

Wir leben Fundraising – mit Freude und im Dialog.

Herzlich willkommen





Prof. Giuseppe Ugazio,

Edmond de Rothschild Foundations Chaired Assistant Professor in Behavioral Philanthropy Geneva Finance Research Institute, University of Geneva





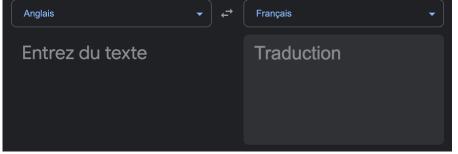
Artificial Intelligence













A system that performs tasks that would require intelligence if performed by humans





Artificial Intelligence and Philanthropy

As AI is transforming society

It is not a matter of "if", but of "when" and "how"

Little awareness and attention from the philanthropic sector

Philanthropy operates with private logics to promote common good

→

Al is bound to transform philanthropy

philanthropic organisations will adopt AI and confront its ethical impact



For use in Philanthropy Impact on society



Privileged position to guide Al towards common good





Al and Philanthropy: two areas of focus



Al for Philanthropy: developing tools that support philanthropic organizations' operations and strategies



Philanthropy for AI: Leveraging philanthropic organizations' potential to ensure the ethical and inclusive development of Artificial Intelligence





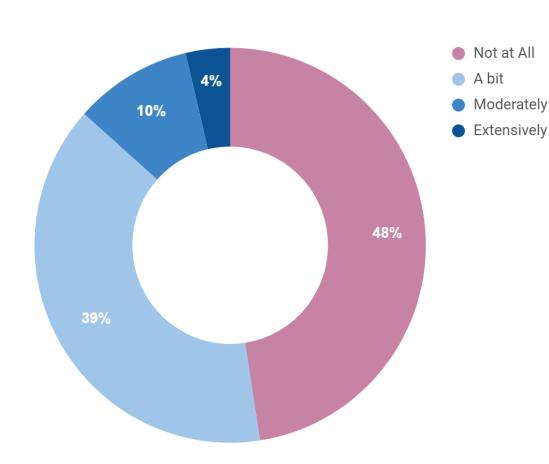
A Survey of AI and Philanthropy among Swiss POs

- > Do organizations use Al?
- If yes, for what purposes is AI used?
- > In which areas would organizations want to have Al assistance?
- > What are the main concerns on the use of AI?





Is AI used by Swiss POs?

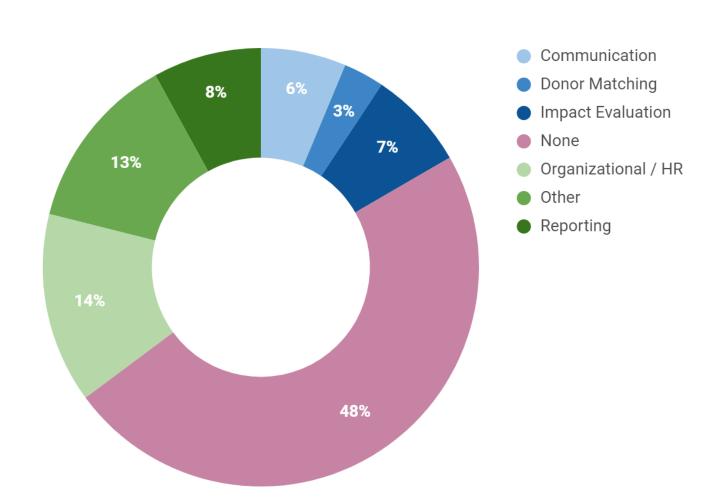


A word of caution: our results are likely biased (organizations using Al more likely to participate in the survey)





