

### Supported by Open Source





- Abstract extraction and and matching
- General data processing
- BSD-Zero Clause



# Selenium

- EGU Session/Topic extraction
- Apache 2.0



## • BERT/ HuggingFace **DistilBERT**

- Transformer model for prediction
- Apache 2.0



- Analysis and prototyping
- BSD-3 Clause









### Summary



- Large corpus (2004 to 2023) of permissively licenced Geoscience Abstracts available (EGU Conference)
- Abstracts are readily available but classifying the abstracts to topics using EGU session information required considerable effort
- With topics assigned, supervised learning could be used to build an ML model that can assign the most relevant geoscience topic of a geoscience abstract
- ML model is significantly better than random model in classifying abstracts
- Accuracy of abstract to topic matching is generally successful
- Training data will be released publicly and codebase will be made open source









### Data Modelling



### **Data Collection**

### **Abstracts Extraction**

- PDF to Text and XML source files
   Topics Extraction
- EGU Websites with Selenium Web Driver

### Abstracts to Topics Matching

Matched on EGU identifiers

### Data Modelling

### Training

- Supervised learning
- BERT and DistilBERT

### Evaluation

 Accuracy of single true label vs multiple true label Geoscience Abstract Topic Classification Model











1. Abstract Extraction

2. Topic Extraction

3. Abstract To Topic Matching

4. Training and Evaluation

5. Model Accuracy
BERT vs
DistilBERT?

6. Potential Applications











### Data Collection – Abstracts extraction – Stats



Year	Original Data	<b>Converted Format</b>
2015 - 2023	XML	JSON
2000 - 2014	PDFs	JSON

Year	Num Abstracts	Topics	
2015	14K	N/A	
2014	15K	N/A	
2013	13K	N/A	
2004	7K	N/A	

### Unavailable **Topics**















### Data Collection – PDF Abstracts extraction





Geophysical Research Abstracts Vol. 17, EGU2015-150, 2015 EGU General Assembly 2015

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Root and microbial respiration from urban, agricultural and natural soils within the Moscow megapolis

Authors • Affiliations •

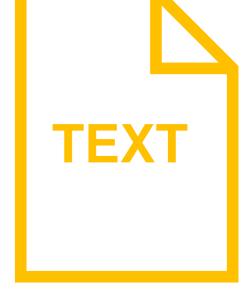
Viacheslav Vasenev (1,2), Simona Castaldi (3), Marya Vizirskaya (1), Nadezhda Ananyeva (4), Kristina Ivashchenko (4), Riccardo Valentini (), and Ivan Vasenev ()

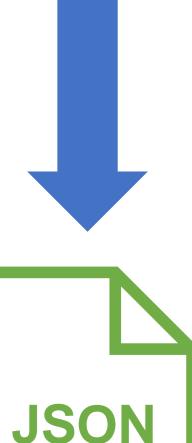
(1) Russian State Agrarian University, Moscow, Russian Federation (vasenyov@mail.ru), (2) Peoples' Friendship University of Russia, (3) Seconda Università di Napoli, (4) Institute of Physico-chemical and Biological Problems in Soil Science

**Abstract** 

Urbanization is an important process of land-use change, which is increasing with the growth of population and abandonment of rural areas. Urbanization alters profoundly soil features and functions, among which soil respiration, which is one of the main carbon fluxes to the atmosphere. Soil respiration is the result of heterotrophic and autotrophic components, which are driven by biotic and abiotic factors. Little is known about soil respiration and its components in urban environments, which represent highly variable systems, characterized by different functional zones, types and intensities of urban management. In the present study we analyzed the spatial variability and temporal dynamics of total soil respiration (Rs) and its components, autotrophic (Ra) and heterotrophic respiration (Rh), from soils of different environments included in the Moscow megalopolis area. In particular we compared highly impacted areas urban green lawns with less anthropized ecosystems within the Moscow city: arable lands and urban forest sites. Experiments were set after snow melt and respiration fluxes were analyzed during the whole summer period till the beginning of the autumn. Data showed that Rs was significantly higher in the most disturbed sites, the green lawns, and showed the highest variability among the three analyzed land use types. Rh was the dominant component of soil respiration in all sites and did not vary significantly during the study period. However, significant differences was shown for the metabolic quotient qCO2, estimated as heterotrophic respiration ratio to microbial carbon (Rh/Cmic). The most disturbed sites showed the highest qCO2 within the lawn land use, followed by arable sites and forest sites, characterized by the lowest qCO2. Ra contributed to total Rs only at a minor extent (26%) and increased in all study sites along the season following the phenological cycle of the plant communities. Ra absolute values and relative contribution to Rs did not change significantly among land use types. Overall, the high observed fluxes of CO<sub>2</sub> in urban lawns seemed to be driven by land management and disturbance impact on the microbial community





















