MEET EIFFEL: ROLE, CHALLENGES, EXPECTATIONS

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OUTLINE


EIFFEL Overview
LIBRA's Role
Technical/NLP Introduction
LIBRA'S Technical Tasks
Challenges
Impact


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PROJECT INTRODUCTION

- **Title**: Revealing the role of **GEOSS** as the default digital portal for building ClimateChange adaptation & mitigation applications
- **Duration**: 36 months
- **Participants**: 19 partners; 8 countries

Eiffel

GEOSS APPLICATIONS FOR CLIMATE CHANGE

GEOSS

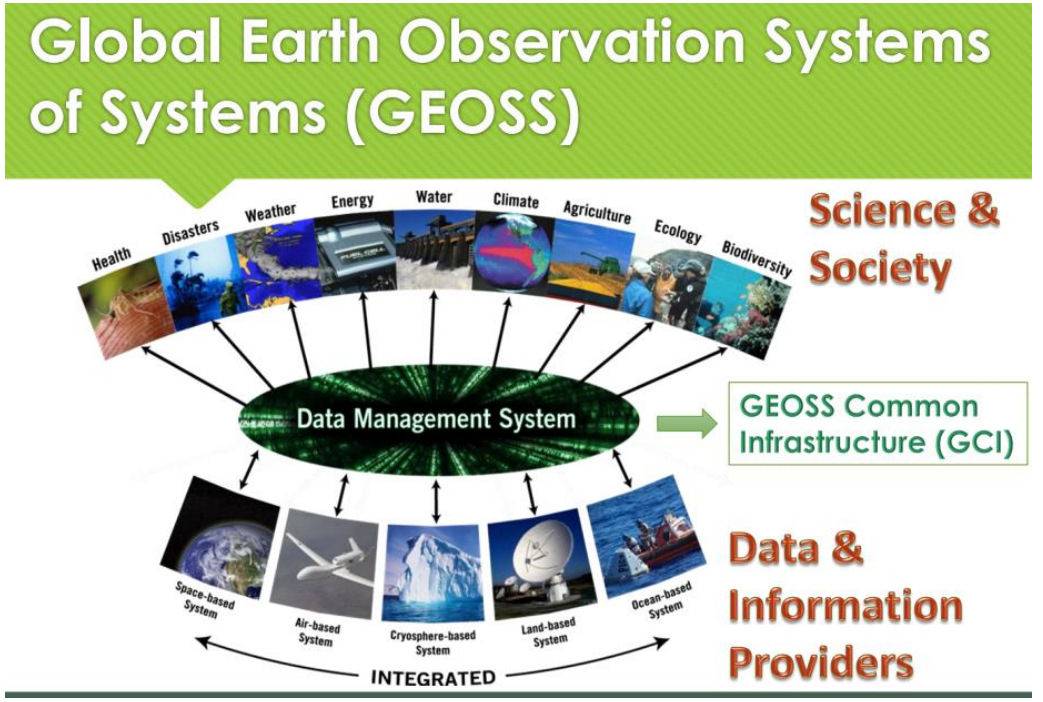
- What is GEOSS?

Global Earth Observation System of Systems is a set of coordinated, independent Earth observation systems.

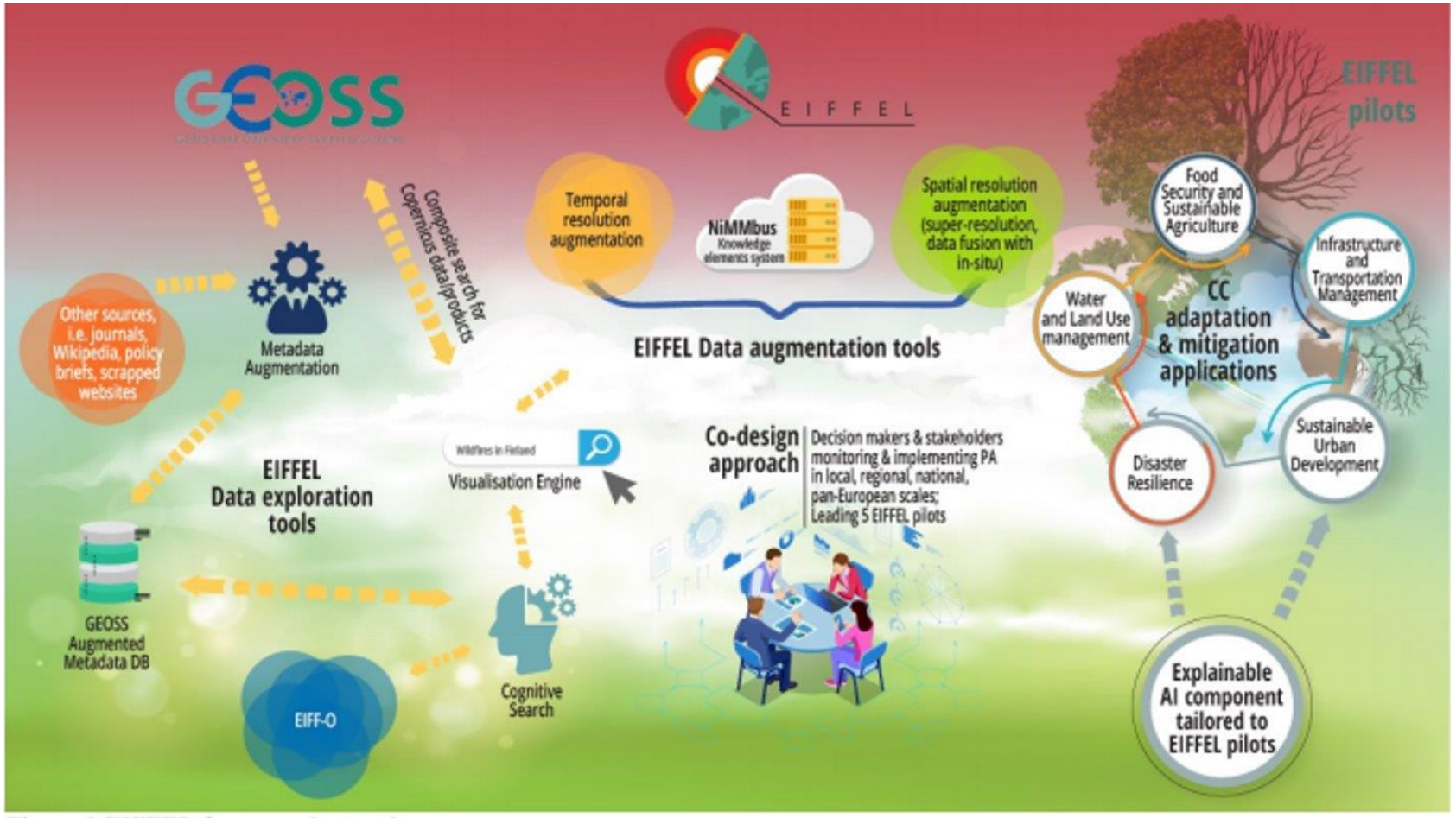
- What is Earth Observation?

