

Wenwen Li and Anna Liljedahl

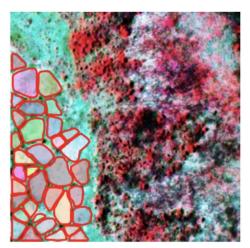




# Al's two big subfields

#### Image analysis/Computer vision

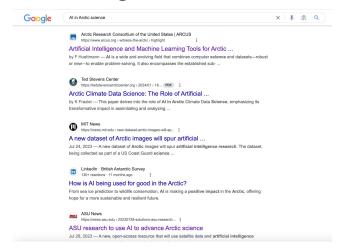
- Image classification
- Object detection and recognition
- Facial recognition
- Medical imaging



Segmenting IWP from Maxar imagery

#### **Natural language processing**

- Chatbots and virtual assistants
- Text and sentiment analysis (search)
- Language translation
- Content generation

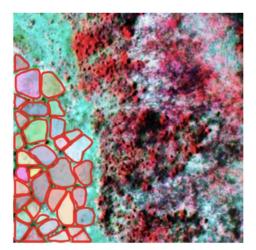


Google search for "AI in Arctic science"

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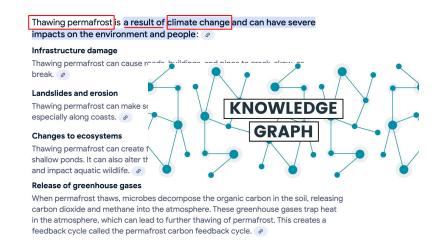
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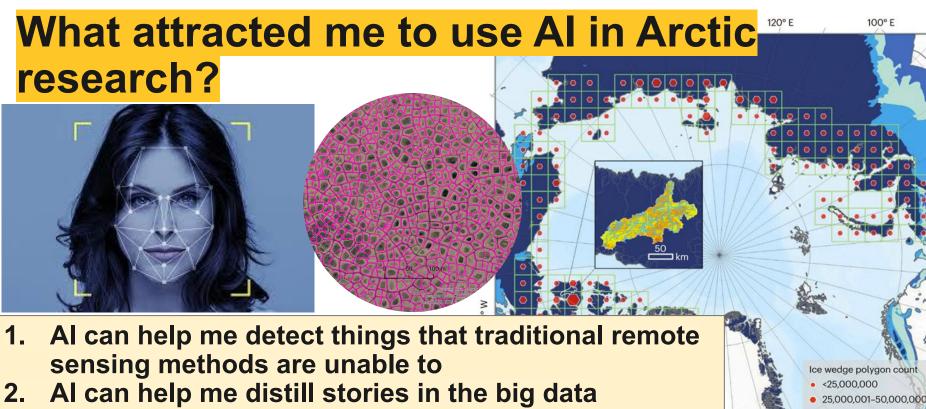
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Text understanding and knowledge representation



75.000.001-100.000.000

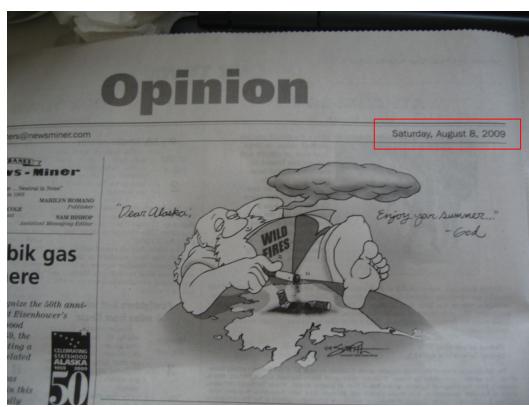
om PAO NOAA, USGS Esri, USGS

- 3. Al can help me keep up with the changing landscape
- 4. Al can help me make connections with data/people Al is my new tool in the toolbox, not my creative partner.

