

# **THUIR at the NTCIR-15 Micro-activity Retrieval Task**



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<b>*Task Definition</b>	*3. Models
Recognize 20 types of Micro-activities	Ranking Model
Ranking Task	Test Instance <i>I</i>
> Input: query $Q_i$ – Finding activity i	Kanking score
Model: Similarity-based ranking model	$ \begin{array}{c c} & & & \\ \hline & & \\ \otimes & \otimes & \\ \end{array} \end{array} $ Feature vector vector

 $\succ$  **Results:** Ranking list of instances  $I_i$  according to similarity  $P(I_i | Q_i)$ 

#### **Classification Task**

- > Input: Instance  $I_i$  Classify instance to a type of activity
- > Model: Multi-level classifier
- $\succ$  **Results:** Probability that  $I_i$  belongs to activity i  $Q_i P(Q_i | I_i)$

## ➤Task Equivalency:

$$P(I_j \mid Q_i) = rac{P(Q_i \mid I_j) P(I_j)}{\sum_{k=1}^n P(Q_i \mid I_k) P(I_k)} \propto P(Q_i | I_j)$$

## **\$1.** Feature Extraction

### > Visual Features:

- > Photos histogram similarity between instance & in instance
- > Concepts and tags from multiple object detection methods

## > OCR Features:

- > URL extraction from screenshot
- > OCR detection of the URL and head

#### > Temporal Features:



### > Instance Score

- Similarity between Instance I and I':  $S_{I,I'} = \sum_{i=1}^{n} \alpha_i * i_i \bigotimes i'_i$
- **Ranking Score**:
  - $\succ$  Similarity between Instance I and Activity J:

$$S_{I,A} = \sum_{I' \in A} S_{I,I'} + \beta * \sum_{I'' \in A_u} S_{I,I''}$$

**Classification Model** 

#### > Basic model selection:

Classifier	LR	SVM	MLP	RF	XGboost	GBDT
Accuracy	0.825	0.821	0.811	0.779	0.826	0.836

Feature type	Extraction Approaches
EOG	Time-domain statistic + Frequency-domain energy +LPC coefficients -> 62 features
Acceleration	3 position * 3 axes * Time-domain statistic + frequency-domain spectrum + axis correlation -> 523 features
Heart rate	Time-domain statistic -> 26 features
Mouse movement	Time-domain statistic + number of peaks -> 14 features

**\*2.** Feature Selection

- > Chi2: Choose features with larger variance.
- > Correlation: Remove similar features (high correlation)
- **GBDT:** Select features according to importance in GBDT
- **PCA:** Use Principal Component Analysis(PCA) to reduce dimension of visual features.

- $\succ$  GBDT performs the best in all basic models.
- > Rules detection:
  - $\succ$  Partitioning the activities to find rules.
  - > Screen-relevant activities are similar, and can be classified with URL
  - $\succ$  Static activities, i.e. Zoning out and closing eyes, are similar.

#### > Rule-based Multilevel classifier:



## **\*4. Results**

Classifier	Accuracy	mAP (classify)	mAP (ranking)	Submission results
<b>Basic GBDT Classifier</b>	0.836	0.901	0.947	0.895
<b>Ranking Model</b>	0.789	0.843	0.836	0.782
Two-level Classifier (Cluster partition)	0.796	0.88	0.931	0.886
Two-level Classifier (Impurity partition)	0.875	0.921	0.971	0.901
Two-level Classifier (Similarity partition)	0.875	0.926	0.97	0.928
<b>Rule-based Classifier</b>	0.889	0.933	0.974	0.95

#### > Hybrid: Combine the above rules

#### Results on different feature selection method



#### > Rule-based multi-level classifier performs the best

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