# PAN@CLEF 2022 Style Change Detection Task

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# Given a document, find the **positions** at which the **authorship changes**.

## Style Change Detection Task

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# Looking back...

2017

Is the document multi-authored? If yes, find the positions of switches.

2018

Is the document written by a single or multiple authors?

2019

Is the document multi-authored? If yes, find the positions of switches. How many authors have collaborated on the document?

2020

Is the document multi-authored? For each pair of paragraphs, is there a style change between them?

2021

Is the document multi-authored? For each pair of paragraphs, is there a style change between them? Assign all paragraphs to an author id.

**[Task1] Style Change Basic:** for a text written by two authors that contains a single style change only, find the position of this change (i.e., cut the text into the two authors' texts on the paragraph-level),

[Task2] Style Change Advanced: for a text written by two or more authors, find all positions of writing style change (i.e., assign all paragraphs of the text uniquely to some author out of the number of authors assumed for the multi-author document)

[Task3] Style Change Real-World: for a text written by two or more authors, find all positions of writing style change, where style changes now not only occur between paragraphs, but at the sentence level.

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[Task3] Style Change Real-World: for a text written by two or more authors, find all positions of writing style change, where style changes now not only occur between paragraphs, but at the sentence level.

#### **Example Document A**

#### **Author 1**

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#### **Author 2**

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#### **Author 2**

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Task 1: [1,0]
Task 2: [1,2,2]

#### **Example Document B**

#### Author 1

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#### Author 2

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#### Author 2

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#### **Author 3**

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#### Task 2: [1,2,2,3]

#### **Example Document C**

#### **Author 1**

Author 2

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#### **Author 3**

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Task 3: [0,0,1,0,0,0,0,0,0,1,....]

### Dataset

- Realistic, non-artificial and comprehensive dataset
- Requirements
  - Find multiple authors that write about the same topic
  - Multi-authored texts need to contain the same topic / subtopic
  - Find texts that are freely available and of sufficient length
- StackExchange consists of several sites (topics), data freely available
- Each question/answer associated with
  - Site
  - Set of tags (subtopics)
- Example: photography site subtopics: lens, canon, nikon, lightroom, photoshop, ...





As the post suggests it can be called the

34

#### flesh



NOUN



1.2 The edible pulpy part of a fruit or vegetable.

halve the avocados and scrape out the flesh

Calling it 'pulpy' doesn't mean it has been squashed. It means the soft part that isn't the skin/rind or pips/stone.

You don't have to call it 'fruit flesh', just 'flesh' will do when the fruit has already been mentioned.

Hey I wanted to make some smoothie with this avocado but I found it over-ripe and part of its **flesh** has changed color.

From Lexico.

### Dataset

- Cleaning
  - Remove links, images, code snippets, bullet lists, block quotes
  - Remove very short questions/answers
  - Remove edited questions/answers
- Randomly choose question/answer thread
- Randomly choose number of authors n in [1,5]
- Randomly select n authors
- Take their paragraphs in thread, shuffle, concatenate
- Check length of document (> 2 paragraphs, between 1,000 and 10,000 characters)

### Dataset

Parameter	Setting
Number of authors	1-5
Document length	1,000-10,000 characters
Minimum paragraph length	100 characters
Minimum number of paragraphs	2
Change positions	Between paragraphs (tasks 1 and 2) Between sentences (task 3)
Document language	English

Task 1: 2,000 documents

Tasks 2 and 3: 10,000 documents

70:15:15 training/validation/test split

### Evaluation

- Score each document by  $F_1$  measure (macro-averaged)
- Three tasks evaluated independently
- Additionally for tasks 2 and 3:
  - Diarization Error Rate (DER): fraction of wrongly attributed paragraphs
  - Jaccard Error Rate (JER): author-specific error rate

