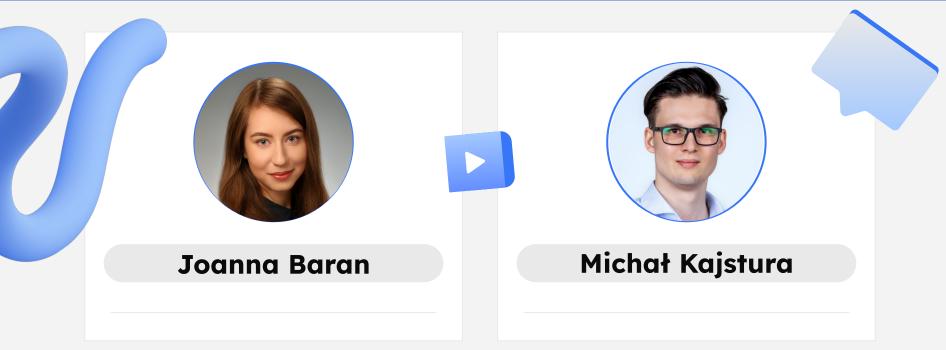


## Nasz zespół





Opiekunowie: prof. dr hab. inż. Przemysław Kazienko, dr inż. Jan Kocoń





## Jak udoskonalić Twittera?

Wyobraźmy sobie świat social mediów skrojony idealnie pod nas...







### @asia065

Publikuje post



### @janusz666

Wystawia złośliwy nic niewnoszący komentarz



### @mihalkaj

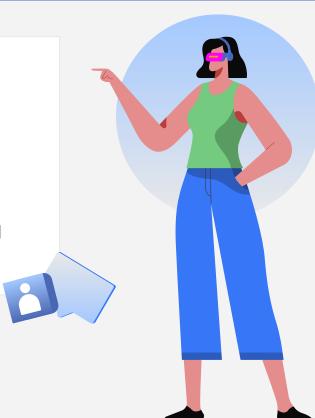
Pisze konstruktywny miły komentarz



### Każdy z nas spotkał się na Twitterze z:

- mową nienawiści
- tematyką, która wzbudza w nas niekoniecznie dobre emocje
- nieciekawymi wpisami (marnowanie czasu)

Rozwiązanie: kreowanie kontentu dopasowanego pod indywidualną osobę, swoiste "różowe okulary"



# Co mógłby zrobić właściciel Twittera?





### Filtracja



Blokowanie treści uznanych za ofensywne przy użyciu ogólnych modeli do detekcji mowy nienawiści.

### Wolność słowa



Należy jednak pamiętać o drugiej stronie medalu — prawa do wolności słowa. Modele ogólne go nie uwzględniają...







77%



przyrostu usuniętych kont do poprzedniego półrocza

1,126,990

ukaranych kont

3,8 mln

usuniętych postów

Dane z II połowy 2020 r. - źródło: time.com

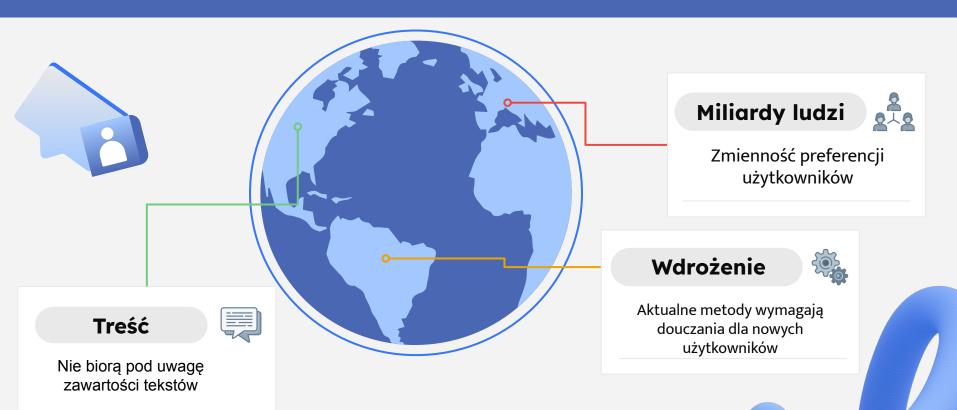


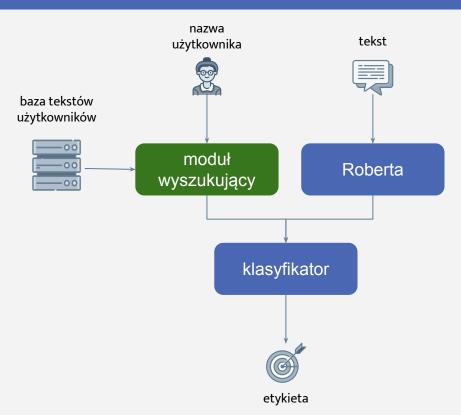
# Personalizacja!

Zamiast bezwzględnej cenzury, dopasujmy wyświetlane treści pod konkretną osobę...

