification: SA-ABLSTM

- **Unsupervised solution**

Clustering: SA-DBSCAN

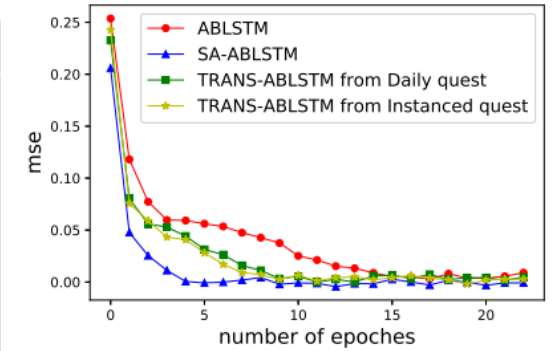
1. Extract the vector representation of event sequences from the well-trained Sequence Autoencoder
2. Perform DBSCAN on the vector representation of event sequence

- **Humanset Resampling**
- **New Botset Detection**
Known game bots, **Mutated game bots**, **Unknown game bots**
clusters of SA-DBSCAN
- **Short-term Auto-iteration**
ABLSTM
- **Long-term Auto-iteration**
SA-ABLSTM

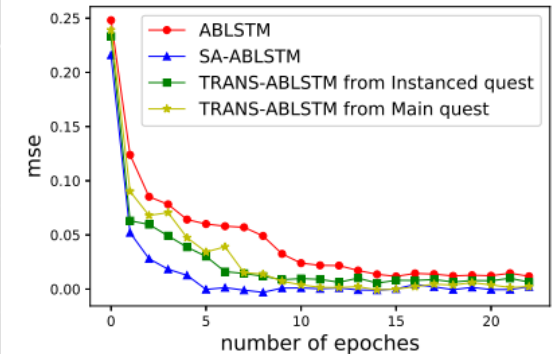
Experiments

Comparisons of supervised solutions

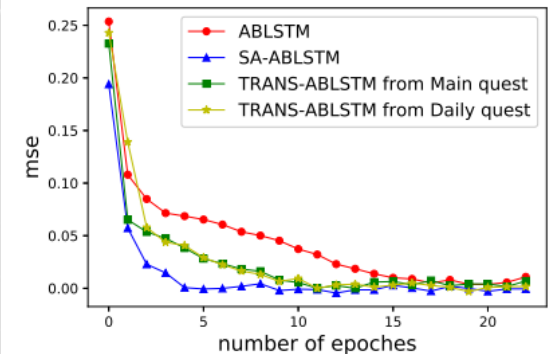
Type of game bot	Model	Precision	Recall	F1
Main quest	MLP	0.9618	0.9773	0.9694
	CNN	0.9721	0.9807	0.9764
	Bi-LSTM	0.9809	0.9865	0.9837
	ABLSTM	0.9851	0.9882	0.9866
	TL-ABLSTM from Daily quest	0.9893	0.9906	0.9902
	TL-ABLSTM from Instanced quest	0.9878	0.9896	0.9881
	SA-ABLSTM	0.9904	0.9912	0.9908
Daily quest	MLP	0.9528	0.9609	0.9568
	CNN	0.9633	0.9712	0.9672
	Bi-LSTM	0.9709	0.9728	0.9718
	ABLSTM	0.9716	0.9774	0.9745
	TL-ABLSTM from Main quest	0.9771	0.9742	0.9785
	TL-ABLSTM from Instanced quest	0.9736	0.9721	0.9763
	SA-ABLSTM	0.9838	0.9861	0.9815
Instanced quest	MLP	0.9441	0.9571	0.9506
	CNN	0.9552	0.9643	0.9597
	Bi-LSTM	0.9612	0.9732	0.9672
	ABLSTM	0.9674	0.9786	0.973
	TL-ABLSTM from Daily quest	0.9698	0.9801	0.9742
	TL-ABLSTM from Main quest	0.9704	0.9808	0.9753
	SA-ABLSTM	0.9721	0.9816	0.9768