$$JER(a) = 1.0 - \frac{|ref_a \cap pred_a|}{|ref_a \cup pred_a|}$$

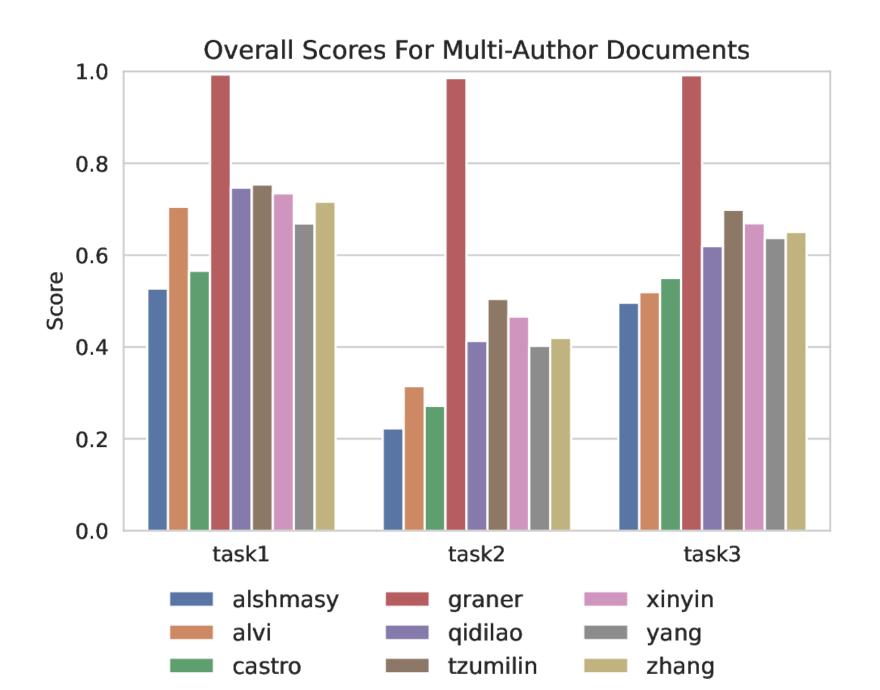
### Submissions

- Nine submissions (eight intrinsic, one extrinsic)
- Alshamasi and Menai (alshmasy): lexical and syntactic features, k-means clustering
- Alvi et al. (alvi): handcrafted discourse markers as features
- Jiang et al. (xinyin): Electra (masked LM), one model per task
- Lao et al. (qidilao): BERT, convolution layer, max pooling for binary class prediction
- Lin et al. (tzumilin): majority-vote ensemble of three LLMs (BERT, RoBERTa, ALBERT)
- Rodriguez-Losada and Castro (castro): transformer and lexical features, similarity for final prediction
- Zhang et al. (zhang): BERT (masked training) and prompt modeling for final prediction
- Zi et al. (yang): BERT (masked training), Bi-LSTM to capture context, fully connected layer for predictions
- Graner and Ranly (graner): extrinsic approach