# What attracted me to use Al in Arctic Research?



A visit to Fairbanks, AK in August 2009 for IPY GeoNorth 2009



Data to knowledge production pipeline



NSE

PolarHub: large-scale web crawler for distributed polar data

#### **Metrics:**

- No. of datasets: 194,000
- 7 types of data
- 1,730,000 unique data layers
- from 150+ countries
- Access from 27 countries

#### Significance:

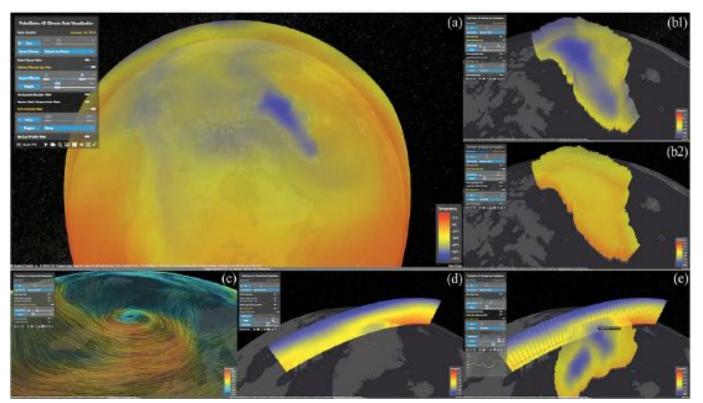
- Support polar science
- Spatial decision support consortium
- Wildfire, water quality, biodiversity conservation communities