(a) Main quest



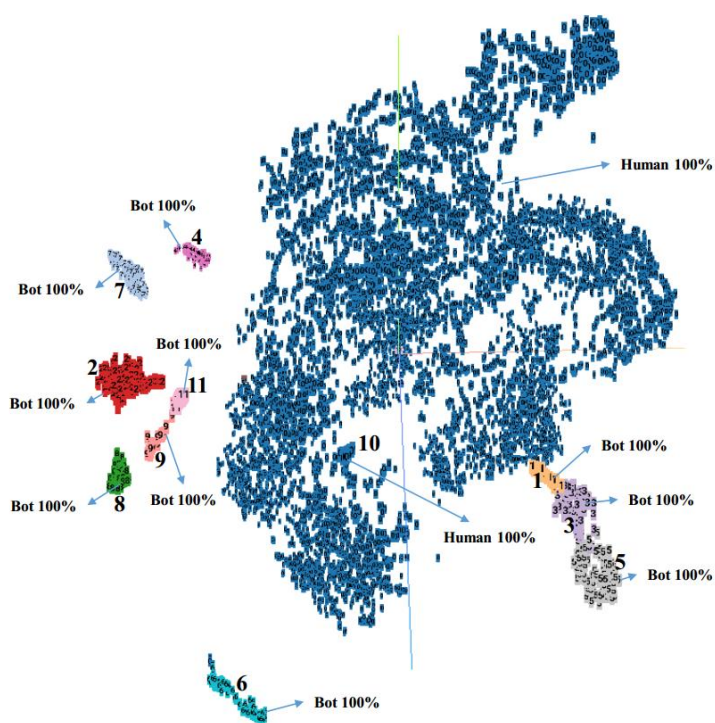
(b) Daily quest



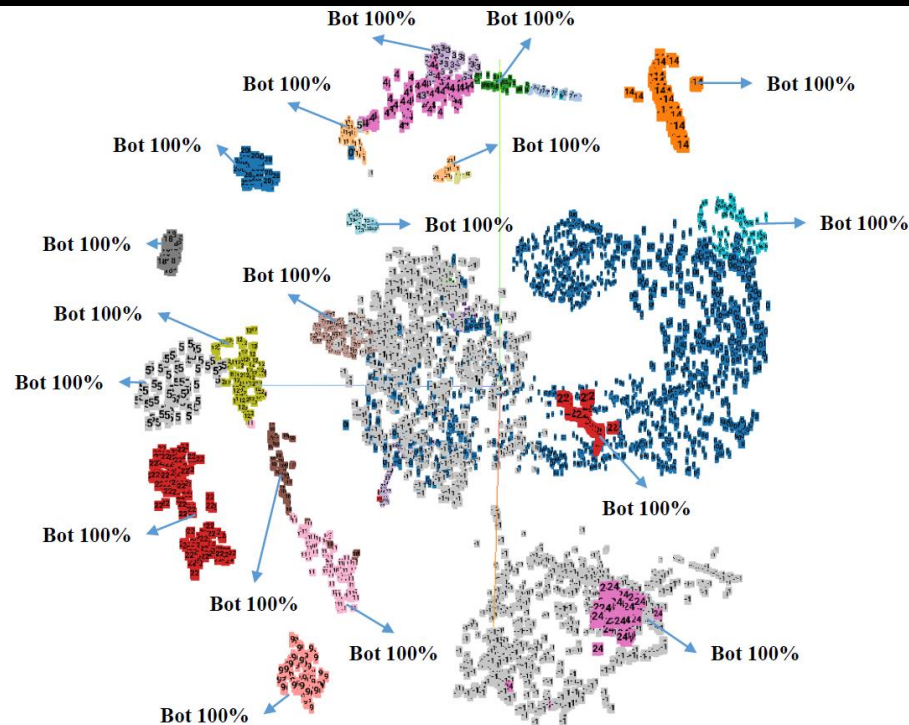
(c) Instanced quest

Experiments

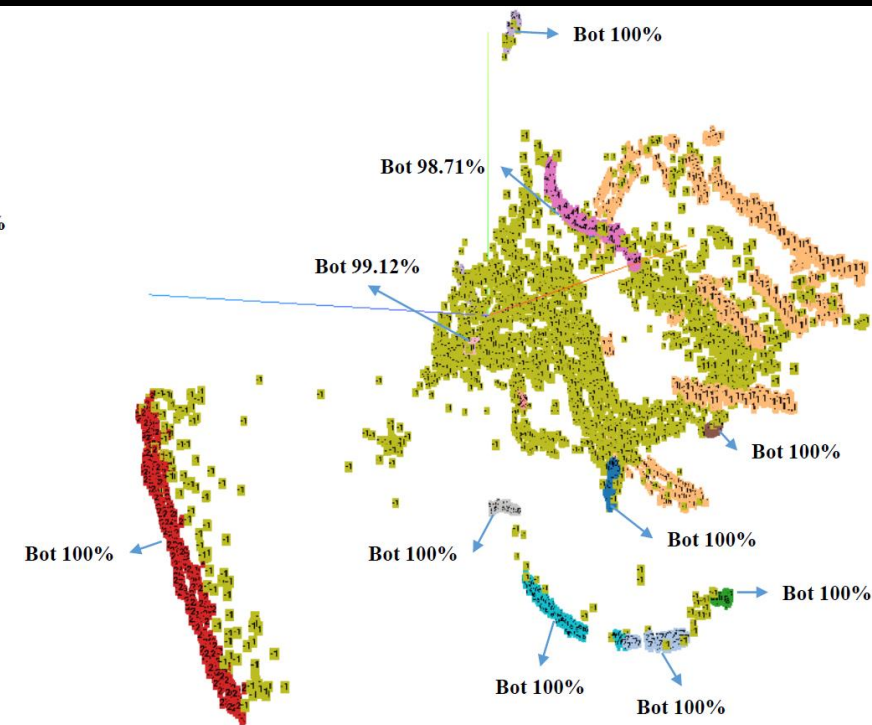
Evaluations of unsupervised solution: SA-DBSCAN