The gathering of information about planet Earth's physical, chemical and biological systems.

- Why is Earth Observation important?
  - is used to monitor and assess the status of, and changes in, the natural and manmade environment
  - **invaluable for assessing and mitigating the negative impacts of CC**
  - use for exploiting new opportunities, such as the sustainable management of natural resources



PROJECT OVERVIEW



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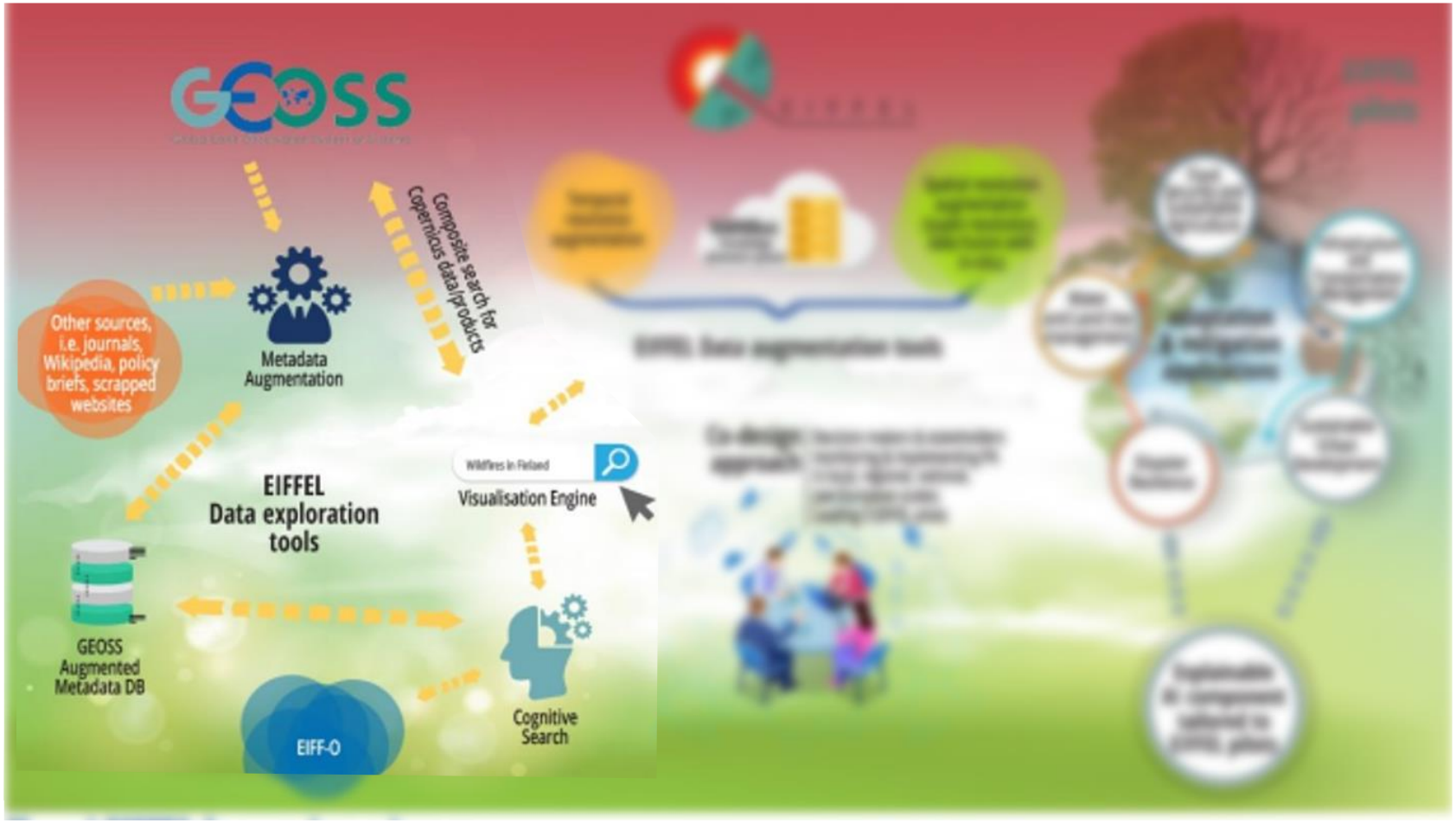

