

# A KERNEL-BASED APPROACH FOR IRONY AND SARCASM DETECTION IN ITALIAN

UNITOR @ evalita2018

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UNITOR approach for Irony and Sarcasm detection

- Task is modeled as a cascade of kernel-based SVM classifier

**AIM:** derive an effective task modeling

- Each kernel function is based on a specific set of irony/sarcasm specific features

The adopted modeling allows to rank **first** in the sarcasm detection task

# SOME ISSUES IS IRONY/SARCASM DETECTION AND (POSSIBLE) SOLUTIONS

 Lexical Sparseness (especially in short messages and reduced datasets)

 Negative evaluation towards a target

 Irony is often not explicit in (short) messages

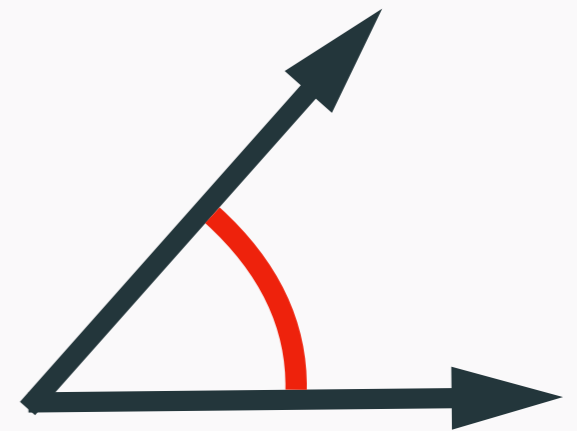
# IRONY/SARCASM FEATURES

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**Goal:** define capture irony features

- Irony Specific BOW (*ISBOW*)

Cosine similarity between the word embedding vector of the considered word and mean topic centroid



“La mia macchina è veloce e scattante come un *bradipo*”

# POLARITY BASED FEATURES

**Goal:** exploit negative evaluation towards a target

**DPL:** large scale Polarity Lexicon (~150K)

**Buono:** **0.7** **0.3** **0.0**

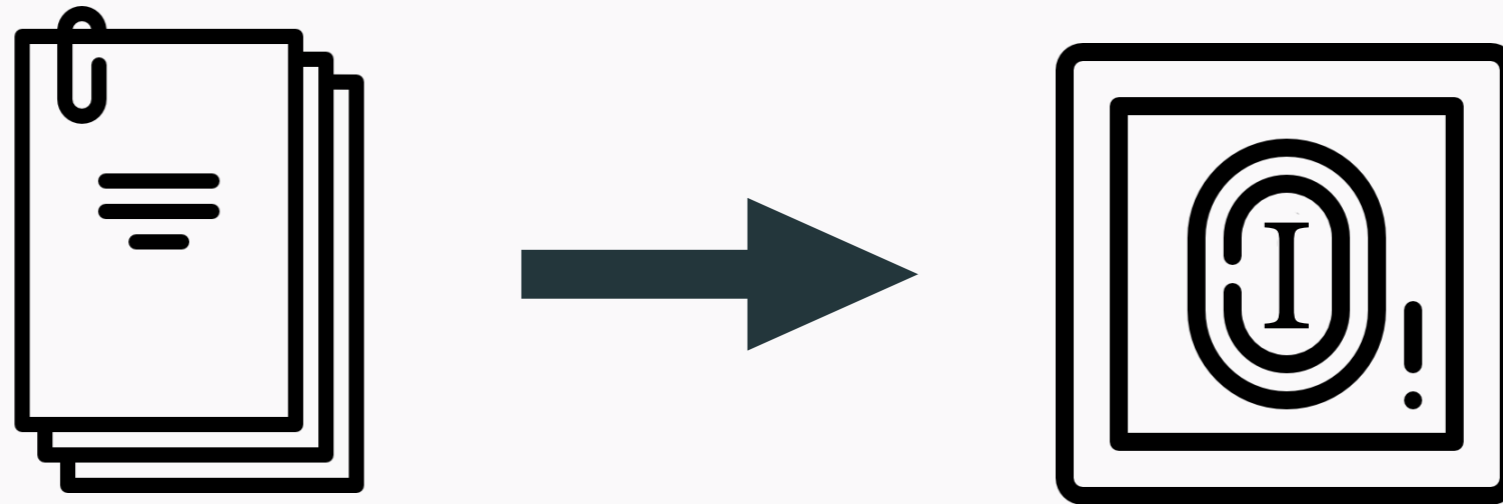
- Distributional Polarity Lexicon Sum (*DSUM*)

$$\sum_{w \in Noun} \frac{w^p}{|Noun|} \oplus \sum_{w \in Verb} \frac{w^p}{|Verb|} \oplus \sum_{w \in Adj} \frac{w^p}{|Adj|} \oplus \sum_{w \in Adv} \frac{w^p}{|Adv|} \oplus \sum_{w \in \tau} \frac{w^p}{|\tau|}$$

- Distributional Polarity Lexicon BoW (*DBOW*)

buono\_pos: **0.7** buono\_neu: **0.3** buono\_neg: **0.0**

**Goal:** define an Irony specific Lexicon and Irony patterns



- Irony Corpus BOW (ICBOW)
- Irony Corpus weighted BOW (ICwBOW)
- Irony Corpus weighted Mean (ICM)

# EXPERIMENTS AND RESULTS

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Training Set		
Ironic	Not-Ironic	Total
2023	1954	3977

**10-fold** cross validation

Train effettuato sullo split di:

