

Natural language processing for enriching real world evidence from electronic health records

...NLP @ Health Campus The Hague

Den Haag

Spring Symposium Young Epidemiologists, @UMC Utrecht, 21 March 2024, Marco Spruit

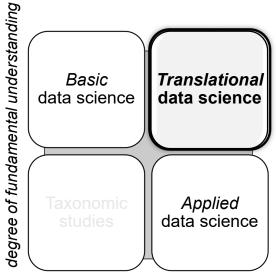




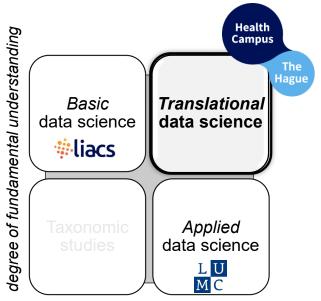








degree of practical use consideration



degree of practical use consideration

Basic data science data science data science data science data science LUMC

degree of practical use consideration

Chapman, P., Clinton, J., Kerber, R., Khabaza, T., Reinartz, T., Shearer, C., & Wirth, R. (2000). CRISP-DM 1.0: Step-by-step data mining guide. *SPSS inc*, *9*(13), 1-73. https://books.google.com/books/about/CRISP_DM 1_0.html

Translational Data Science in Population Health 2 **Domain** Data Understanding Insight 3 Data Preparation Cross-6 Model Domain Deployment Health Data 4 Data Modelling (5) Model **Evaluation** Data Engineering Data Analytics e-Health Implementation



Preprints are preliminary reports that have not undergone peer review. They should not be considered conclusive, used to inform clinical practice, or referenced by the media as validated information.

Extracting Patient Lifestyle Characteristics from Dutch Clinical Text with BERT Models

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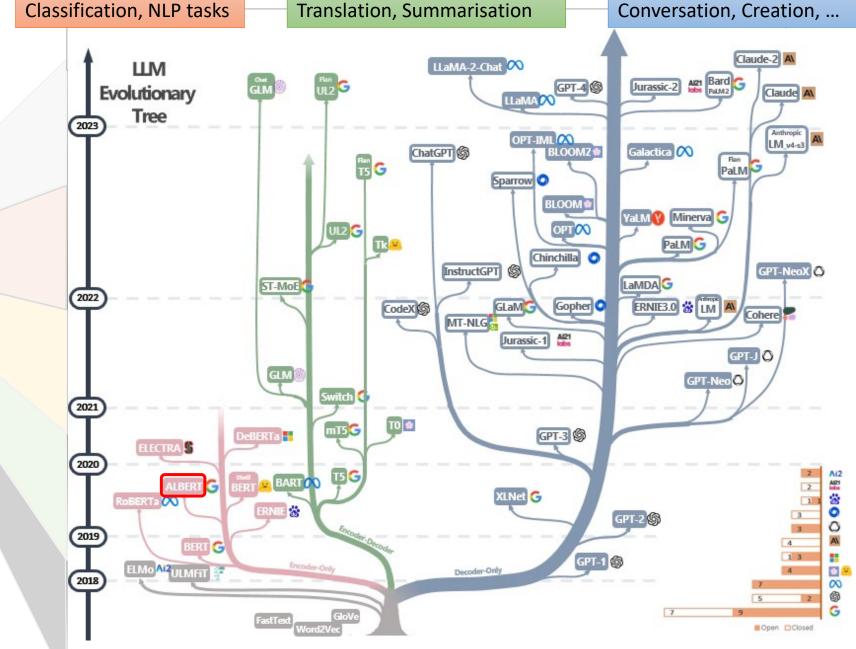


Research Article

Keywords: BERT, BERT Clinical Research, Clinical NLP, NLP Clinical Lifestyle Classification

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DOI: https://doi.org/10.21203/rs.3.rs-3831694/v1



Yang, J., Jin, H., Tang, R., Han, X., Feng, Q., Jiang, H., ... & Hu, X. (2023). Harnessing the power of Ilms in practice: A survey on chatgpt and beyond. *ACM Transactions on Knowledge Discovery from Data*. https://doi.org/10.1145/3649506

Which LLM is best suited for clinical text classification?