### Data Collection – XML Abstracts extraction





















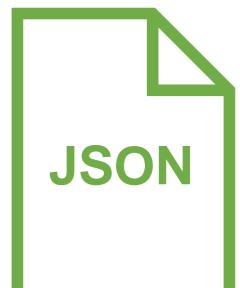
### Web Scraping with Selenium





EGU Conference Websites





Year	Num Topics	EGU Conference Links
2015	22	https://meetingorganizer.copernicus.org/egu2015/sessionprogramme
2014	23	https://meetingorganizer.copernicus.org/egu2014/sessionprogramme
2013	23	https://meetingorganizer.copernicus.org/egu2013/sessionprogramme

# Available Topics

- 2015: exclude IS
- 2013 2014: exclude IS and PSD











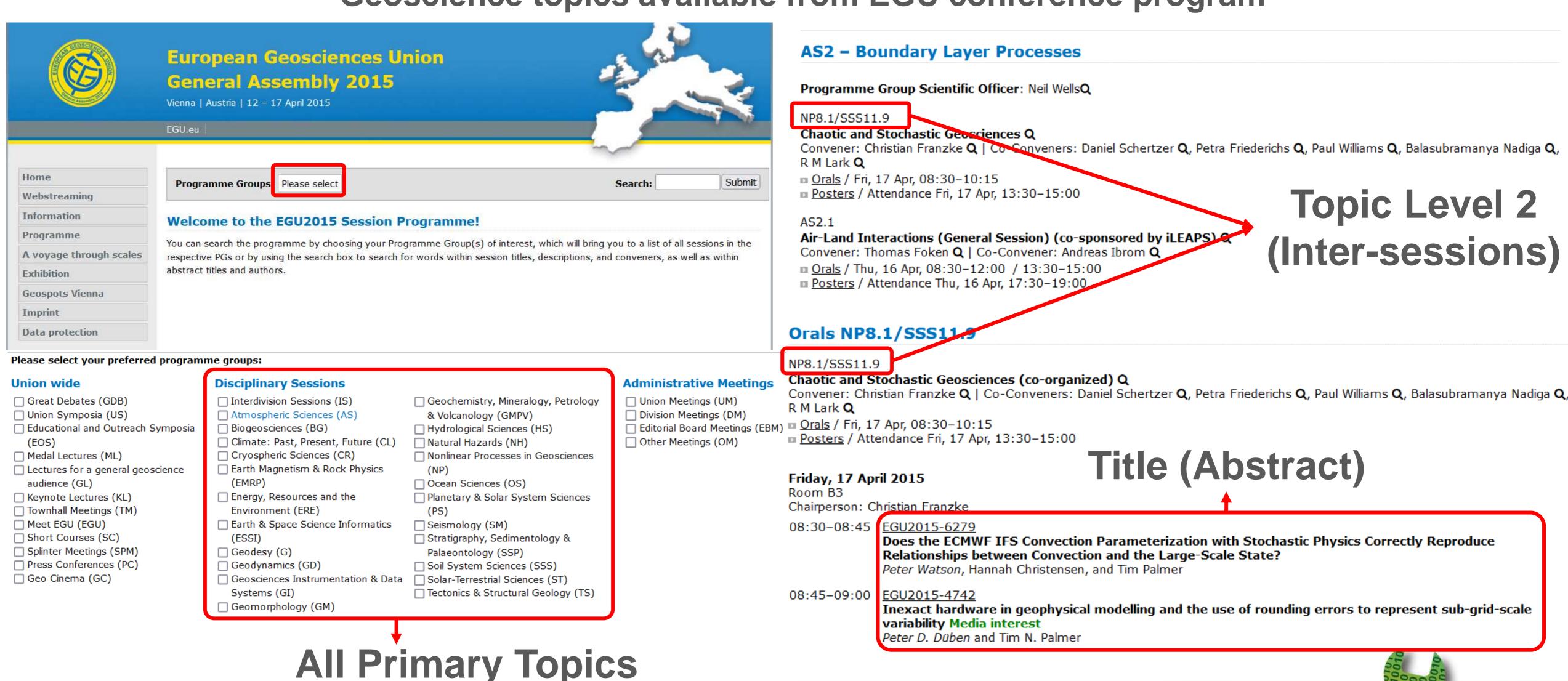




### Data Collection – Topics extraction - Example



### Geoscience topics available from EGU conference program











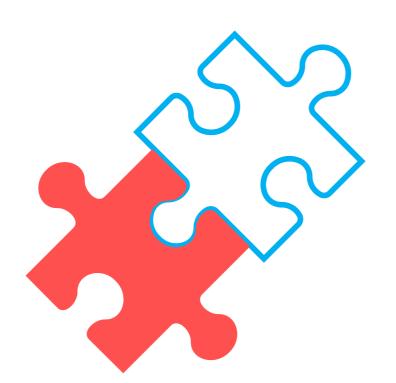






### Data Collection – Abstract/Topics Matching





ID: EGU-\*\*\*\*

Year	Abstracts	Primary Topics
2015	14K	22
2014	15K	23
2013	13K	23