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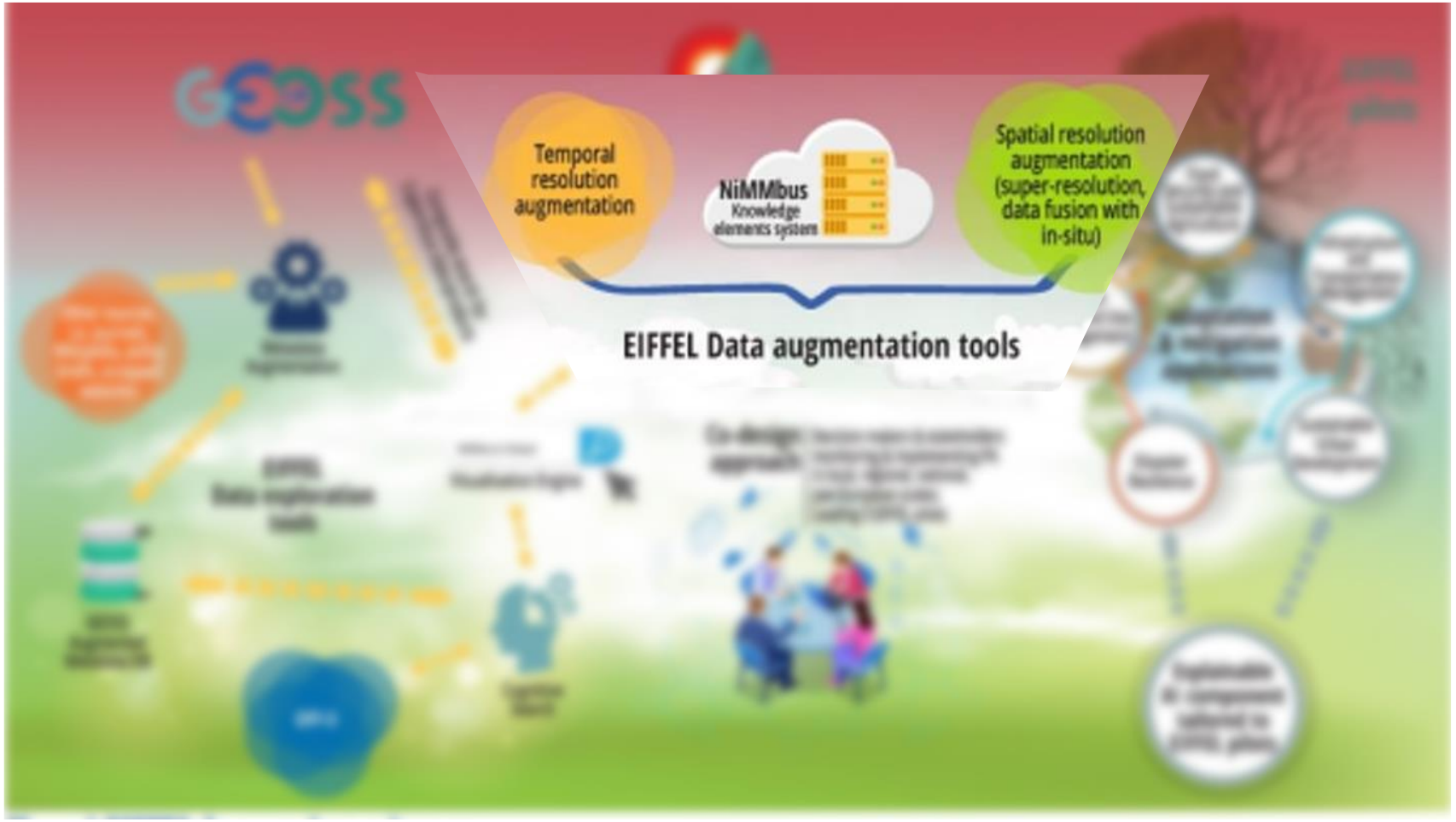