Main quest

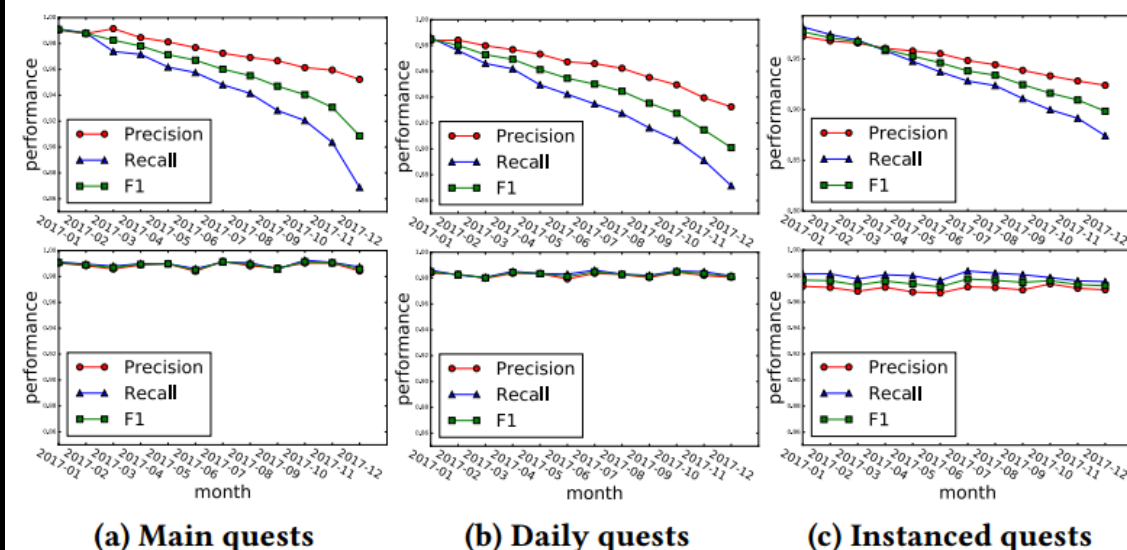
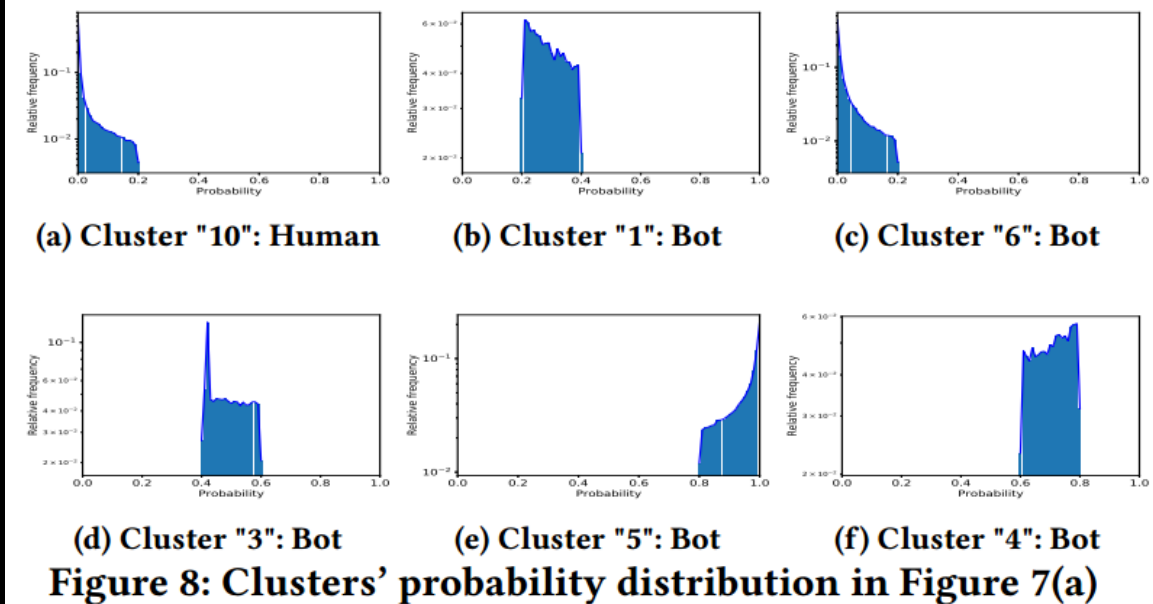
