

Automatic Construction of Expiration Time Expression Dataset from Retweets

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1. Introduction

- **Problem:**

Distinguishing time expressions that specify expiration times.

"The deadline is 23:00." ✓

"Campaign began at 23:00." ✗

- **Our Contribution:**

Developed a method for automatically generating a labeled dataset of expiration time expressions.

- **Expected Applications:**

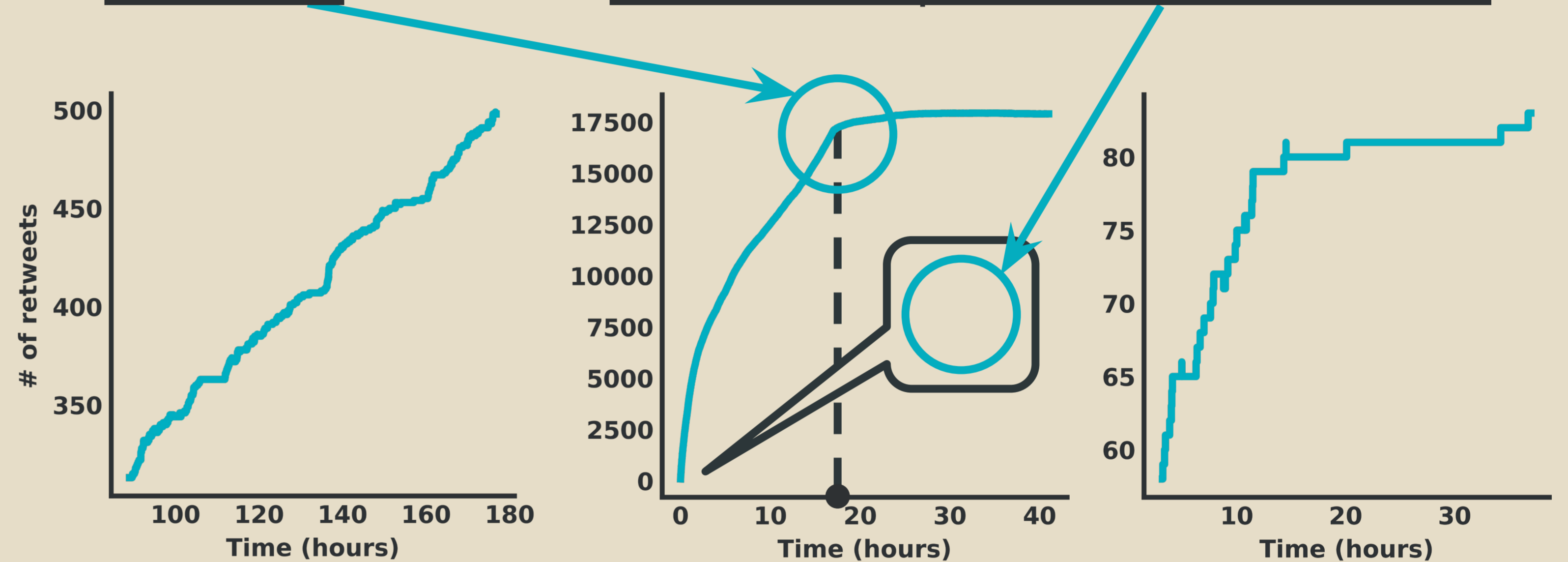
- Detection of approaching deadlines.
- Filtering of obsolete information.

2. Methods

I. **Tracking:** Monitoring tweets with time expressions.

II. **Detection of Expiration Time Specifications:**

Determine if the inflection point of the number of retweets matches the time expression in the tweet.



Without expiration time

With expiration time

No clear inflection point

III. **Dataset Construction:** Label the matching tweets as positive and others as negative.

3. Experiment

- **Dataset:**

Collected 25,046 tweets with time expressions, labeled them both manually and automatically.

- **Classifier Training and Evaluation:**

- Trained classifiers using automated labels.
- Compared performance with classifiers trained on manually assigned labels.

Achieved 70.7% accuracy.

Achieved 75.9% of Manual's improvement over Random.

| Model | Precision | Recall | F1-score | Accuracy | MCC |
|-----------------|-----------|--------|----------|----------|--------|
| Random | 0.255 | 0.508 | 0.339 | 0.499 | -0.001 |
| Our Method | 0.823 | 0.455 | 0.580 | 0.707 | 0.520 |
| Manual Labeling | 0.750 | 0.629 | 0.676 | 0.773 | 0.576 |
| Gain Ratio | 1.15 | -0.438 | 0.715 | 0.759 | 0.903 |

4. Conclusions

- **Effectiveness:**

- Demonstrated reliable dataset creation using proposed method.
- Reduced manual labeling effort significantly.

- **Future Work:**

- Refine accuracy.
- Expand the dataset for broader applications.