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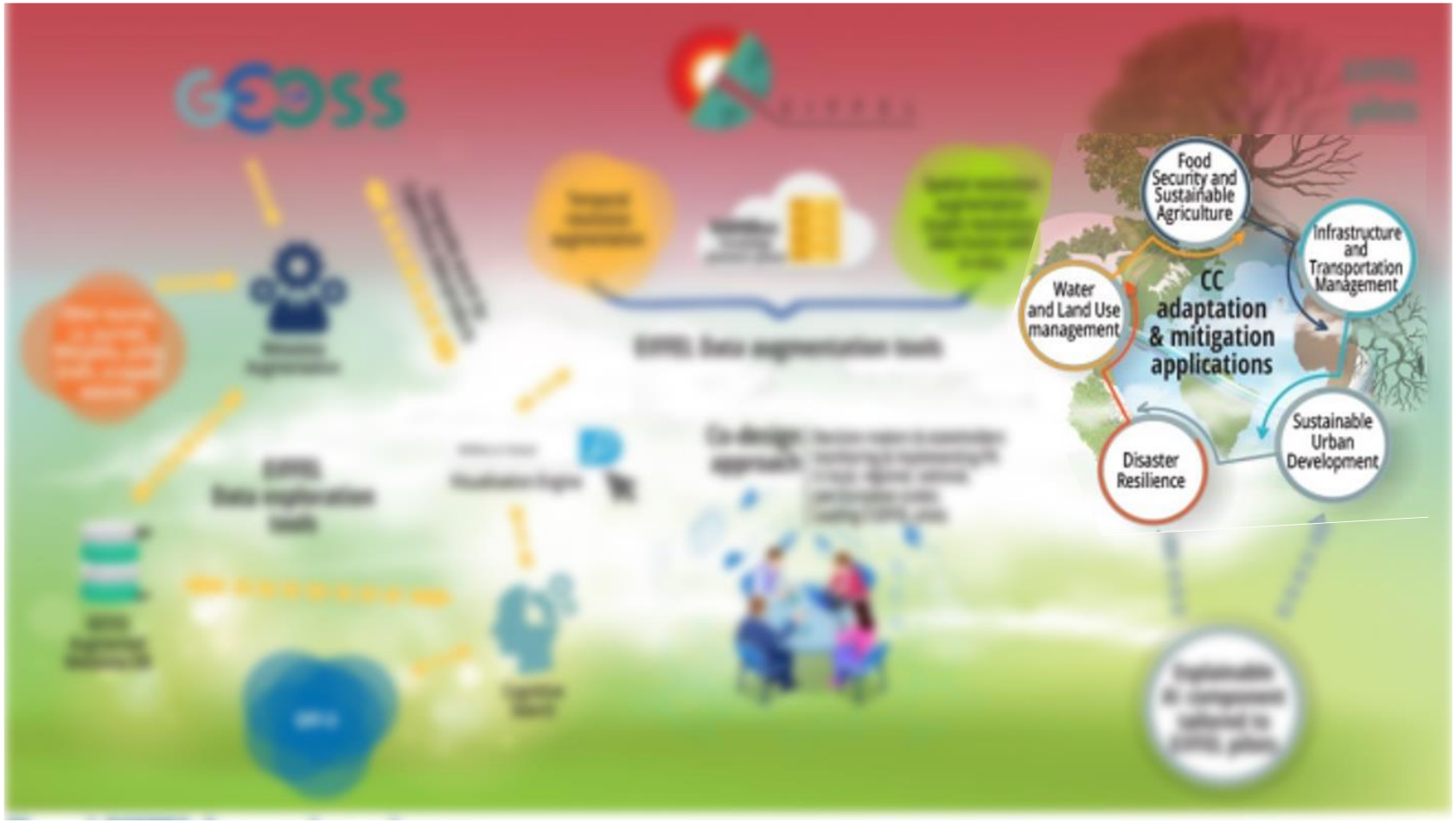