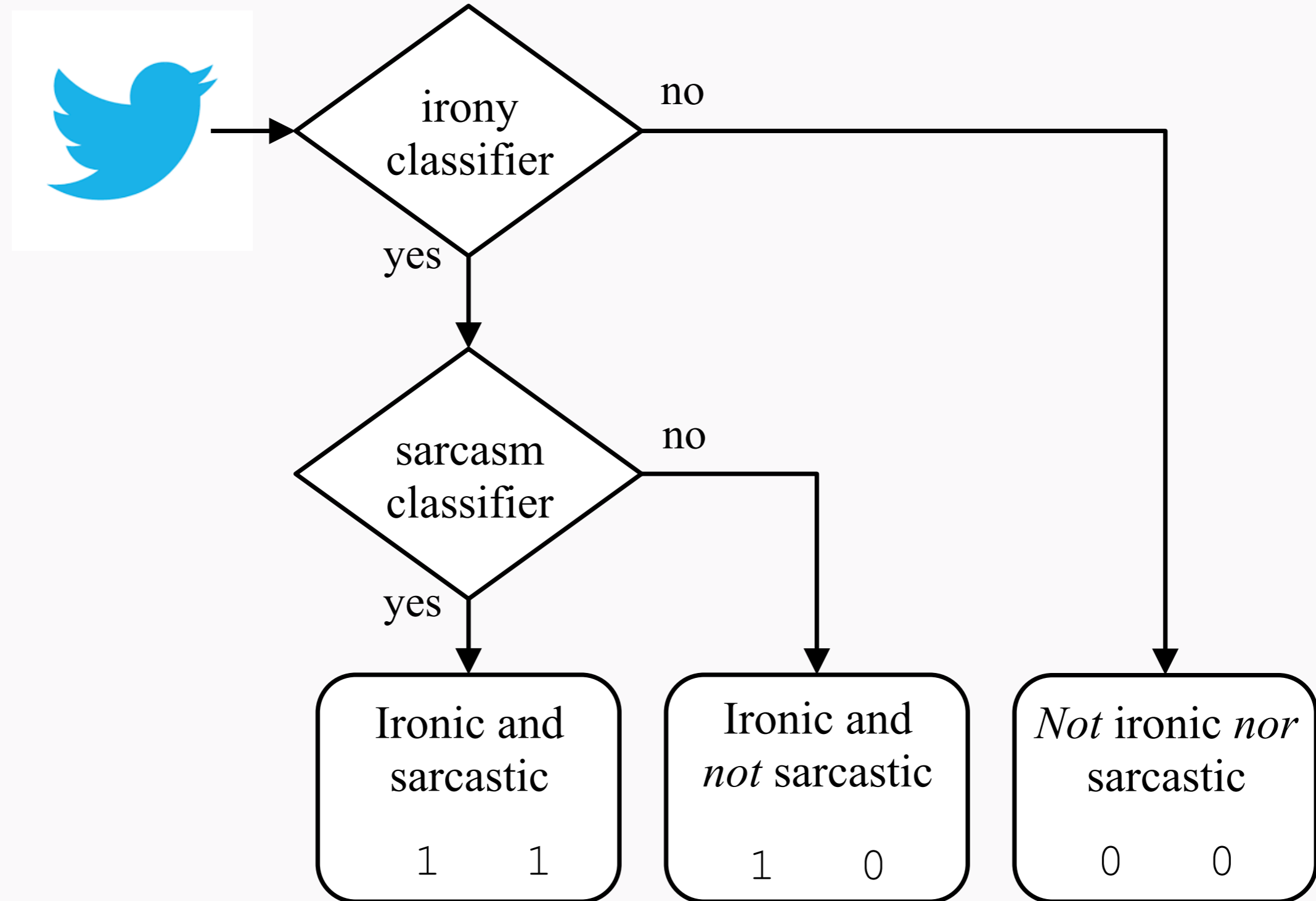
- **Irony:** All dataset (3977)
- **Sarcasm:** Ironic tweets (2023)

Each feature modeling generate a separate feature space

We also added some “standard” feature modeling

- Bag-of-Words
- Bag-of-Word Ngrams
- Bag-of-Character Ngrams
- Some syntethic feature (capitalized letters, elongations, ...)

The resulting system is implemented as two binary classifiers in cascade



## Irony Detection

	Not Irony			Irony			Mean
	P	R	F1	P	R	F1	F1
1st	.785	.643	.707	.696	.823	.754	<b>.731</b>
2nd*	.771	.617	.686	.680	.816	.741	.714
6th( <i>u</i> )	.778	.577	.662	.662	.834	.739	.700
7th( <i>c</i> )	.764	.593	.668	.666	.816	.733	.700
BL	.501	1.00	.668	1.00	.000	.000	.334

## Sarcasm Detection

	Not Sarcastic			Sarcastic			Mean
	P	R	F1	P	R	F1	F1
1st( <i>c</i> )	.362	.584	.447	.492	.407	.446	<b>.520</b>
2nd( <i>u</i> )	.355	.553	.432	.469	.449	.459	.518
4th*	.344	.566	.428	.344	.566	.428	.508
BL	.296	.132	.183	1.00	.000	.000	.199

# CONCLUSION

- 😎 UNITOR ranked **first** in the sarcasm detection task
- 😎 The proposed irony and sarcasm specific features, especially **Polarity Based Features**, were beneficial to detect irony and sarcasm
- 😬 Non trivial for human reader, lot of information *out of context*

# Thanks for your attention



**Code available:** <https://github.com/andry9454/ironySarcasmDetection>