P@larHub® For Polar Data Discovery 194K >

https://cici.lab.asu.edu/polarhub3

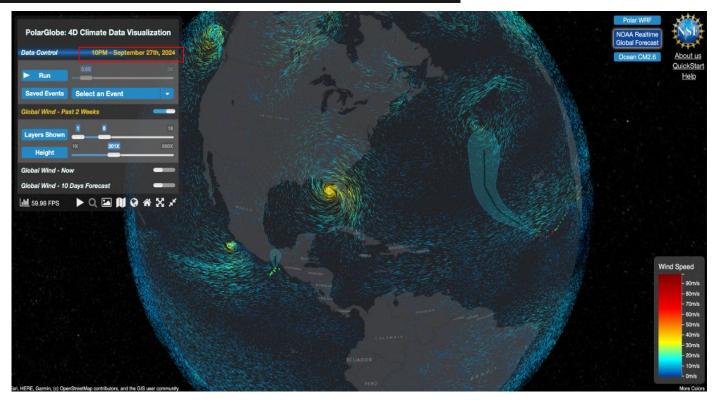
NST

PolarGlobe: Real-time climate data visualization





PolarGlobe: Real-time climate data visualization



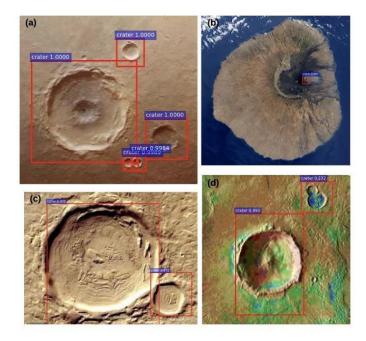
# **GeoAl for Intelligent Mapping**



### Mapping the Earth's natural features - craters



Li et al. 2017

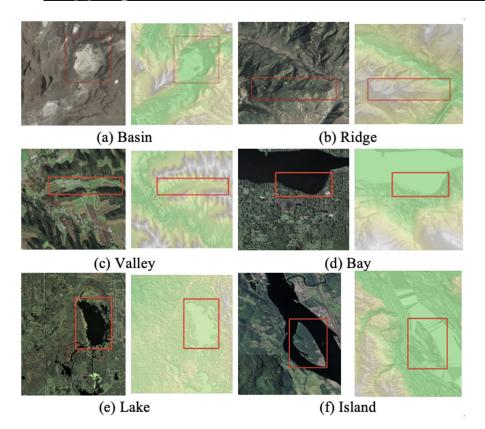


Crater detection using CNN mAP(crater) = 99%

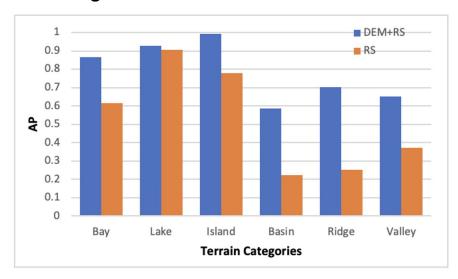
# **GeoAl for Intelligent Mapping**



Mapping the Earth's natural features - multi-source, multi-type



#### GeolmageNet



**AP**: Average Precision

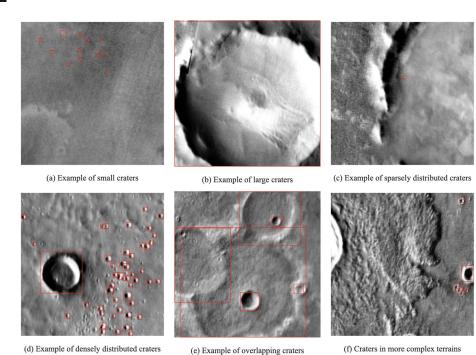
Wang et al. (2021)

# **Mapping Mars craters**

#### Incubator for the NSF Cyber2A project

- 102,675 images scenes covering global Martian surface
- THEMIS image at 100m resolution
- 640k craters





https://cici.lab.asu.edu/martian

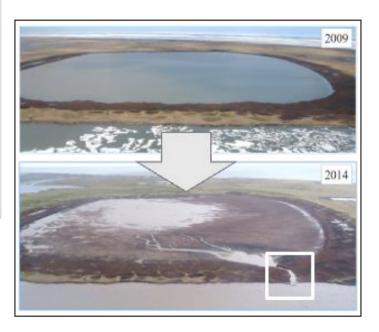
# **Mapping the Arctic Permafrost**







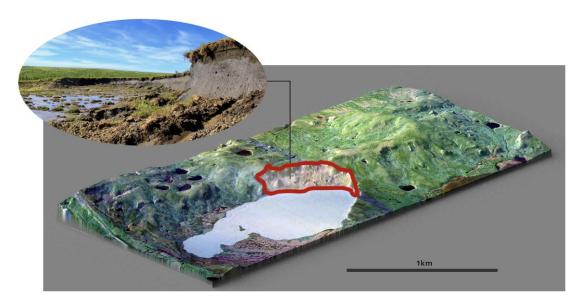




# **Mapping the Arctic Permafrost**



**Ice-Wedge Polygons** 



**Retrogressive Thaw Slump (RTS)** 

# **Mapping Arctic Permafrost Thaw**

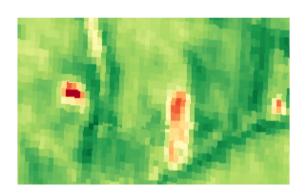
A multimodal + feature fusion approach





Optical image





NDVI
(Normalized Difference
Vegetation Index)

The higher the value, the greener it is

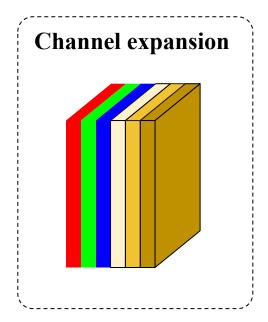


ArcticDEM
(Digital Elevation Model)

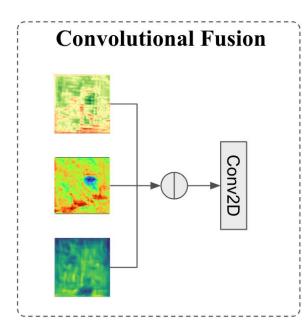
# **Mapping Arctic Permafrost Thaw - Methods**

Feature-level data fusion

**Prior solutions** 

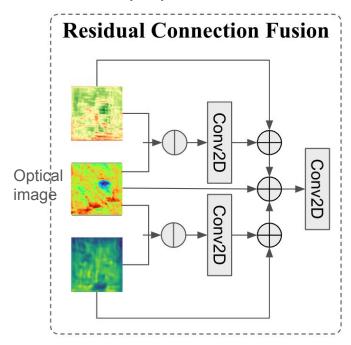


Yang et al. 2023



Ophoff et al., 2019; Wang and Li, 2021

Our proposed solution



Gu, Li, et al. 2024 (under review)