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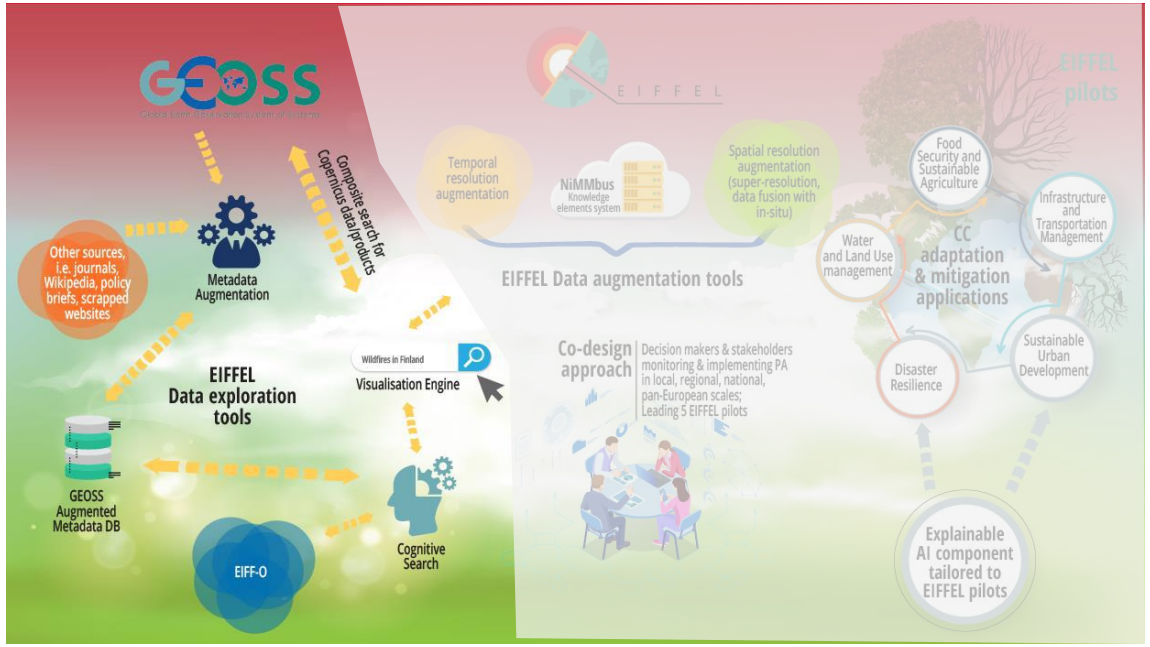
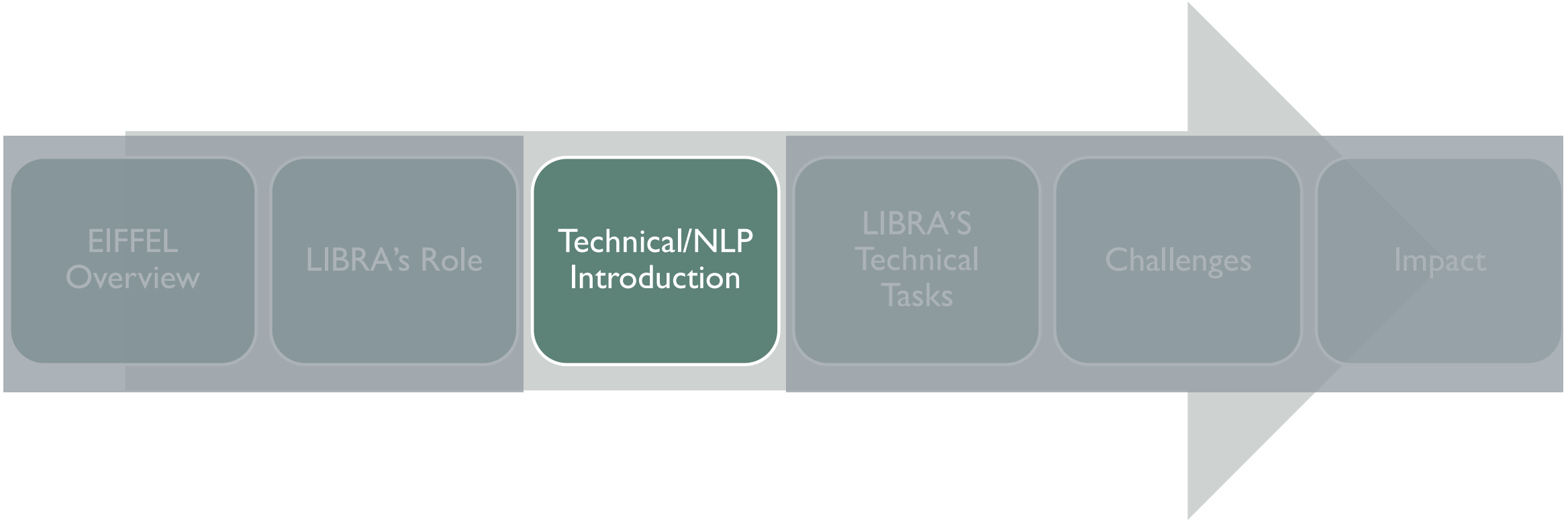
LIBRA'S ROLE



- Leading the Augmenting GEOSS data exploration activities of the project, including:
  - NLP-based cognitive search engine
  - Visualisation engine of the EIFFEL cognitive search tool
  - Metadata curation and augmentation
  - EIFFEL CC-focused ontology
- Development of the cognitive search framework
- Design and deployment of metadata enrichment mechanisms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- What is NLP?

  Machine learning applied to text / speech

- Computers only understand numbers, not characters, words, or sentences -> **text representation**