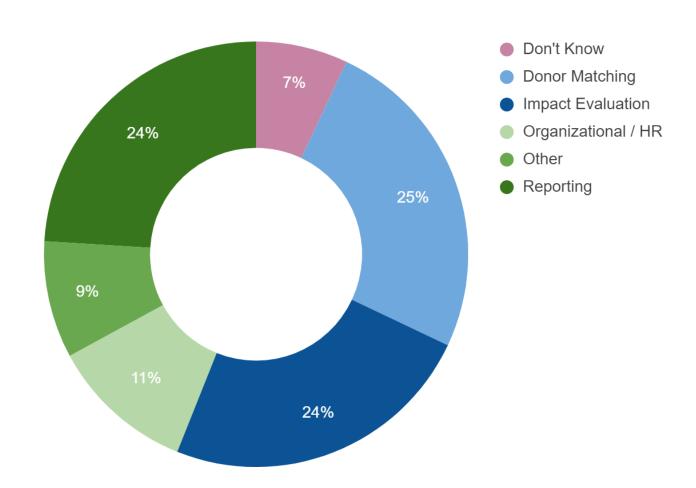
In what area is AI Used?







Al assistance would be helpful for my organization in:







What are the main concerns of using AI?







A promising form of AI for Philanthropy: Natural Language Processing

- Decision tree classification of foundations in International
 Classification of Nonprofit Organizations (INCPO –
 International.)
 categories data = foundations' names Litofcenko et al., 2019
- Several methods for classifying foundations in National Taxonomy of Exempt Entities (NTEE – U.S.) categories – Ma et al., 2021

Limitation: Classifications are static, philanthropy & society is dynamic





Supporting Philanthropy Strategic Planning with NLP

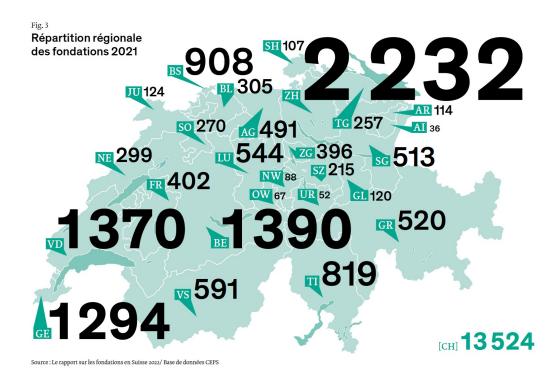
- > In what and in how many areas do POs operate?
- According to which Type do these work (Donors vs. Operating)?
- Are different Types of POs using a similar language?
- > What motives determine POs engagement?
- > To which extent are the U.N. SDGs addressed by POs?





Philanthropy in Switzerland a use case on AI for Philanthropy

- About 17.000 POs in Switzerland
- No comprehensive database of these POs







Big NLP Data on Swiss POs

- Collected Mission Statements from ≈ 13.000 Philanthropic Organizations
- > Identified 8485 unique and meaningful words from 12766 POs.

Translation → Word Tokenization → Lemmatization → TF-IDF scoring & pruning

Fname [‡]	word [‡]	n [‡]	total [‡]	tf [‡]	idf [‡]	tf_idf [‡]
Foundation 1	community	1	4	0.25	0.6931472	0.1732868
Foundation 1	parish	1	4	0.25	0.6931472	0.1732868
Foundation 1	promotion	1	4	0.25	0.6931472	0.1732868
Foundation 1	youth	1	4	0.25	0.0000000	0.0000000
Foundation 2	clubs	1	4	0.25	0.6931472	0.1732868
Foundation 2	leaders	1	4	0.25	0.6931472	0.1732868
Foundation 2	training	1	4	0.25	0.6931472	0.1732868
Foundation 2	youth	1	4	0.25	0.0000000	0.0000000



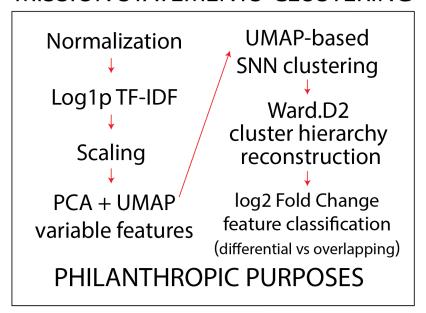


Our approach in a nutshell

Machine Learning algorithms (NLP) to analyze Philanthropic Communications

Unsupervised Learning – Philanthropic Landscape (Cluster organizations in Areas of Activity)