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	ID	Geoscience Topics 2015		
	AS	Atmospheric Sciences		
	BG	Biogeosciences		
	CL	Climate: Past, Present, Future		
	CR	Cryospheric Sciences		
	EMRP	Earth Magnetism & Rock Physics		
	ERE	Energy, Resources and the Environment		
	ESSI	Earth & Space Science Informatics		
	G	Geodesy		
	GD	Geodynamics		
	GI	Geosciences Instrumentation & Data Systems		
	GM Geomorphology			
	GMPV	Geochemistry, Mineralogy, Petrology & Volcanology		
	HS	Hydrological Sciences		
	NH	Natural Hazards		
	NP	Nonlinear Processes in Geosciences		
	OS	Ocean Sciences		
	PS	Planetary & Solar System Sciences		
	SM	Seismology		
	SSP	Stratigraphy, Sedimentology & Palaeontology		
	SSS	Soil System Sciences		
	ST	Solar-Terrestrial Sciences		
	TS	Tectonics & Structural Geology		
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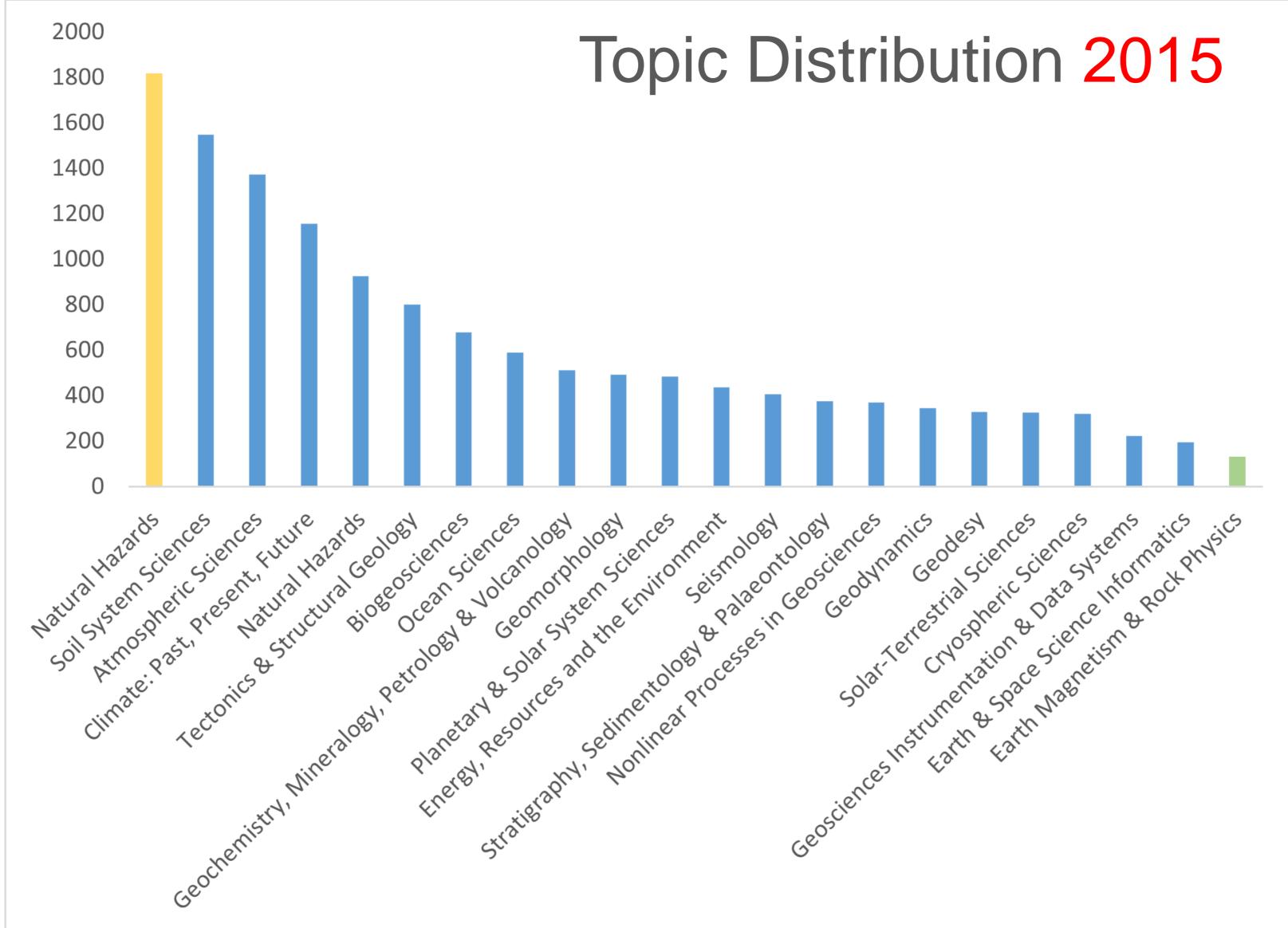