TRADITIONAL CONTEXT-FREE REPRESENTATIONS

- **Bag of Words**

  each element in the vector corresponds to a unique word in the vocabulary

  if word exists in document, element is marked as 1, else as 0

- **TF-IDF**

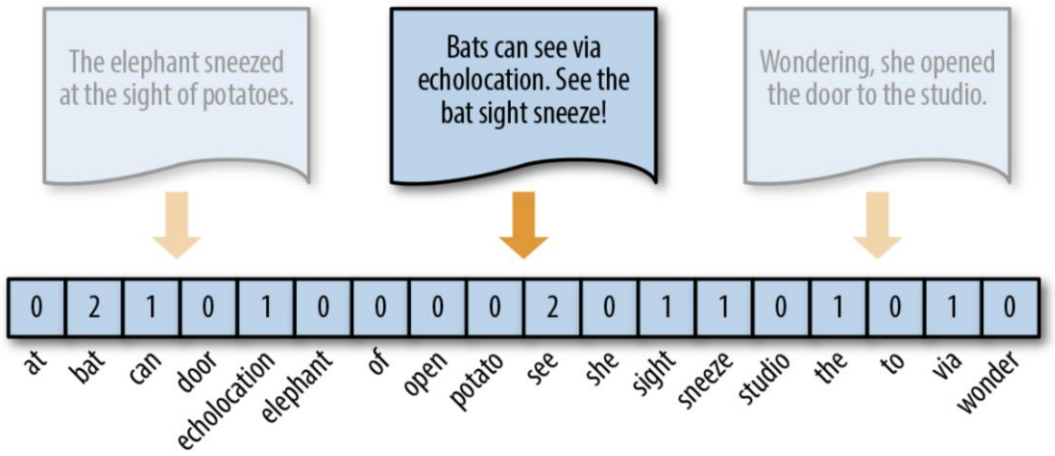
  TF: scoring of the frequency of the word in the current document

  IDF: scoring of how rare the word is across documents

  TF-IDF score = TF * IDF

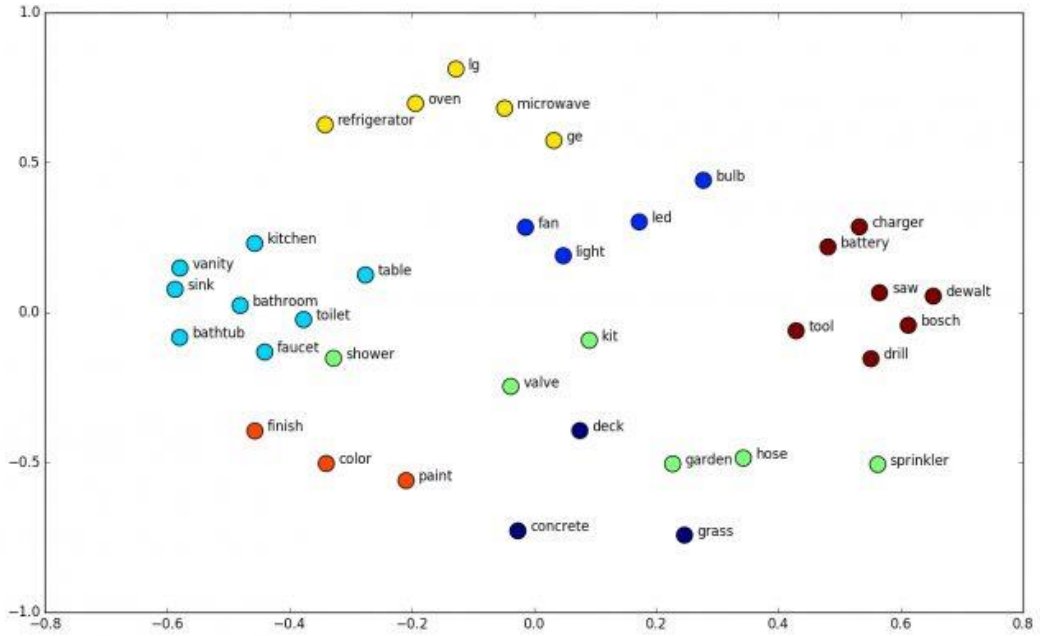
  Contains information on the more/less important words

  UNABLE to capture word meaning/word similarity!



WORD EMBEDDINGS – DISTRIBUTIONAL SIMILARITY BASED REPRESENTATIONS

- Numeric vector representations of a particular word, that encodes the meaning of the word

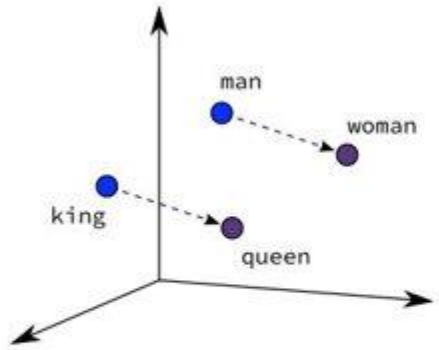


Distributional Hypothesis

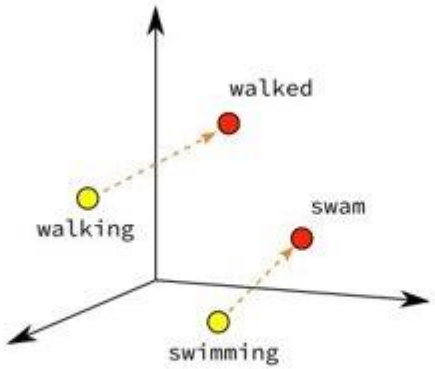
Words that occur
in similar contexts
tend to have
similar meanings