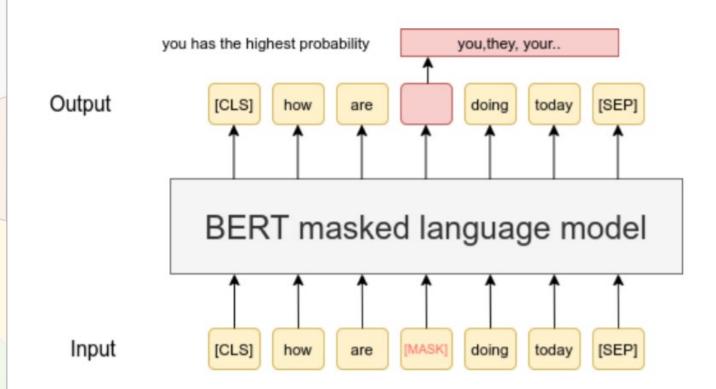
Lifestyle information extraction for personalised prognoses

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Which LLM is best suited for clinical text classification? Lifestyle information extraction for personalised prognoses

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BERT high-level architecture



[MASK]: Masked Language Modelling (MLM) → Word prediction

[CLS] : Classification for Next Sentence Prediction (NSP)

degree of fundamental understanding BERT-based Dutch NLP Translational on sloppy data science informal medical text snippets Lifestyle information extraction for personalised prognoses degree of practical use consideration

van Haastrecht, M., Sarhan, I., Yigit Ozkan, B., Brinkhuis, M., & Spruit, M. (2021). SYMBALS: A systematic review methodology blending active learning and snowballing. Frontiers in research metrics and analytics, 6, 685591. https://doi.org/10.3389/frma.2021.685591

Domain understanding [1/6]: Determining Objective

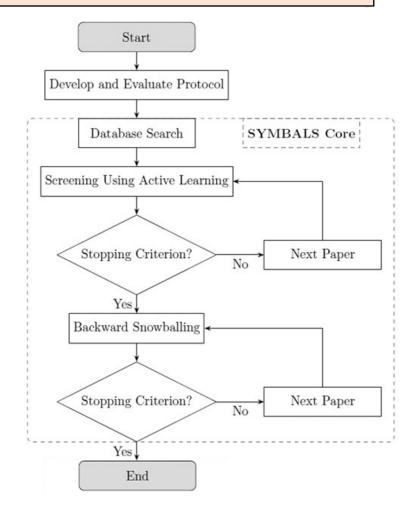


Haga Ziekenhuis

- Case study
- <u>Previous</u> research
- Lifestyle characteristics
- NLP task \rightarrow BERT

Systematic review

- Active learning
- Backward snowballing
- $SYMBALS \rightarrow$
- 85 papers



Objectives

- Pretraining BERT model from scratch
- Pretraining on top of Dutch BERT models
- 3. Translation strategy

Data Insight [2/6]: Data Description

"Anamnesis: Painful, cold soles of the feet. Especially in the cold months. Is now known to have hip complaints. Doesn't know if those originated from back issues. Tract history: Tract history: DM - chol - Smoking - Hypertension. Intervention for spinal stenosis. Physical examination: adp and atp +/+ now normal color. Additional examination: Index: pressure of 1.2 bdz. Conclusion: Conclusion: Foot complaints related to the phenomenon of Raynaud. Possible in case of neurogenic dysregulation in case of possible recurrent spinal canal stenosis.

No indication for PAOD Policy: Policy: Explanation. Wait and see analysis of the hip and, if necessary, start neurology traject."

Descriptive statistics

Type of label	#labelled texts	Current users	Former Users	Non-users	Unknown
Smoking	148.768	7.015 (4.72%)	32.230 (21.66%)	44.677 (30.03%)	64.846 (43.59%)
Drinking	143.166	16.017 (11.25%)		39.119 (27.32%)	87.940 (61.43%)
Drugs	147.999	1.443 (0.98%)		53.005 (35.81%)	93.551 (63.21%)

degree of fundamental understanding

BERT-based
Dutch NLP
on sloppy
informal
medical text
snippets

Translational data science

Taxonomic studies Lifestyle information extraction for personalised prognoses

degree of practical use consideration

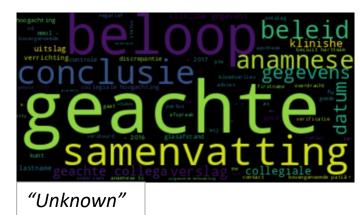
Data Insight [2/6]: Exploratory Data Analysis

Checking label quality after obtaining suspicious performance

Stochastic Gradient Descent class: Alcohol Use	Unknown F1-score	Current F1-score	Non-user F1-score	Macro F1- score
(Ngram 2, Stopwords kept)	1.00	0.99	0.99	0.99
(Ngram 2, No stopwords)	1.00	0.97	0.99	0.99
(Ngram 1, Less stopwords)	0.95	0.61	0.69	0.75

No BERT necessary??