MISSION STATEMENTS' CLUSTERING







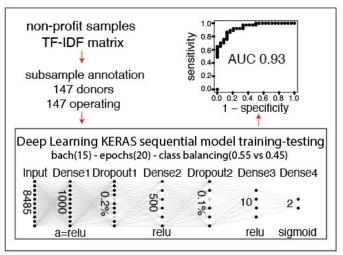
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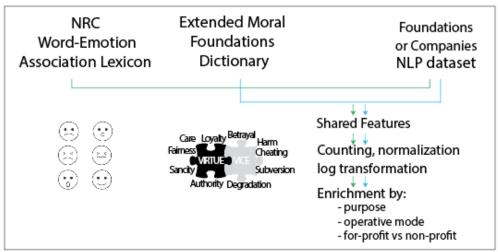
Supervised Learning

Operation Mode, communication overlap, motives, SDGs

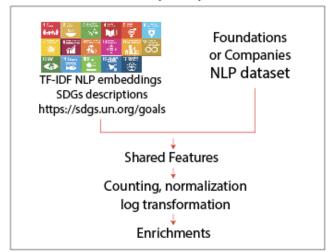
OPERATIVE MODE PREDICTION



LEXICON-BASED ANALYSIS: Emotions & Moral Values



Sust Dev Goals (SDG) COVERAGE

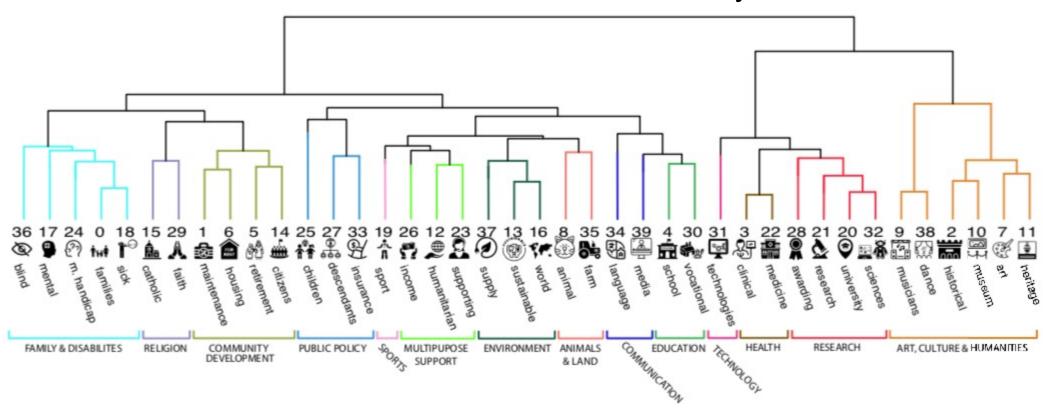