### Data Preprocessing – Datasets Stats



ID	Geoscience Topics	Sharing is encouraged
	<u> </u>	
AS	Atmospheric Sciences	
BG	Biogeosciences	
CL	Climate: Past, Present, Future	
CR	Cryospheric Sciences	
EMRP	Earth Magnetism & Rock Physics	
ERE	Energy, Resources and the Enviro	nment
ESSI	Earth & Space Science Informatic	S
G	Geodesy	
GD	Geodynamics	
GI	Geosciences Instrumentation & Da Systems	ata
GM	Geomorphology	
GMP V	Geochemistry, Mineralogy, Petrolo Volcanology	ogy &
HS	Hydrological Sciences	
NH	Natural Hazards	
NP	Nonlinear Processes in Geosciences	
OS	Ocean Sciences	
PS	Planetary & Solar System Sciences	
SM	Seismology	
SSP	Stratigraphy, Sedimentology & Palaeontology	
SSS	Soil System Sciences	
ST	Solar-Terrestrial Sciences	
TS	Tectonics & Structural Geology	















### Model Training





















### **Evaluation Results**



### For Single True Label:

- Exact match is required between predicted label and the FIRST listed session e.g. NP/SSS
- Only 'NP' is considered an accurate classification

### For Multiple True Label:

- Exact match of predicted label and ANY listed session e.g. NP/SSS
- Either 'NP' or 'SSS' is an accurate classification

This recognises that the first session listed may not always be the most relevant

















**Data Science** 

**Transforming** 

Maintenance

### Classification Performance as percentage (%)

Year	Model	F1 (single true label)	Accuracy (single true label)	Accuracy (multiple true label)
2015	DistilBERT [1]	65.4	65.8	73.9
	BERT [2]	66.7	66.9	75.2

[1] Sanh et al. (2019). DistilBERT, a distilled version of BERT: smaller, faster, cheaper and lighter. <a href="https://arxiv.org/pdf/1910.01108.pdf">https://arxiv.org/pdf/1910.01108.pdf</a>
[2] Devlin et al. (2018). BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding. <a href="https://arxiv.org/pdf/1810.04805.pdf">https://arxiv.org/pdf/1810.04805.pdf</a>