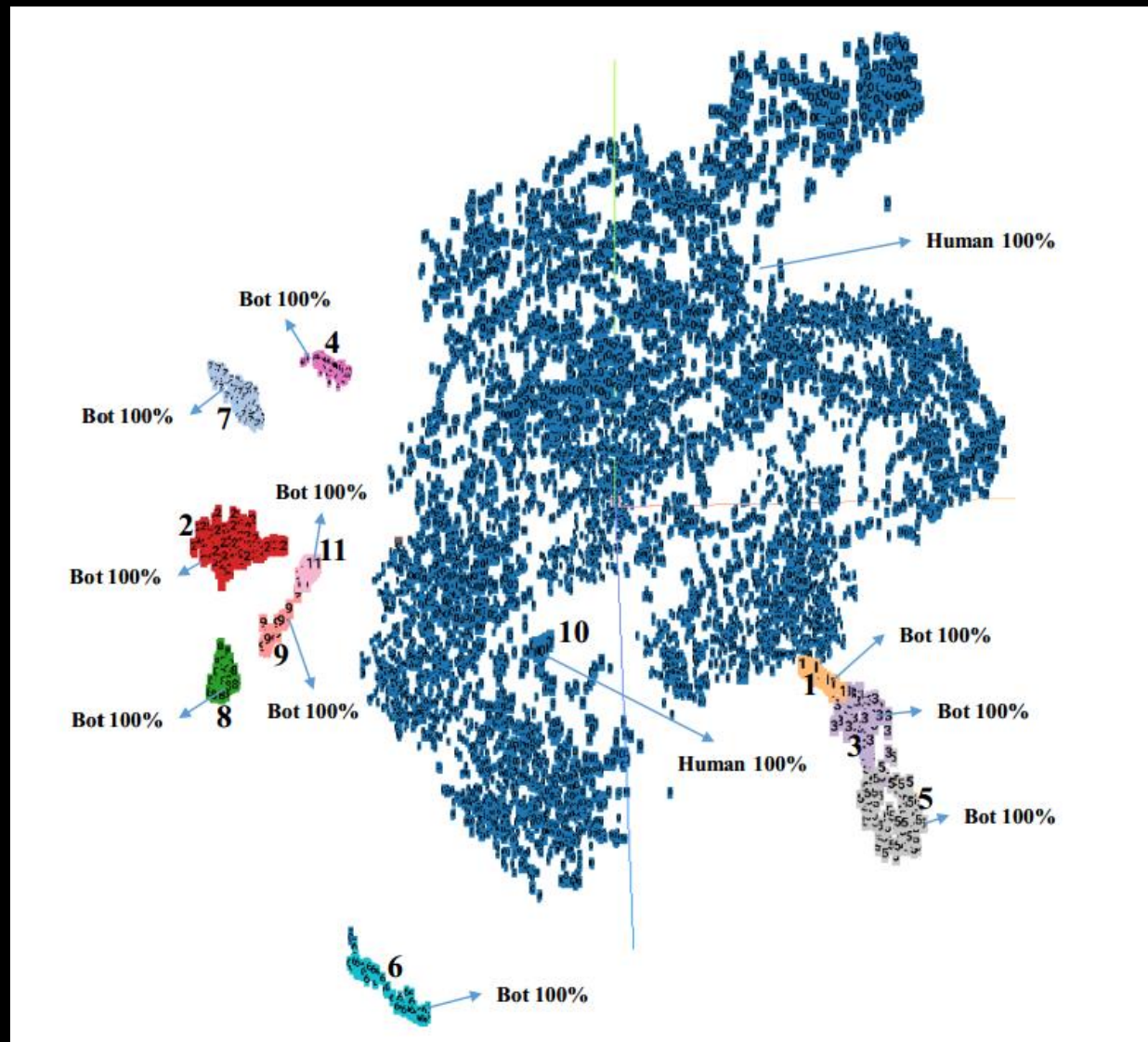


Daily quest



Instanced quest

The necessity of the auto-iteration mechanism



谢谢