In what areas do POs operate

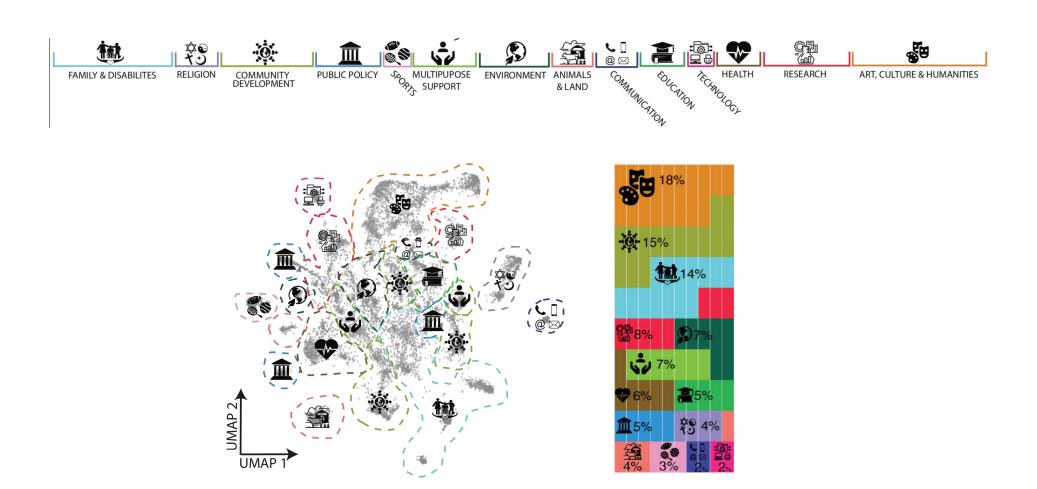
40 different Areas of Activity







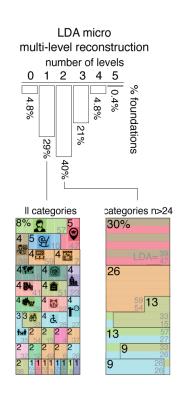
In what areas do POs operate

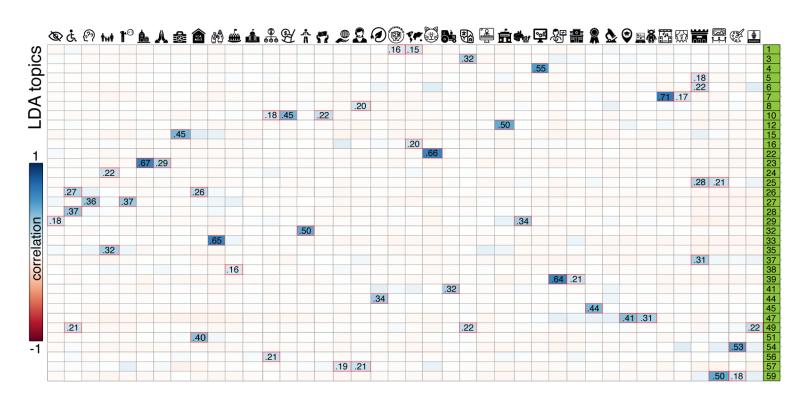






How engaged is a PO in a given area of activity







1° - 66% LDA topic 12 – education macrodomain

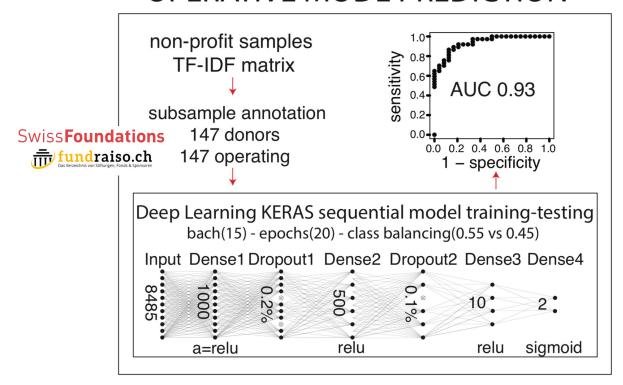
2° - 25% LDA topic 21 – family macrodomain