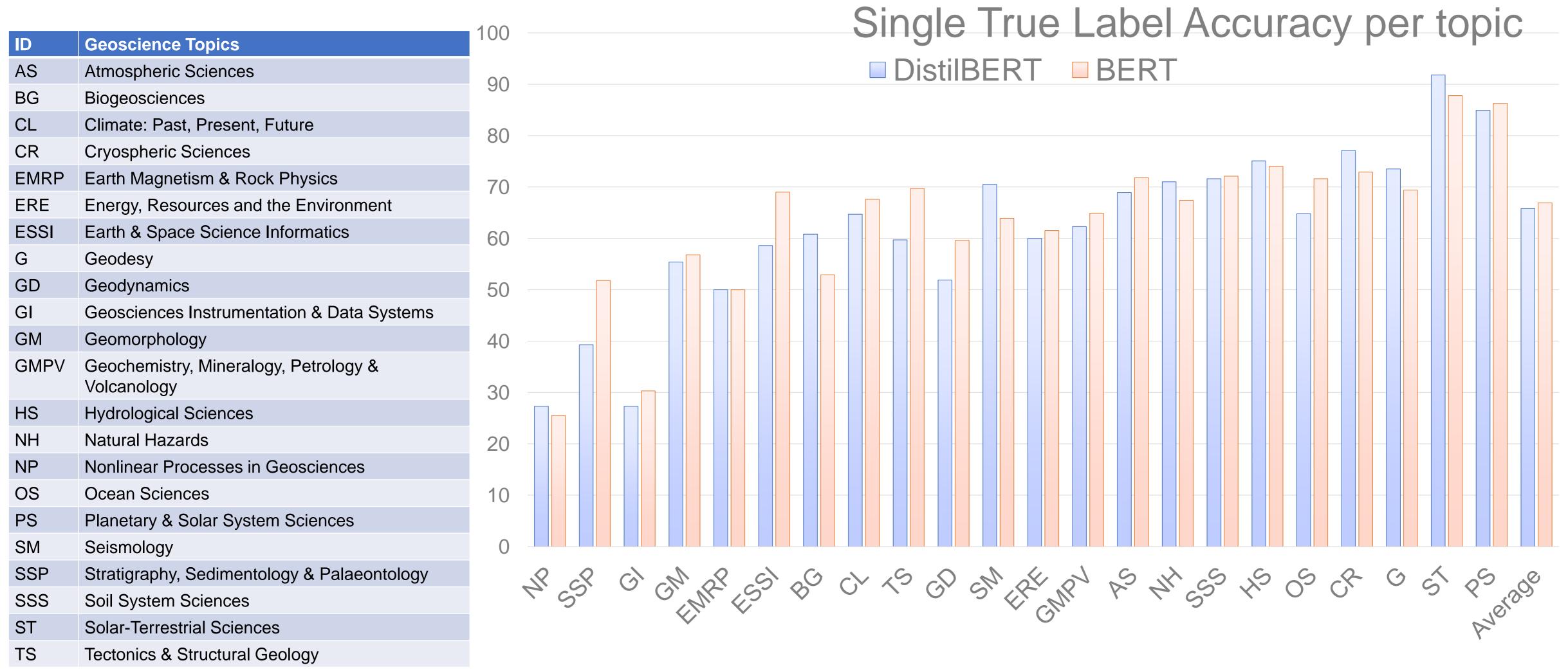






### Single True Label – Accuracy per Topic





[1] Sanh et al. (2019). DistilBERT, a distilled version of BERT: smaller, faster, cheaper and lighter. <a href="https://arxiv.org/pdf/1910.01108.pdf">https://arxiv.org/pdf/1910.01108.pdf</a>
[2] Devlin et al. (2018). BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding. <a href="https://arxiv.org/pdf/1810.04805.pdf">https://arxiv.org/pdf/1810.04805.pdf</a>