Convright @ 2024 Arizona Board of Regents

# **Mapping Arctic Permafrost Thaw - Results**

Feature-level data fusion

	Fusion Strategy	Input Modalities			
Fusion Type		RGB	RGB +	RGB +	RGB +
			NDVI	DEM	NDVI +
					<b>DEM</b>
Data level	Channel Expansion	39.73	42.74	36.59	43.01
Feature level	<b>Convolutional Fusion</b>	(baseline, no	44.51	43.66	44.59
Feature level	<b>Residual Connection Fusion</b>	fusion applied)	46.08	43.39	45.78

#### **Observations:**

- Multi-modality is helpful
- Fusion method is important
- Low quality (vertical accuracy) of DEM lowered the model performance



ArcticDEM

# **Forecasting Arctic Permafrost Thaw**

- We need to have the trigger data, but existing data are not of desired resolution (1km)
  - Climate
  - Disaster (wildfire)
- Thaw indicator data
  - RTS formation
  - Lake change data (pond formation/lake drainage)

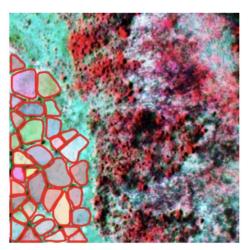
# Al-based medium range weather forecasting

- Significant success in this area
  - Huawei-Pangu (Bi et al. 2023 in Nature)
    - 3D neural network based on Swin Transformer
    - Training data: 30 years of reanalysis (1979-2017) with 69 factors (e.g., T, U, V, Geopotential, humidity) at 13 pressure levels
    - Higher prediction accuracy and 10k times faster than the IFS (Integrated Forecasting System) model from ECMRW
  - Google DeepMind- GraphCast (Lam et al. 2023, Science)
    - More lightweight and higher accuracy than than Pangu

# Al's two big subfields

#### Image analysis/Computer vision

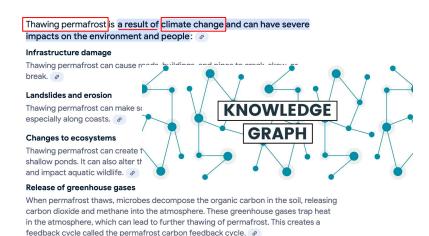
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#### **Natural language processing**

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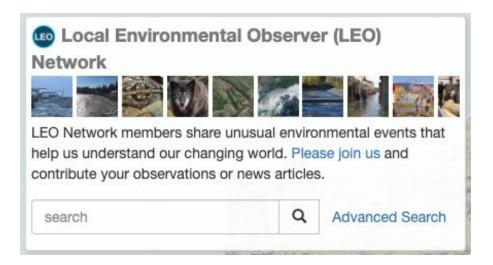


Text understanding and knowledge representation

# Embracing Large Language Models for Arctic Research

# Al enhanced event search and recommendation

Generative AI for deterministic tasks



https://www.leonetwork.org/

#### Humpback found dead near Kodiak gets Alaska's first 2023 whale necropsy

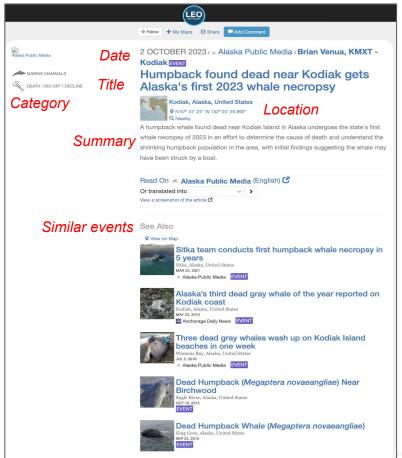
By Brian Versia, KMXT - Kodiak - October 2: 2020



Biologists conduct a necropsy on a humpback whale that washed ashore near Kodiak on Sept. 25, 2023. (Brian Vincum MRT)