WORD ANALOGIES

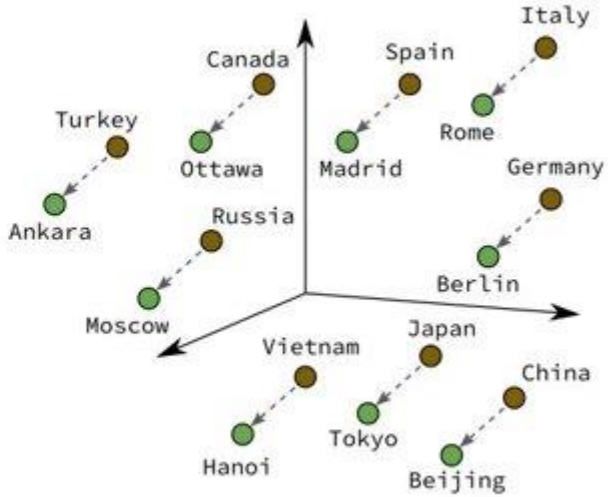
- A fascinating property of trained word embeddings is that the relationship between words is captured through linear relationships between vectors



Male-Female



Verb Tense



Country-Capital

king – man + woman = queen

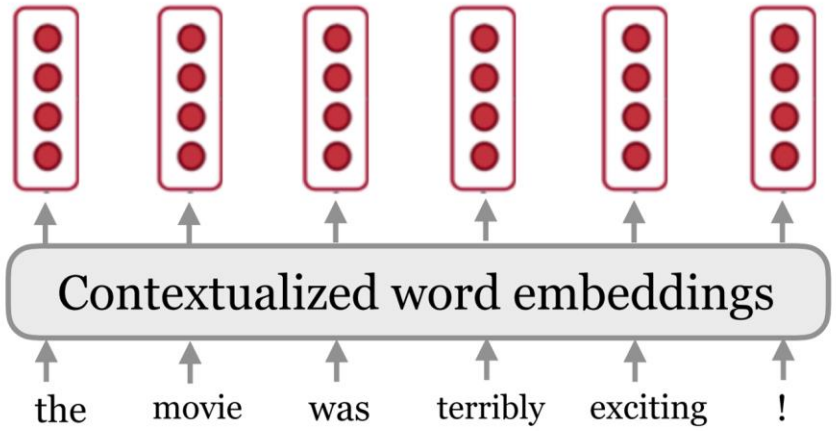
LIMITATIONS OF WORD EMBEDDINGS

- One vector for each word type (static)
- Polysemous words, eg. Bank, mouse
- Words don't appear in isolation. The word use depends on its context.
- Why not learn the representations for each word in its context?

" I like apples" VS "I like Apple macbooks"

CONTEXTUALIZED WORD EMBEDDINGS

- Address the issue of polysemous and the context-dependent nature of words
- Build a vector for each word conditioned on its context!



$$g : (w_1, w_2, \ldots, w_n) \longrightarrow \mathbf{x}_1, \ldots, \mathbf{x}_n \in \mathbb{R}^d$$

- Contextual representations of language leverage the intuition that the meaning of a particular word in a particular text depends not only on the identity of word, but also on the words that surround it at that moment