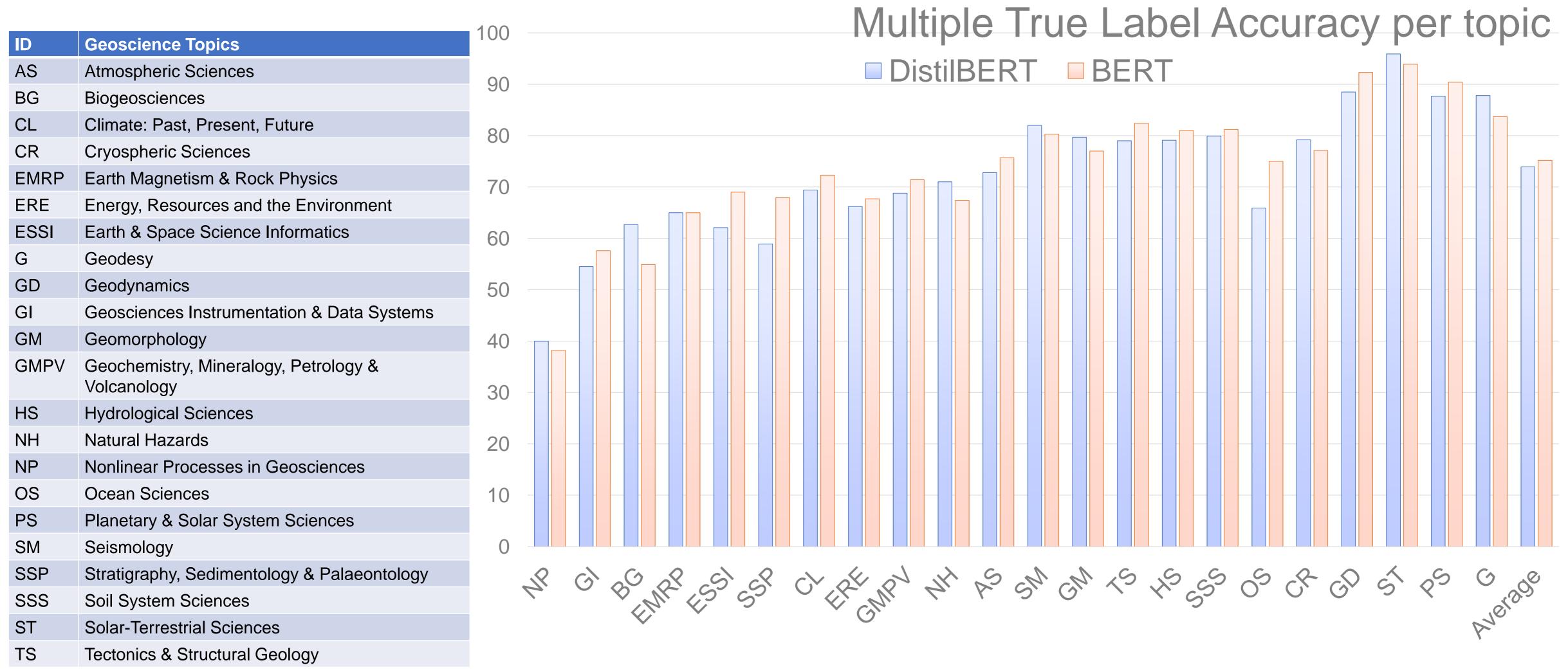






### Multiple True Label Accuracy per Topic





[1] Sanh et al. (2019). DistilBERT, a distilled version of BERT: smaller, faster, cheaper and lighter. <a href="https://arxiv.org/pdf/1910.01108.pdf">https://arxiv.org/pdf/1910.01108.pdf</a>
[2] Devlin et al. (2018). BERT: Pre-training of Deep Bidirectional Transformers for Language Understanding. <a href="https://arxiv.org/pdf/1810.04805.pdf">https://arxiv.org/pdf/1810.04805.pdf</a>















### Potential Applications



### Trained ML Model:

- Session recommender system for attendees
- Identifying abstracts for potential cross listing
- Assisted "binning" of abstracts into relevant sessions
- Vetting of abstracts to ensure they are relevant to sessions
- Abstract reviewer system
- **Topic Distribution Analysis**
- Cross listing impacts and benefits

### Public Dataset Release:

- LLM Training
- Named Entity Recognition
- Knowledge Graph Extraction
- What are your ideas?

















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