"Current user"



"Non-user"





BERT-based Dutch NLP on sloppy informal medical text snippets Taxonomic studies Lifestyle information extraction for personalised prognoses

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Data Insight [2/6]: Exploratory Data Analysis

• Exemplar *edge* cases from the Smoking task in clinical notes that were predicted as *Non-user* class but were misclassified.

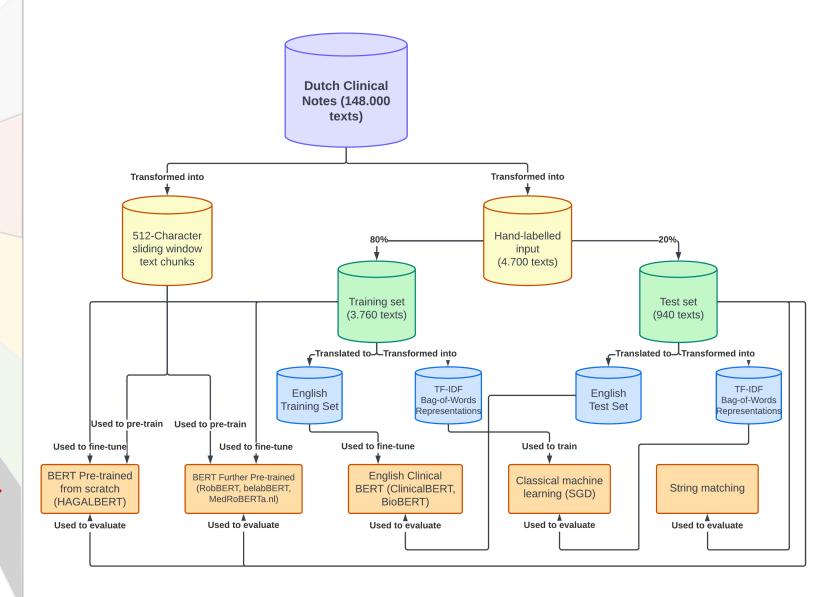
Edge cases in Dutch clinical notes	English translation	Truth
Intoxicaties: roken 2-4 sigaren per dag alleen in de zomer,	Intoxications: smokes 2-4 cigars per day only in the summer,	Current user
roken (+ (20 packyears, is gestopt); Risicofactoren: Familieanamnese (+); roken (-)	smoking (+ (20 packyears, has stopped); Risk factors: Family history (+); smoking (-)	Former user
Roken: 3-4 per dag al jaren lang, vroeger wel meer	Smoking: 3-4 a day for years, in the past more	Current user
Roken: - (20 jaar geleden na 50 py), Cardiovasculaire risico-factoren: roken:+	Smoking:- (20 years ago after 50 py), Cardiovascular risk factors: smoking:+	Former user

- Top 50 texts with highest scores for Current user class that were predicted Non-user
- ... Hand-labeling #4700 texts

True label	#
Unknown	30/50
Current user	5/50
Non user	10/50
Former user	5/50

Computational experiments

Data preparation [3/6]



Data Modelling [4/6] -Experimental Results

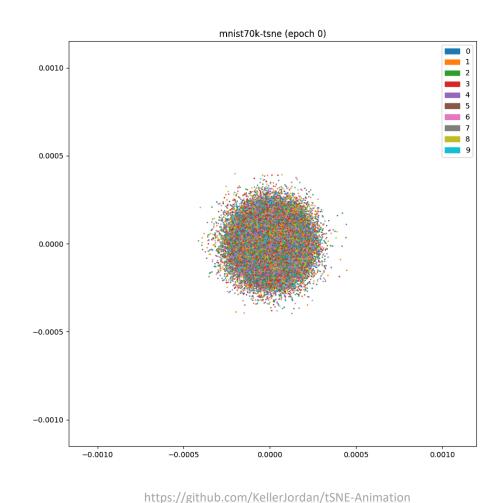
	Model	Smoking	Alcohol	Drugs
	[a] Traditional models			
	String Matching	0.84	0.74	0.68
	Machine Learning (SGD)	0.85	0.71	0.60
	[b] Trained from scratch			
	HAGALBERT	0.66	0.54	0.43
	[c] Fine-tuned models			
	RobBERT-HAGA	0.87	0.71	0.63
	belabBERT-HAGA	0.48	0.64	0.57
	MedRoBERTa.nl-HAGA	<mark>0.93</mark>	0.79	<mark>0.77</mark>
[d] Large English models				
	BioBERT (translated)	0.91	0.72	0.52
	ClinicalBERT (translated)	0.92	<mark>0.80</mark>	0.61