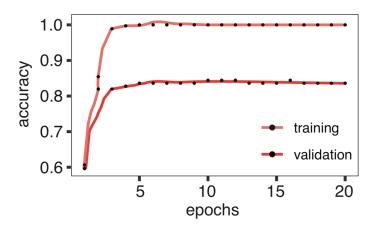




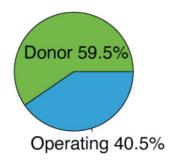
Classifying Type of POs

OPERATIVE MODE PREDICTION





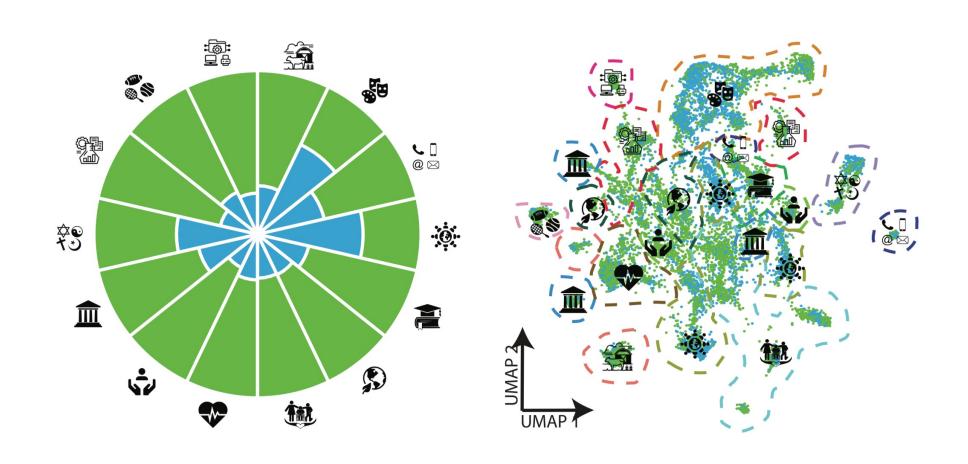








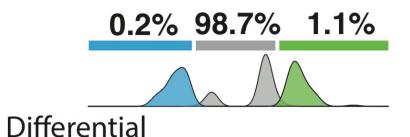
Type of POs Across Areas of Activity



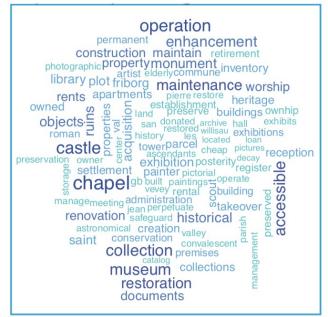




Large overlap in language used by different POs



requests developing poverty deprived situations supports orphanages visually technology responsibility ecology training technology responsibility ecology training training technology responsibility ecology training training training technology responsibility ecology training traini



Overlapping 100 Most Generic Words

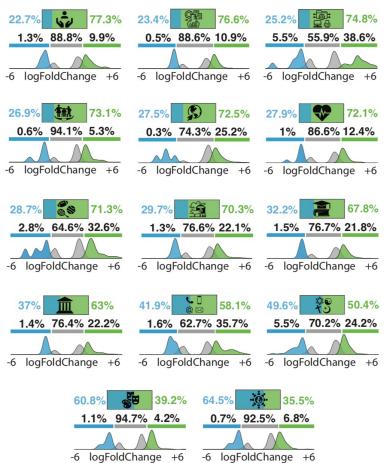
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steering sophisticated minorities appearance infected appearance hedges creators madagascar spouse facing options vegetables encumber endeavor pedagogue resistance emergence in fected facing options vegetables encumber endeavor pedagogue resistance facilitates secretariat facilitates secretariat vegetables encumber endeavor pedagogue resistance facilitates secretariat facilitates secretariat vegetables encumber endeavor pedagogue resistance facilitates secretariat vegetables vegetables secretariat vegetables vegetabl
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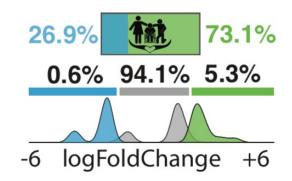


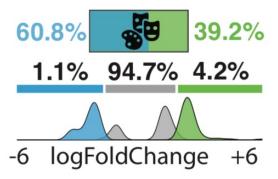


Language Overlap Differs by Type of Activity

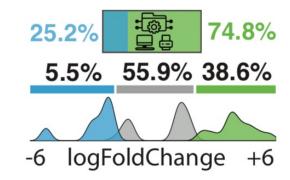
High Shared Language

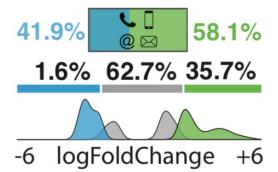






Low Shared Language







"Related Dialect" Spoken in 6/14 areas of activity

Related Dialects = 85% or more words shared across two languages (Eberhard, D. et al, 2020)



89%





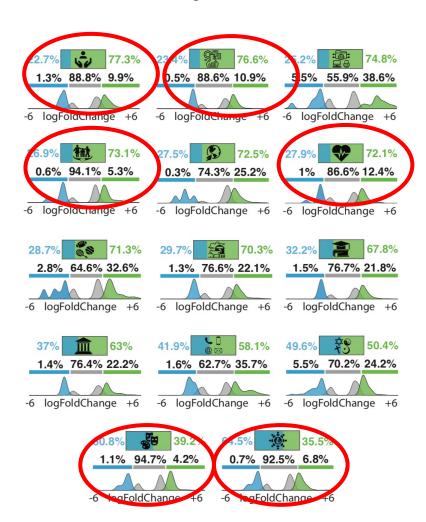
60%





27%

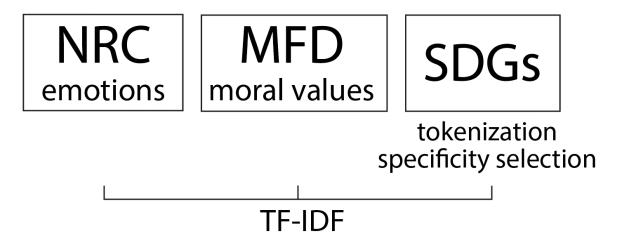








Investigating Motives: "supervised learning"