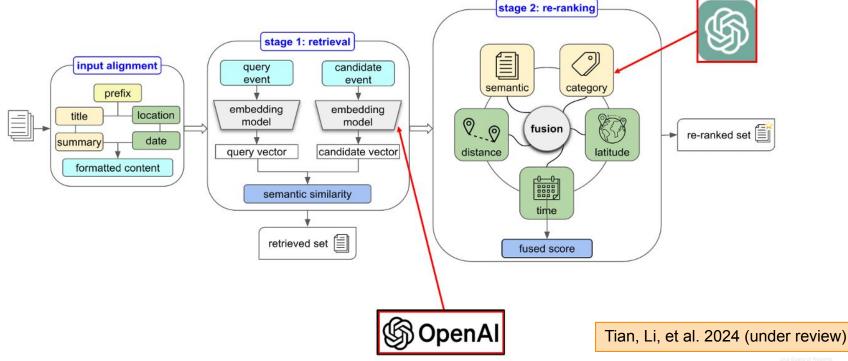
### **LEON Event Search & Recommendation**

- Current method:
  - Full text search (BM25 model)
    - Keyword matching
    - Context matters!
  - Distance & category boosting
  - Manual work:
    - Multiple rounds of keyword-based search and human judgement



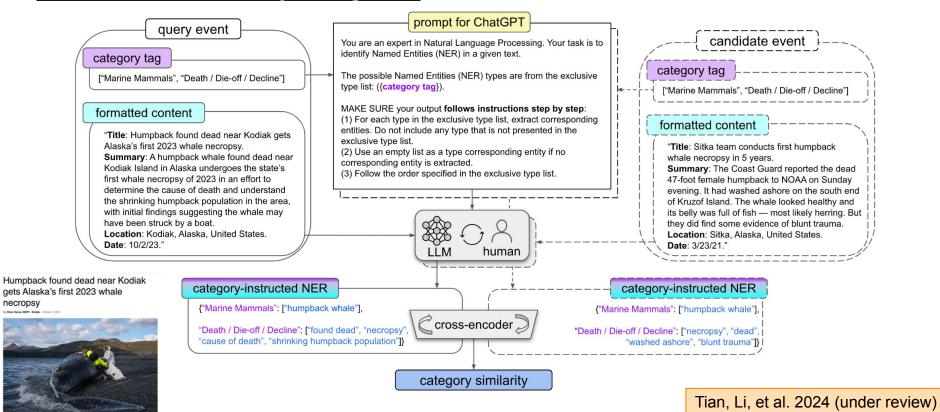
### Al-enhanced search and recommendation

Leveraging large language models (LLM)



### Al-enhanced search and recommendation

#### LLM-based Named Entity Recognition

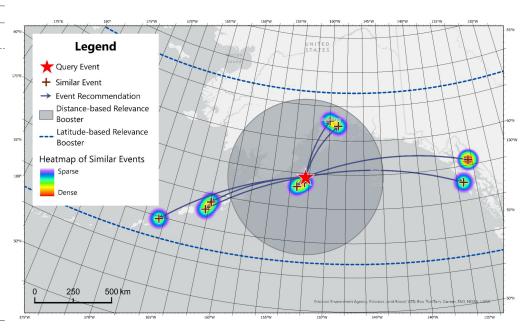


### Al-enhanced search and recommendation

#### Our results

#### LEON human subject ranking

	Rank	Label	Title	Location	Date
query event			Humpback found dead near Kodiak gets Alaska's first 2023 whale necropsy	Kodiak, Alaska, United States	10/02/23
	#1	hit	Alaska's third dead gray whale of the year reported on Kodiak coast	Kodiak, Alaska, United States	05/23/19
	#2		Dead Humpback Whale (Megaptera novaeangliae)	Old Harbor, Alaska, United States	08/23/12
	#3	hit	Sitka team conducts first humpback whale necropsy in 5 years	Sitka, Alaska, United States	03/23/21
	#4		The dead whale floating in Cook Inlet has washed ashore at Kincaid Park	Cook Inlet, Anchorage, Alaska, United States	09/25/17
similar	#5	hit	Dead Humpback Whale (Megaptera novaeangliae)	King Cove, Alaska, United States	09/23/15
events	#6		Beluga whale found dead south of Anchorage will help scientists better understand the endangered animals	Girdwood, Alaska, United States	05/27/21
	#7		Dead whales wash up near Unalaska, but pandemic complicates necropsies	Unalaska, Alaska, United States	08/27/20
	#8		Dead humpback whale calf washes up near Juneau, may have been struck by vessel	Juneau, Alaska, United States	08/28/23
	#9		Dead Humpback Whale (Megaptera novaeangliae) Floating by Moller Point	King Cove Alaska	08/18/20
	#10		Whale's body spotted near Tenakee Inlet	Juneau, Alaska, United States	02/10/22