BERT-based Dutch NLP on sloppy informal medical text snippets Lifestyle information extraction for personalised prognoses

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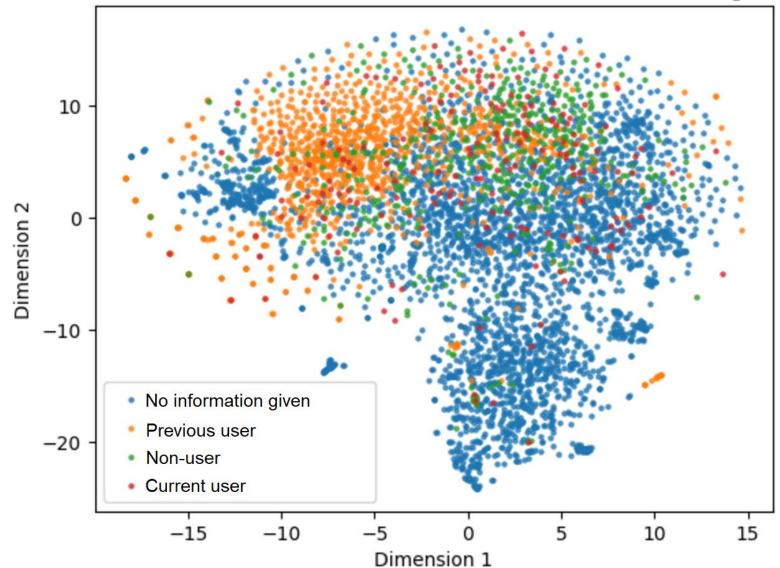
Model Evaluation [5/6] - t-SNE viz

• t-SNE (t-distributed Stochastic Neighbor Embedding): for nonlinear dimensionality reduction (e.g. PCA)