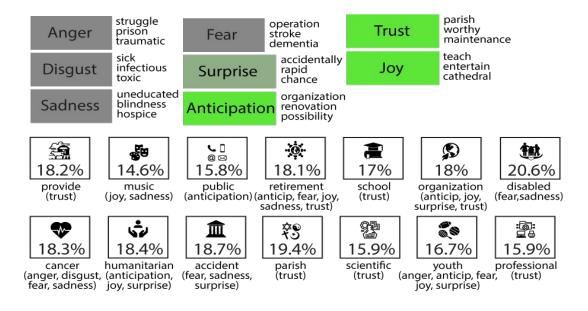
Defining salient words related to Emotions, Moral Values and the SDGs.

Using these salient words to identify which organizations mention them in their mission statements to infer motives





Emotions

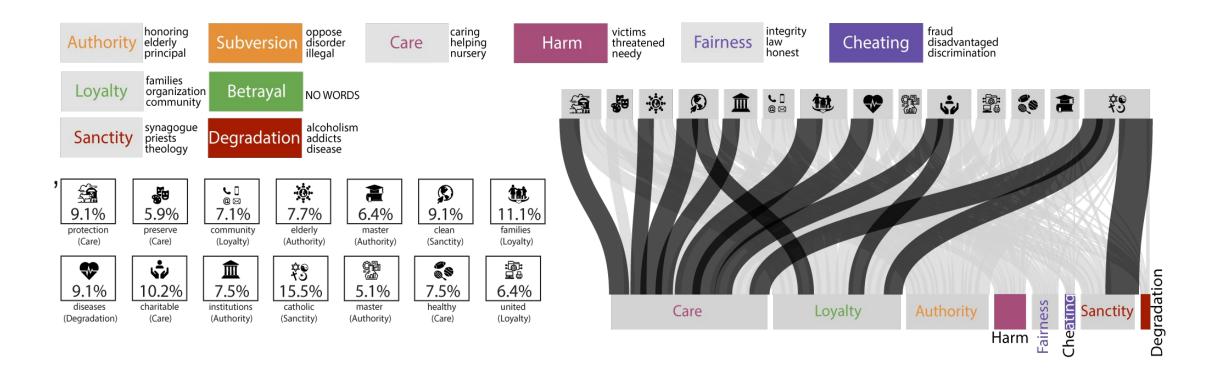








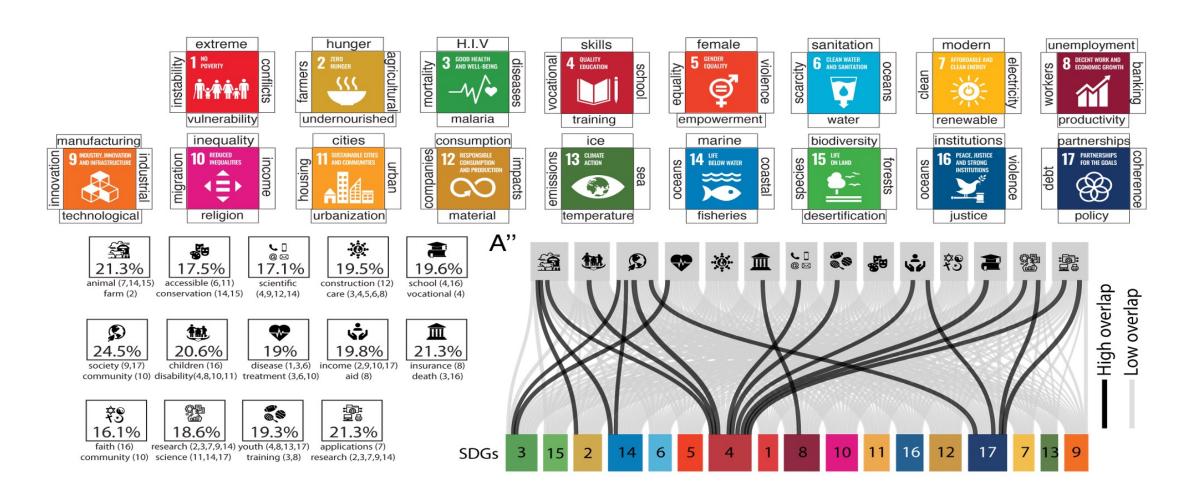
Moral Values







Sustainable Development Goals

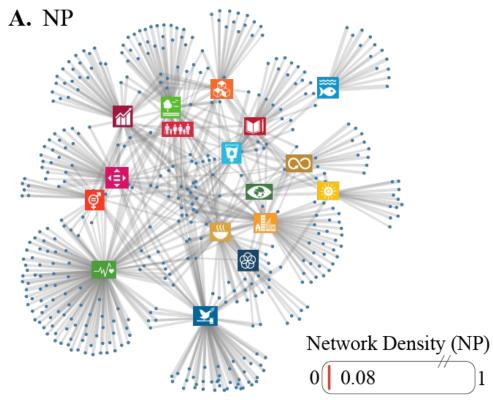






Weak collaborations among POs in tackling the SDGs

"SDG Connection-Network based on For-Profit missions"



- organisations
- SDG connections (shared words)



































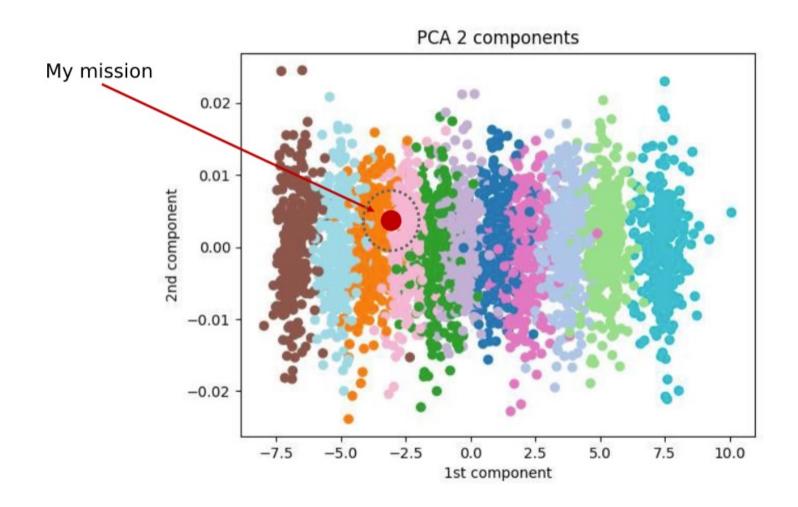








From Classification to Flexible SDG-Localization







Conclusions

- > All can empower philanthropic collaboration by mapping POs:
 - Areas & Types of Activity
 - Geographical Distribution
 - Motivations, emotions & moral values
 - Informing Strategic Thinking to align with SDGs
- > Key Advantages:
 - Can be used to find patterns too hard for humans to detect
 - Open Access, can benefit from inputs from anyone and easily replicable





Thanks for your attention!

The AI and Philanthropy Team:

- Lucia Gomez Teijeiro, PhD
- Camilla della Giovampaola
- Hubert Halope
- Maria Cristiana Tudor
- Nisa Thomas



https://www.unige.ch/BehavioralPhilanthropyLab

Join us for the 10:45 workshop: **Deep-Dive «Empowering philanthropy with AI and big data»**



Fragen und Anregungen?

Die Geschäftsstelle von Swissfundraising steht bei Fragen oder Anliegen immer gerne zur Verfügung.

Swissfundraising

Rosenbergstrasse 85 9000 St.Gallen

Telefon: <u>+41 71 777 20 11</u>

E-Mail: info@swissfundraising.org



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