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# Results

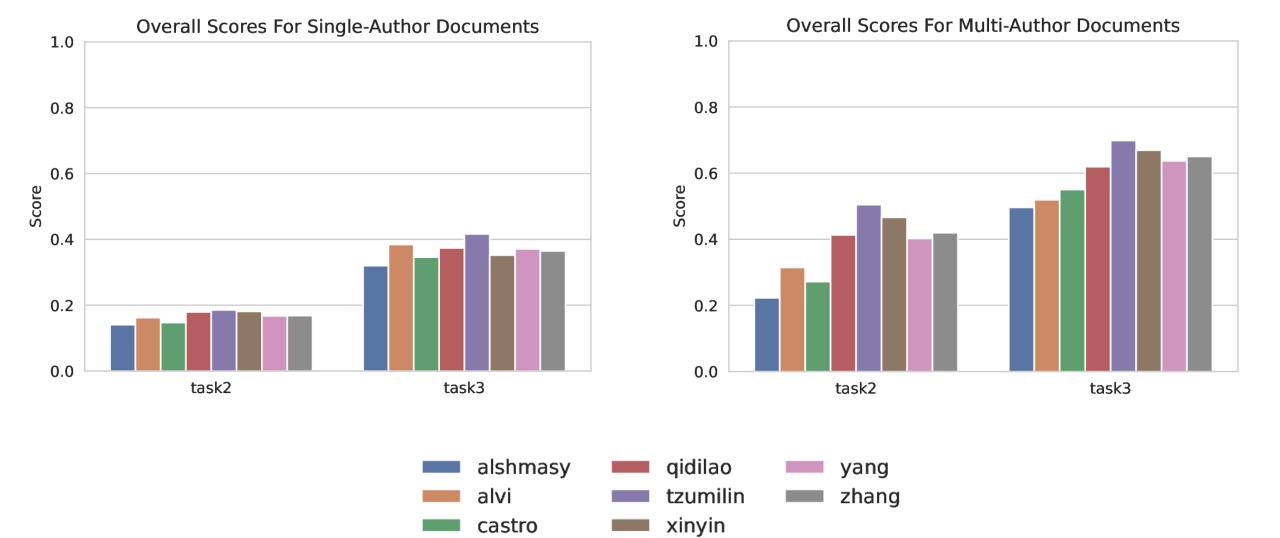


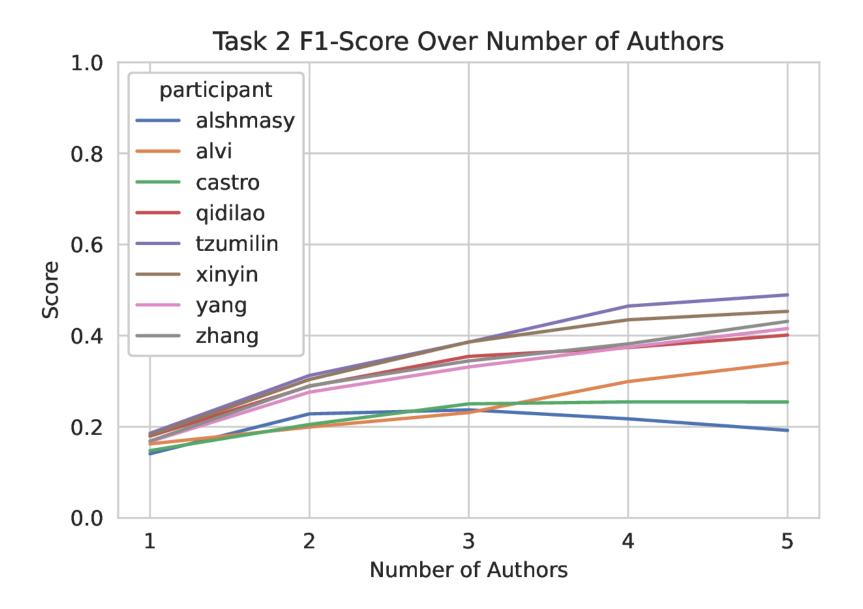
## Overview

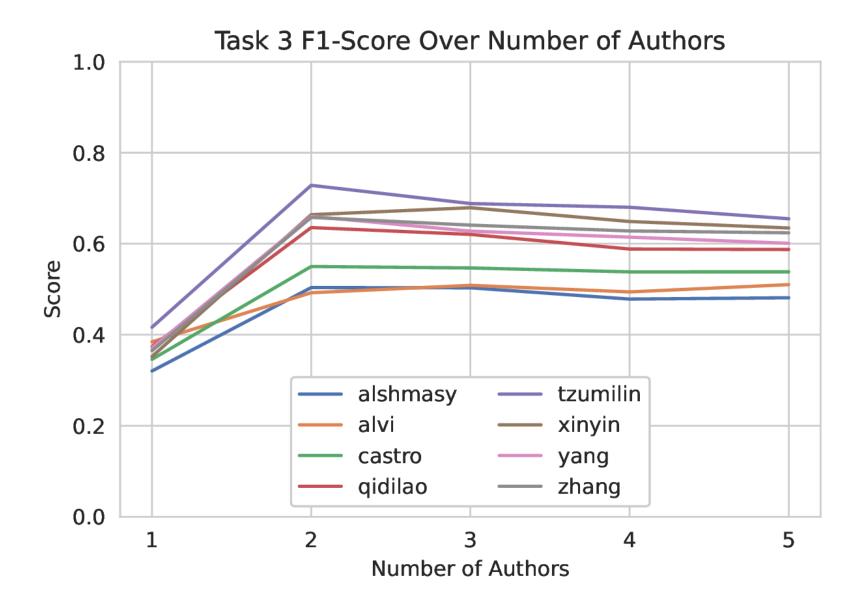
Participant	ST1 F <sub>1</sub>	ST2 F <sub>1</sub>	<b>ST3 F</b> <sub>1</sub>	ST3 DER	ST3 JER
	Ir	ntrinsic Approa	ches		
tzumilin22	0.7540	0.5100	0.7156	0.1941	0.3095
xinyin22	0.7346	0.4687	0.6720	0.2380	0.3138
qidilao22	0.7471	0.4170	0.6314	0.2636	0.3641
zhang22	0.7162	0.4174	0.6581	0.2886	0.3556
yang22	0.6690	0.4011	0.6483	0.2964	0.3677
alvi22	0.7052	0.3213	0.5636	0.3924	0.5218
castro22a	0.5661	0.2735	0.5565	0.4035	0.5771
alshmasy22	0.5272	0.2207	0.4995	0.4240	0.6444
	Ex	xtrinsic Approa	ches		
graner22	0.9932	0.9855	0.9929	0.0040	0.0040
		Baseline			
Random	0.3222	0.2651	0.4809	0.4568	0.5938

## Overview

	Participant	$ST1 F_1$	$ST2 F_1$	$ST3 F_1$	ST3 DER	ST3 JER			
	Intrinsic Approaches								
	tzumilin22	0.7540	0.5100	0.7156	0.1941	0.3095			
_	xinyin22	0.7346	0.4687	0.6720	0.2380	0.3138			
ILM	qidilao22	0.7471	0.4170	0.6314	0.2636	0.3641			
_	zhang22	0.7162	0.4174	0.6581	0.2886	0.3556			
	yang22	0.6690	0.4011	0.6483	0.2964	0.3677			
	alvi22	0.7052	0.3213	0.5636	0.3924	0.5218			
	castro22a	0.5661	0.2735	0.5565	0.4035	0.5771			
	alshmasy22	0.5272	0.2207	0.4995	0.4240	0.6444			
	Extrinsic Approaches								
	graner22	0.9932	0.9855	0.9929	0.0040	0.0040			
	Baseline								
	Random	0.3222	0.2651	0.4809	0.4568	0.5938			







- Style change detection often benefited from topic changes within documents
- Control for change of authorship and topic simultaneously
  - > control for topic and style similarity in assembling documents
  - → Uncover ability to capture/model style variations

- New dataset based on Reddit
- New "perspective" on task complexity: single task carried out on three datasets of increasing task complexity

Core task: Detect change of authorship on the paragraph level

- 1. Easy dataset: paragraphs cover a variety of topics
- 2. Medium dataset: topical variety is limited, focus on style information
- 3. Hard dataset: all paragraphs cover the same topic, focus only on style information

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Any feedback welcome ©