# Integration of vision and language models

# Vision language model for LEON search and recommendation

Image as an important cue for theme



3 AUGUST 2021 / Anchorage Daily News / Ray Levy Uyeda, The Washington Post

Climate change is endangering sacred land. For these Native women, it threatens 'everything we are.'



Newell, California, United States

O N 41° 36' 39 44" W 121° 33' 13 09'

Wildfires in Western states have razed structures, displaced residents and altered air quality - and fundamentally changed the relationship between Native women and the land they have historically stewarded.



AND MAMMALS

UNUSUAL ANIMAL BEHAVIOR

29 SEPTEMBER 2020 / A Alaska Public Media EVENT

#### Bear in Denali National Park euthanized over safety concerns



Denali National Park, Alaska, United States

Q N 63° 32' 26.808" W 151° 43' 24.996"

Q Nearby

The bear had entered buildings and food caches, according to National Park Service

officials.

Read On **A Alaska Public** 

Or translated into

View a screenshot of the article &



San Francisco Chronicle



31 MARCH 2021 / € San Francisco Chronicle / Steve Rubenstein EVENT

#### A gray whale carcass washed ashore at S.F.'s Crissy Field



San Francisco, California, United States

Q N 37° 48' 23 04" W 122° 27' 58 656"

Q Nearby

A gray whale carcass washed ashore at Crissy Field on Wednesday morning. Veterinarians and biologists from the center and from the California Academy of Sciences plan to perform a necropsy in coming days on the island to determine the cause of death

# **Extracting visual cues with GPT-4o**





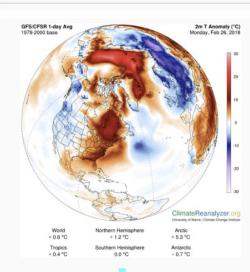
#### **Cue Category**

The image features polar bears, which are marine mammals.

#### **Cue Location**

Polar bears are native to the Arctic region.







#### Cue\_Category

The image shows a temperature anomaly map, which is directly related to weather patterns.

#### Cue\_Location

The map prominently displays the Northern Hemisphere, with clear indications of temperature anomalies in this region.

#### Cue\_Time

The date on the image is February 26, 2018, which falls in the winter season in the Northern Hemisphere.

# **Extracting visual cues with ChatGPT**

Location + category





#### Cue\_Category

The image features polar bears, which are marine mammals.

#### Cue\_Location

Polar bears are native to the Arctic region.



#### Cue\_Category

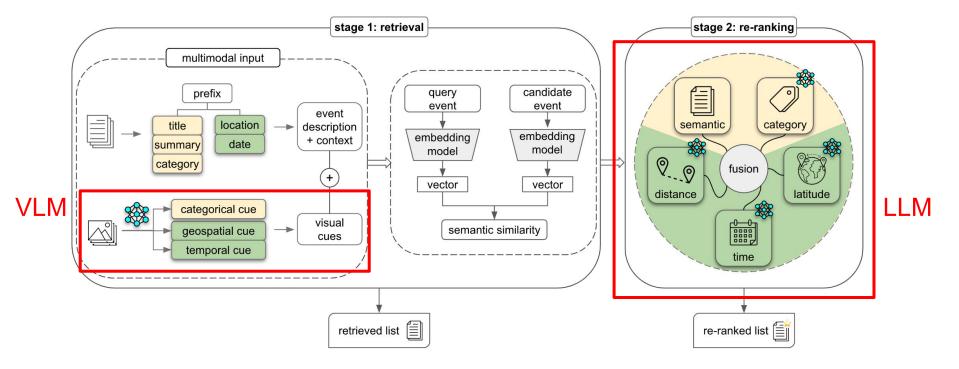
The presence of crosses indicates a burial site. The visible ground and ice suggest changes in permafrost.