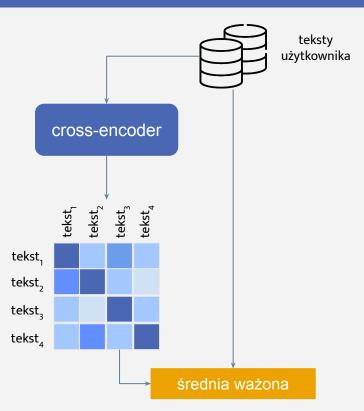
# Wady istniejących metod







- Wyszukuje podobne teksty użytkownika
- Średnia ważona dotychczasowych ocen





Nie trzeba douczać modelu



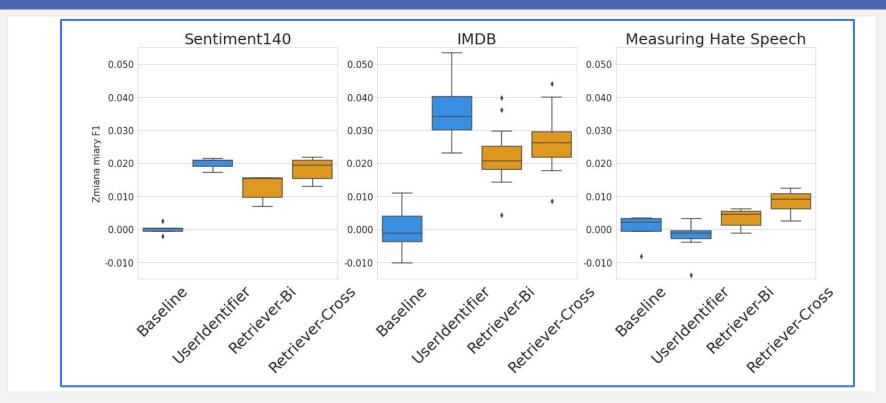
Bierze pod uwagę podobieństwo tekstów



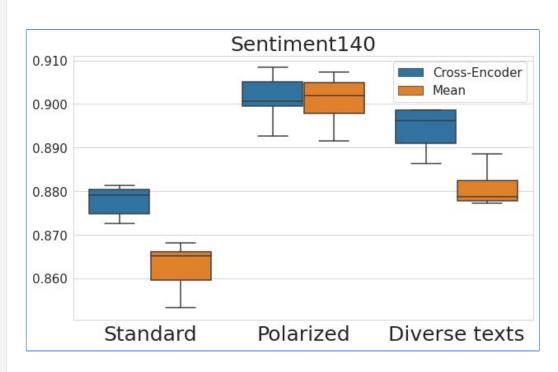
Szybciej się uczy



# Personalizacja pomaga



## Pozytywne scenariusze



 Przetestowaliśmy Retrievera na różnych podziałach danych

Użytkownicy "ekstremalni"

• Różnorodne teksty

#### Retrieval based approach for subjective tasks in NLP

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#### Abstract

Generalizing models are often unable to produce a personalized prediction for a individual user. Personalized methods often yield better results, but at the cost of an increased training complexity, especially in big-data scenarios, when frequent retraining is necessary. We propose a retrieval based approach that performs similar to SOTA personalization algorithms, but does not require any further optimization for new users.

#### 1. Introduction

Popular classical approaches to natural language processing tasks have for a long time focused solely on developing a general classifier that does not take into account the characteristics of a single person. However, everyday language is influenced by many individual factors like one's mood, emotion, world view, and sociodemographic circumstances. This has become the reason for the dynamic development of the personalization trend in NLP, which has been going on for over a dozen years. Perfect for subjective tasks, where obtaining a single label is hard or even impossible, human-based approaches were exploited in various ways. Starting with the inclusion of user traits, modelling entire social groups, ending with focusing on individual features - thus creating perfectly tailored NLP systems. From the user's perspective, personalization can occur as an explicit input, when the person is providing information themselves, or by using implicit inference made by specially designed models (Flek 2020). In the scope of this research, we consider the last approach since it is more convenient for users (no additional action is required on their part) and thus better applicable in business.

The example of a highly subjective task is sentiment classification, where the polarity of a text depends on a person's experiences and a character. This existing methods require frequent re-training for each new user, which usually is done by sending the data to a centralized server. Although federated learning algorithms allow for in-device gradient update computation (Mahlool and Abed 2022), they are still not feasible to use in many situations, because of significant performance overhead, especially in big-data scenarios (Gadekallu et al. 2021).

In this work, we propose a retrieval based approach, which provides personalized results using the text similarity and is easily extendable to the large number of users. We experiment with different ways of capturing the text similarity, including state-of-the-art biencoders and cross-encoders. The analysis performed on three datasets shows in which situations the personalized methods work best and where the improvements come from.

#### 2. Related Work

The existing approaches to human-based NLP can be divided into two main groups - based on users personal metadata, and focused on using their past digital traces such likes, ratings, posted texts in social media, etc. The first works that addressed the need to adapt NLP approaches to subjective language have appeared since the 2010s.





# Dziękujemy!

Czy są jakieś pytania?









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