### Gotong Royong in NLP Research A Mobile Tool for Collaborative Text Annotation in Indonesia

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# Introduction

- The Indonesian language Bahasa Indonesia counts as a "low-resource" language
- Machine learning technology advanced the development of NLP tools in Indonesia

#### BUT:

Machine learning based NLP methods depend on the availability of annotated training data

# Annotated Training Data

**Example – Named Entity Recognition** 

### MAY DAY: Buruh KSPI Ancam Mogok Kerja Jika Tuntuan Tak Digubris. Time Per Org

May Day: KSPI workers threaten to strike if their demands are ignored

# The Problem

A	D	0	U	L	1
Token	ACTOR	TRIGGE	TARGET	LOCATION	TIME
MAY					x
DAY					x
:					
Buruh	x				
KSPI	x				
Ancam		x			
Mogok					
Kerja					
Jika					
Tuntuan					
Tak					
Digubris					
Angkot					
di					
bogor					
pada					
mogok					
kerja					
,					
jalanan					
tuh					

### Manual annotation of data is

- tedious and
- time-consuming



# Solution Approach





#### Gotong Royong



#### Mobile First Culture



The first Mobile Collaborative Annotation Tool

## Gotong Royong

### Mobile First Culture



In today's Indonesia, 93% of online users access the Internet via their smartphone (Andrews et al., 2015)

### Solution



#### The First Mobile Collaborative Annotation Tool



# Shortcomings for their application in Indonesia1) Not Mobile Friendly

### **Existing Systems**

#### BRAT



BRAT, P. Stenetorp, 2012

### **Existing Systems**

GATE



#### GATE, H. Cunningham, 2011

Shortcomings for their application in Indonesia

- 1) Not Mobile Friendly
- 2) Interface does not support Bahasa Indonesia

# The Solution

- We propose a tool
- that makes data annotation more efficient
- allows data to be annotated by several users at the same time
- and can be used anywhere, anytime – using a mobile phone

Klik tombol label dan kemudian klik token kata   #NewsGibol Otamendi   Siapkan Aksi Mogok I   http://t.co/z1zFrzvU1G   Orang/Organisasi Lokasi Waktu Lainnya						
Klik tombol label dan kemudian klik token kata   #NewsGibol Otamendi Siapkan Aksi Mogok I http://t.co/z1zFrzvU1G   Orang/Organisasi Lokasi Waktu Lainnya						
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Orang/Organisasi Lokasi Waktu Lainnya	#NewsGibol Otamendi Siapkan Aksi Mogok   http://t.co/z1zFrzvU1G	)				
	Orang/Organisasi Lokasi Waktu Lainnya					
Tidak relevan Hapus >	Tidak relevan Hapus >					

#### **Example Binary Classification**

Is this Tweet related to labour strikes or protests?

Dear mahasiswa yg lg 'aksi' demo, bagus sih merjuangin hak rakyat. TAPI YA GAK NUTUPIN JALAN JUGA. HUH. Pengguna jalan raya juga RAKYAT btw.



- 15 Indonesian students/alumni from 5 Universities
- Labelled 100 Tweets each
- Using one of three NLP data annotation tasks:
  - Named Entity Recognition
  - Semantic Role Labeling
  - Binary Classification
- In one week, using KataKita on their mobile phones

### **Evaluation Criteria**



### Usability

I could use KataKita from mobile phone so I can annotate anytime and anywhere.

When I use KataKita, I need to wait couple of minutes until all the tokens were loaded on the screen.

KataKita annotation guideline is easy to understand.

I think KataKita is too complicated.

I think KataKIta is easy to use.

I think I need technical support to use KataKita

I imagine that most of KataKita users could learn to use KataKita quickly

I think KataKita is impractical to use

I feel very confident when doing the annotation using KataKita.

I must learn a lot of things before using KataKita.



### User Activity – Annotation Speed



**Time per Record - Density** 

#### **Median Time**

Binary Classification	5s
Named Entity Recognition	17s
Semantic Role Labelling	41s

### User Activity – Annotation Time per User



### **Experimental Evaluation**



### **Experimental Evaluation**



Task	Fleiss' Kappa	Interpretation	
Binary Classification	0.45	Moderate Agreement	
Named Entity Recognition	0.22	Low Agreement	
Semantic Role Labelling	0.41	Moderate Agreement	

0 = no agreement ,1 = perfect agreement

- How to improve annotation quality? What are the factors and user attributes influencing the quality?
- How to present guidelines and provide training on the phone?

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https://github.com/strikesensor/

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