#### **Cue Location**

The landscape, including the frozen ground and sparse vegetation, is characteristic of the Arctic region.

## A new VLM-enhanced search strategy

VLM: Vision language model



## A new VLM-enhanced search strategy

#### Results

Ranking performance metrics

**nDCG**: normalized discounted cumulative gain

MRR: Mean Reciprocal Rank

Table 2: Comparison of Re-ranking Performance on Unimodal Representation and Multimodal Representation(metric@10)

Representation	Retrieval	Re-ranking	nDCG	MRR
unimodal	Ada-002	-1	44.70	52.20
Text-based	Ada-002	MiniLM	42.10	49.00
	Ada-002	BGE reranker	36.40	44.40
	Ada-002	Cohere rerank	33.10	38.90
	Ada-002	GPT-3.5 Turbo	32.30	42.50
	Ada-002	RankGPT	41.40	49.00
	Ada-002	GT-R	47.40	55.80
multimodal	emb3-large	_	50.48	57.28
Text + image	emb3-large	GT-R	50.80	57.07
	emb3-large	ASTRA (ours)	51.22	57.60

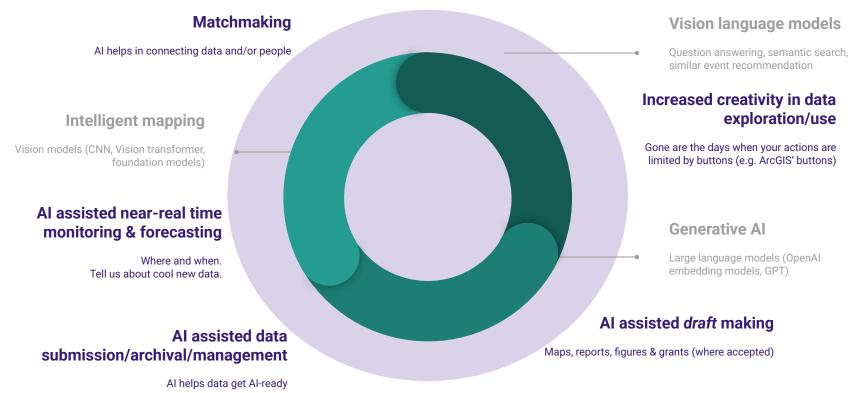
# A quick summary of our research under Permafrost Discovery Gateway



# Big vision and aspiration for PDG

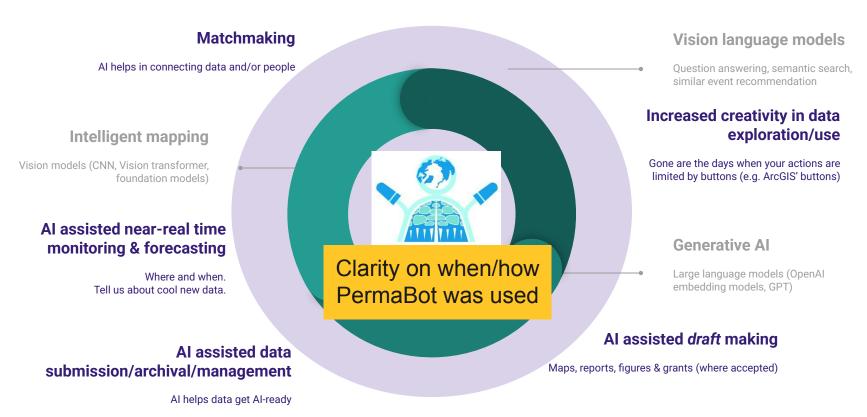






# Big vision and aspiration for PDG

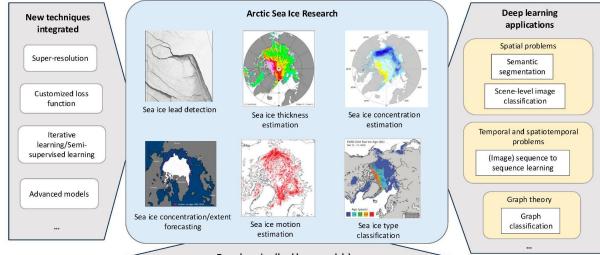
Al tools → More time for our creativity and (real-world) action!