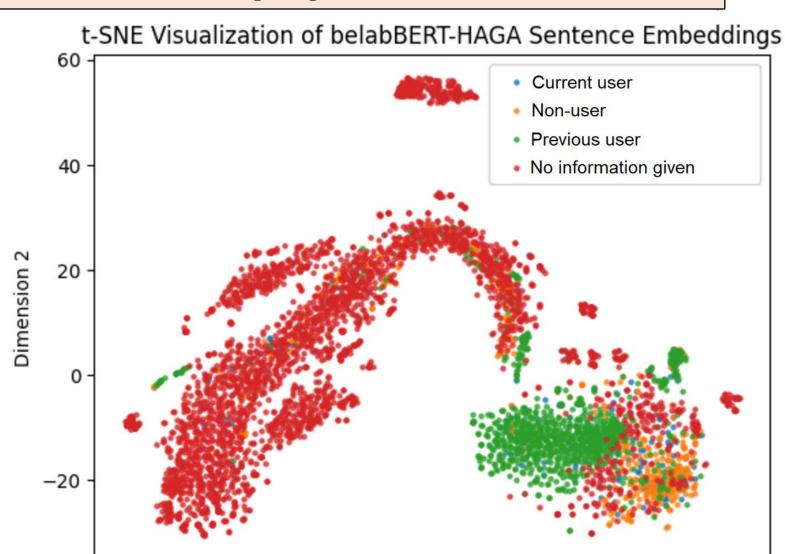


Model Evaluation [5/6] - t-SNE viz: HAGALBERT





Model Evaluation [5/6] – t-SNE viz: belabBERT-HAGA



20

Dimension 1

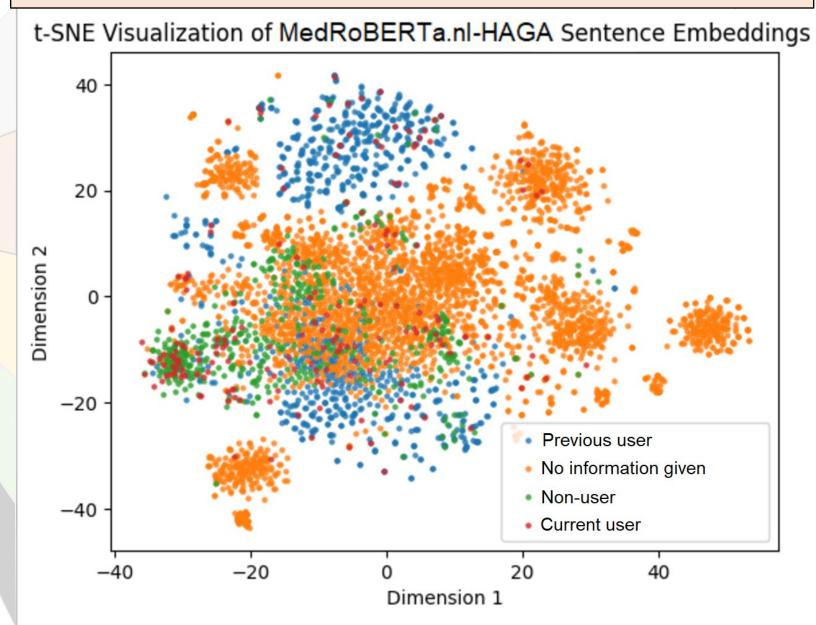
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60

-40

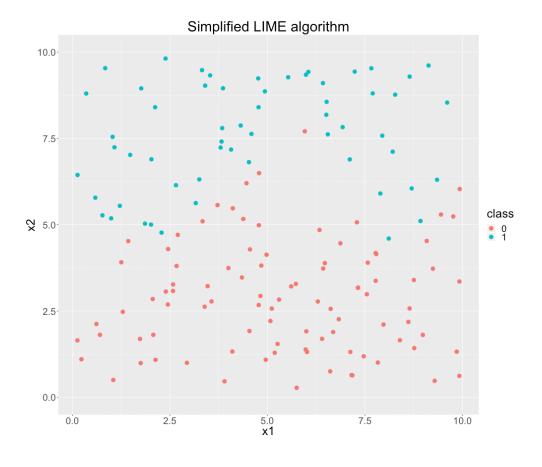
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Model Evaluation [5/6] - t-SNE viz: MedRoBERTa.nl-HAGA



Model Evaluation [5/6] - LIME

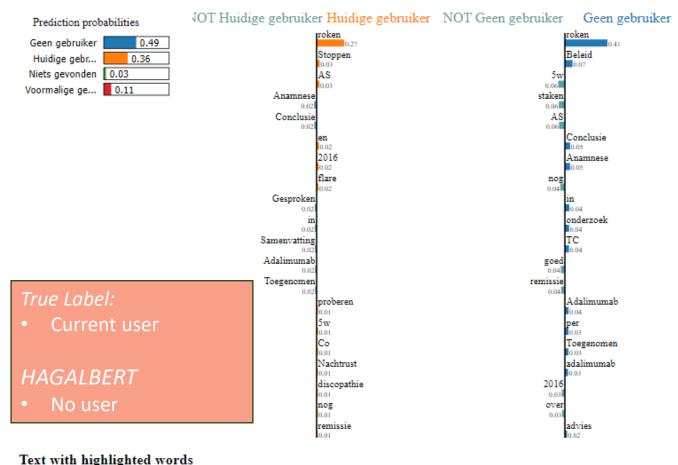
 LIME: Local interpretable model-agnostic explanations, to train local surrogate models to explain individual predictions



https://slds-lmu.github.io/iml methods limitations/lime.html

Model Evaluation [5/6] – LIME: HAGALBERT

"Summary: Course of radiology: 2014 Increased spondyloarthrosis and disc disease. 2016 flare after discontinuation. Case history: Talked about smoking, advice to stop. Sleeps well, also functions well in daily life. Adalimumab still per 4w. Physical examination: Hand osteoarthritis. Conclusion: Conclusion: AS in remission under adalimuma.b Policy: Policy: Co 1j (combi) + TC. Stop smoking. Try adalimumab per 5w."



Beloop radiologie: 2014 Toegenomen spondylartrose en discopathie. 2016 flare na staken

Conclusie: AS in remissie onder adalimumab

Gesproken over roken, advies staken Nachtrust goed, functioneert ook goed in dagelijks leven Adalimumab per 4w nog

Conclusie:

Lichamelijk onderzoek: Handartrose

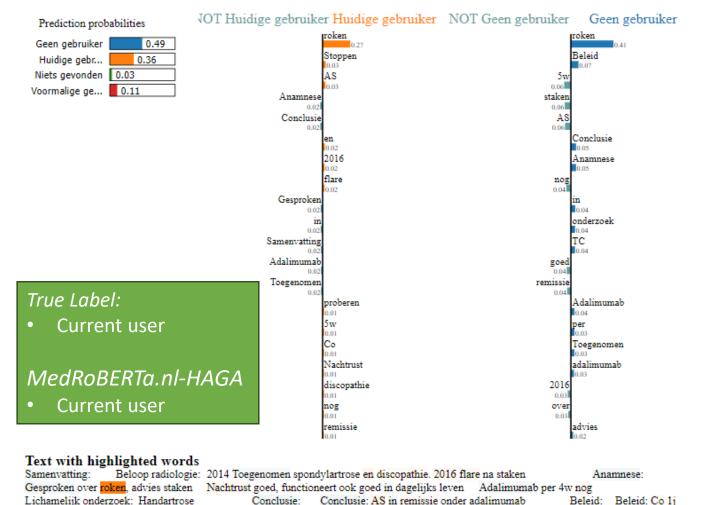
(combi) + TC Stoppen met roken Adalimumab per 5w proberen

Anamnese:

Beleid: Beleid: Co 1i

LIME Model Evaluation [5/6] - MedRoBERTa.nl-HAGA

"Summary: Course of radiology: 2014 Increased spondyloarthrosis and disc disease. 2016 flare after discontinuation. Case history: Talked about smoking, advice to stop. Sleeps well, also functions well in daily life. Adalimumab still per 4w. Physical examination: Hand osteoarthritis. Conclusion: Conclusion: AS in remission under adalimuma.b Policy: Policy: Co 1j (combi) + TC. Stop smoking. Try adalimumab per 5w."

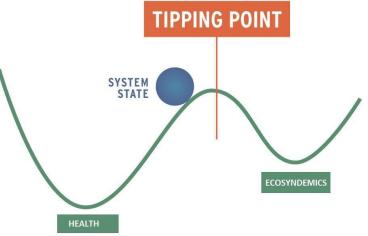


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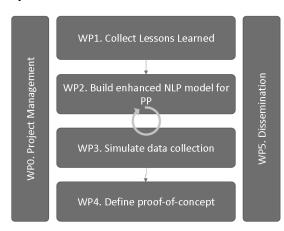
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Model Deployment [6/6]

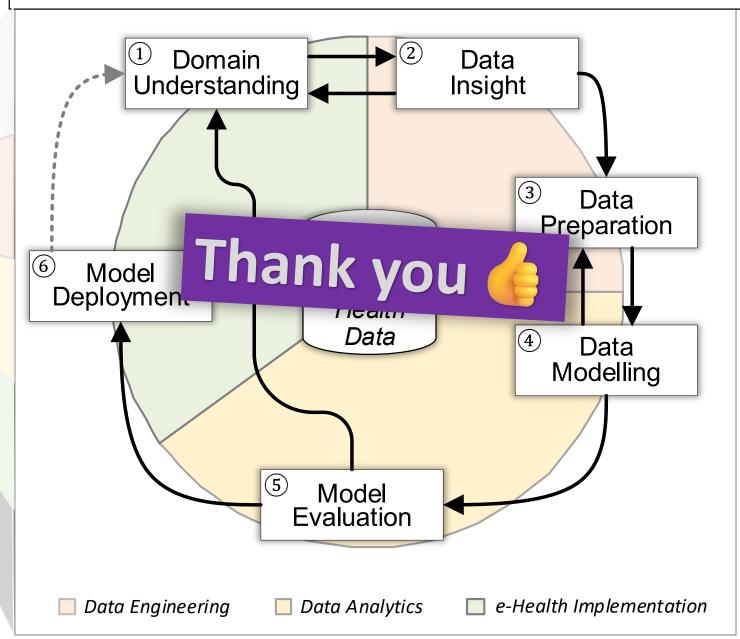
Follow-up research grant for NWA ECOTIP: Identifying tipping points of the effects of living environments on ecosyndemics of lifestyle-related illnesses

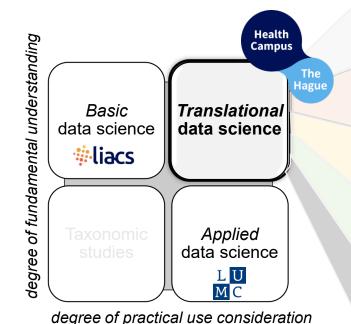


Follow-up research plan for NWO Pandemic Preparedness



Translational Data Science in Population Health







Spruit, Marco. (2022). *Translational Data Science in Population Health* (p. 20). Inaugural lecture. Leiden University. https://doi.org/10.5281/zenodo.7665858