- ELMO, GPT-3 and BERT
- state-of-the-art performance on many popular tasks

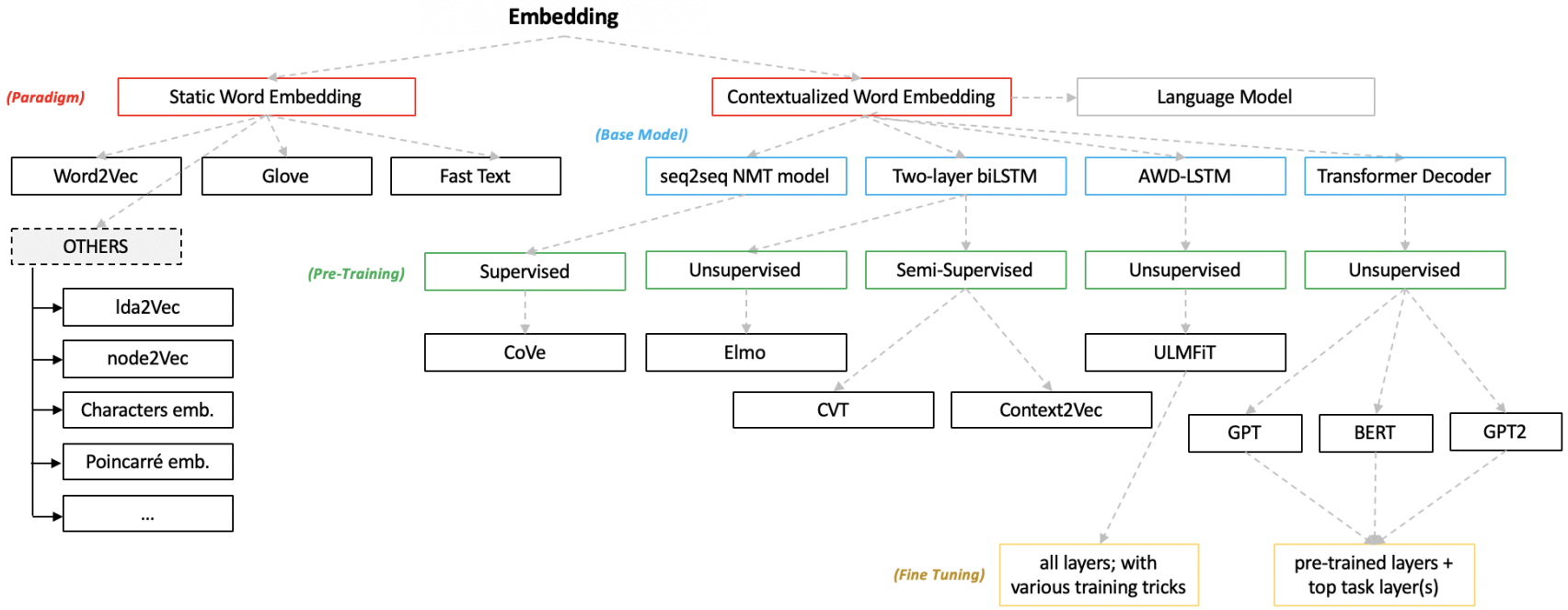
Take into account the word context

Trained on a large text corpus

Open-source

- Models pre-trained on language modeling (**unsupervised task**) and fine-tuned (supervised) with labeled data specific to a task

WORD EMBEDDING TECHNIQUES



More specific technical details… at following workshops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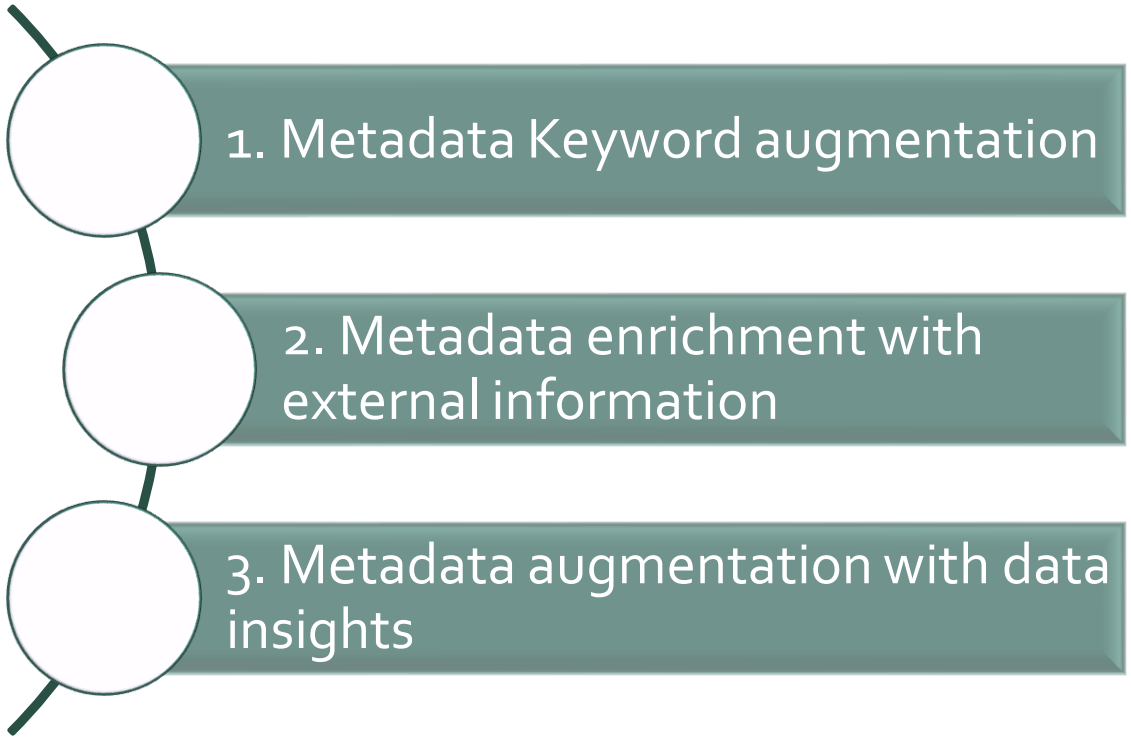
TASK 1: NATURAL LANGUAGE PROCESSING (NLP)-BASED COGNITIVE SEARCH FOR GEOSS DATASETS

- Build the core of a search engine for GEOSS data based on AI-powered Natural Language Processing (NLP)
- The search engine will allow advanced query-related capabilities that add **semantic relevance** to search results & allows searching based on free text
- Domain specific
- Contextually rank the most relevant search results

Google for GEOSS data

TASK 2: METADATA ENRICHMENT MECHANISMS

- **Goal**: Diverse toolset for automatic metadata curation and augmentation, specialised to the GEOSS context
- **New Metadata DB** will be developed – not physically linked with GEOSS

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CHALLENGES FOR LIBRA

- **Very large and unstructured corpus** of geo-science documents.
- Current metadata not sufficient / lack a keyword list.
- GPT-3 produced the equivalent **of 552 metric tons of carbon dioxide** during its training. That's the same amount that would be produced by driving 120 passenger cars for a year.
- CC Applications depend on the quality of augmenting GEOSS data exploration activities.
- Leading role of WP: Coordination of many teams.


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IMPACT ON LIBRA

- Deep dive into various NLP tasks:
  - Context-aware Word and document embeddings.
  - Question-Answering.
  - Document Ranking.
  - Information Retrieval.
- Get accustomed with SotA models, such as Deep Bidirectional transformers, Generative pre-trained transformers.
- Make Libra's work known to the expert communities.

Thank you!

Questions?