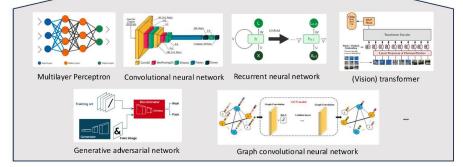
# **Moving beyond Permafrost Research**

## Al for Sea Ice Remote Sensing

#### What is happening in the sea ice world?







Li, W., Hsu, C. Y., & Tedesco, M. (2024). Advancing Arctic Sea Ice Remote Sensing with Al and Deep Learning: Opportunities and Challenges. *Remote Sensing*, 16(20), 3764.

# Al for Sea Ice Remote Sensing

**Table 4.** Deep learning solution techniques for sea ice concentration forecasting.

Sea Ice Application	Deep Learning Problem Formulated	Deep Learning Techniques (Models)	Output	References
Sea ice concentration forecasting	Time-series forecasting	Semantic segmentation (U-Net, SICNet)	Classify each pixel into a SIC type, with multi- channel output and each channel represents the SIC classification for a future timestamp (e.g., month)	[66,68,74]
		RNN (LSTM, attention- based LSTM, gated RNN)	Predict sea ice concentration in future timestamps	[75,76,77,78]
		CNN+RNN (ConvLSTM, multi-task ConvLSTM)	<ul> <li>through sequence-to-sequence learning. The forecast could be for one or multiple timestamps</li> </ul>	[69,71,72,79,80]

## Al for Sea Ice Remote Sensing

#### Future research directions

- Enhanced multimodal deep learning capabilities
- Better ability to quantify uncertainty
- Deeper integration with physics-based models
- Better leveraging AI foundation models
- Open benchmark datasets and open models
  - Al4Arctic (for sea ice charting)
  - SealceWeather (from cruise)
- ...

# Big models, big concerns

 Transformer does not how to reason, they are only advanced pattern matching tools

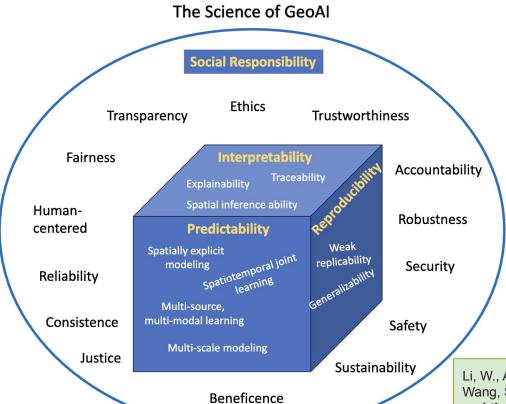
> GSM-Symbolic: Understanding the Limitations of Mathematical Reasoning in Large Language Models

> > Iman Mirzadeh<sup>†</sup> Keivan Alizadeh Hooman Shahrokhi<sup>\*</sup> Oncel Tuzel Samy Bengio Mehrdad Farajtabar<sup>†</sup>

Apple

https://arxiv.org/abs/2410.05229

# Big models, big concerns



Li, W., Arundel, S., Gao, S., Goodchild, M., Hu, Y., Wang, S., & Zipf, A. (2024). **GeoAl for Science and the Science of GeoAl**. *Journal of Spatial Information Science*, (29), 1-17.

### Thank you!

Follow the PDG news here: <a href="https://arcticdata.io/catalog/portals/permafrost">https://arcticdata.io/catalog/portals/permafrost</a>

Follow Cyber2A events here: <a href="https://cyber2a.github.io/">https://cyber2a.github.io/</a>

Contact:

wenwen@asu.edu, aliljedahl@woodwellclimate.org