

Institute for Health Metrics and Evaluation

Generative Al in Public Health: opportunities and risks in NLP, data analysis, and simulation

Abraham D. Flaxman 8/8/2023

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What is Generative Artificial Intelligence?

I'm not the right person to give a definition, but I think I have to give it a try.

Generative Al refers to the use of artificial intelligence techniques to create or generate new content, such as images, videos, text, music, or any other form of media.

Examples: Llama 2, ChatGPT, Stable Diffusion, StyleGAN



Capabilities of AI

"If a typical person can do a mental task with less than one second of thought, we can probably automate it using AI either now or in the near future"

---Andrew Ng, 2018



Al Opportunities in Public Health

AI can automate public health work that people:

- do not like doing
- cannot do fast enough
- cannot afford to do

Outline

- Natural Language Processing
- AI-Assisted Data Analysis
- Simulation with Generative Alagents



Verbal Autopsy as a Case Study



- The Global Burden of Disease is a study that compares how much health is lost to each of over 200 different conditions, geographically and temporally
- To measure disease burden, we must know who is dying of what where.
- Death certificates exist for less than half of all deaths globally

Verbal Autopsy as a Case Study



SECTION 5 <u>MODULE 1. GENERAL ILLNESS LEADING TO DEATH</u> SPECIFIC QUESTIONS TO ELICIT SYMPTOMS AND SIGNS OF THE LAST ILLNESS			
NO.	QUESTIONS AND FILTERS	CODING CATEGORIES	SKIP
501	Did(NAME) have fever during her last illness?	YES1	
		NO2	502
		DON'T KNOW8	502
501A	How many days/months before her death did the fever start and end?	START mons days	
		END mons days	
		DIED WITH FEVER	
		DON'T KNOW/UNSURE9998	
501B	How was the fever like?	HIGH1	
		MILD2	
		DON'T KNOW/UNSURE	
501C	Was the fever continuous or on and off?	CONTINUOUS1	
		AFTER EVERY 1 - 2 DAYS 2	
		AT NIGHT ONLY	

Brief History of Verbal Autopsy

- Projects in Asia and Africa in the 1950s and 1960s used systematic interviews by physicians to assess causes of death
- Field workers at the Narangwal project in India labeled this technique "verbal autopsy" (VA)
- Subsequently spread and developed, particularly during the 1970s, when WHO suggested ay reporting of health information by people with no medical background
- Remains the best available approach for assessing causes of death in communities in which most deaths occur at home

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Natural Language Processing

- Free-text is somewhat common in global public health data
- Large language models make it easy to try to do something with it

Verbal Autopsy (VA)

- The challenge: find out what people are dying of
- The approach: ask someone who was present during the final illness
- The bottleneck: not many docs who have experience necessary to certify deaths based on verbal autopsy interview and they have better things to do with their time



Computer Certification of Verbal Autopsy

- Gates Foundation funded collection of a large database, suitable for training and validation: Murray et al, <u>Verbal</u> <u>Autopsy Data</u>, 2011.
- Including open-text with personal information redacted: Flaxman, et al. <u>A de-identified database of 11,979 verbal</u> <u>autopsy open-ended responses</u>, 2018.

Example of a verbal autopsy interview

Deceased was 53 Year Old Male, reported to have:

• Asthma

Puffiness of the
Society All Over His
Protructing Relly

Underlying Cause (Gold Standard Level 1): COPD

• Difficulty Breathing (On-and-Off, Worse in Walking Position)

 Free Text: Asthma, Breath, Heart, Lung, Swell, Water

Computer Certification of Verbal Autopsy

Input:

the respondant explained that the decease died on the 3 day after being injured himself with rod which caused his neck to become very stiff and died because of tetanus.

Output:



[Cirrhosis, Epilepsy, Pneumonia, COPD, Acute Myocardial Infarction, Fires, Renal Failure, AIDS, Lung Cancer, Maternal, Drowning, Other Cardiovascular Diseases, Other Non-communicable Diseases, Falls, Road Traffic, Bite of Venomous Animal, Diabetes, Other Infectious Diseases, TB, Suicide, Other Injuries, Cervical Cancer, Stroke, Malaria, Asthma, Colorectal Cancer, Homicide, Diarrhea/Dysentery, Breast Cancer, Leukemia/Lymphomas, Poisonings, Prostate Cancer, Esophageal Cancer, Stomach Cancer, Measles, Other Defined Causes of Child Deaths, Violent Death, Other Digestive Diseases, Encephalitis, Sepsis, Other Cancers, Hemorrhagic fever, Meningitis, Birth asphyxia, Stillbirth, Preterm Delivery, Meningitis/Sepsis, Congenital malformation].

If there is a specific cause of death mentioned, return "Cause of Death: {Cause}: Confidence {X}" where X is a confidence score between 0 and 1. For example, "Cause of Death: Drowning : Confidence 0.9".

If there are symptoms mentioned without a clear cause of death, return "Symptoms noted: {Symptoms}". For example, "Symptoms noted: Chest Pain, Stomach Pain, Vomiting".

If there is no relevant information provided, return "No relevant info : Confidence $\{X\}$ ".

Given the provided context, determine the cause of death.

Context:

This was straightforward to prototype. But I bet we can do better!

See Kocoń et al, <u>ChatGPT:</u> <u>Jack of all trades, master of</u>

none



Figure 8: Correlations between the loss of ChatGPT performance compared to the SOTA method and difficulty of the task. Regression lines are drawn separately for pragmatic and semantic tasks. Each data point represents a single task, with the index from Tab. 1.



Computer Certification of Verbal Autopsy

Lifting the curtain a little --- chatbots are not exactly what I need for this, but they seem close ... are they?

 $p(w_{n+1}|w_1, w_2, \dots, w_n)$

where w_i is a token, not a word. And causes of death are not always tokens...

With some additional affordances (benefits of openness in e.g. Llama 2 vs GPT4)



Additional GenAI opportunities in VA

- Adaptive interviewing, to get better data, and get it faster
- AI systems in collaboration with physicians --- better? opportunity to enhance training?
- Moral Distress of VA work, and emotional support for interviewee and interviewer



Other NLP challenges, and lessnatural languages

- Nurses Notes
- Triage for SMS-based advice during pregnancy
- Billing data and diagnosis codes



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Al-Assisted Data Analysis

Potential for AI-assisted data analysis to broaden participation in technical analyses central to evidence-based public health

Continuing VA example: what do you do with your VA data once you have collected it and mapped it to underlying cause of death?



Al-Assisted Data Analysis

Potential for AI-assisted data analysis to broaden participation in technical analyses central to evidence-based public health

Example 1: counting rows with missing data when considering using complete case analysis or multiple imputation

Example 2: exploring Llama 2 predictions with huggingface transformers library



Al-Assisted Data Analysis

How should we be writing libraries (and documentation) so that AI Assistants will be able to help users to use them?



This reminds me of an old fantasy novel

In "Earthsea" the study of wizardry requires learning the true names of things, in a language that is no longer spoken by most of the inhabitants of the Earthsea world.

Like what it might take to construct useful prompts for GenAI systems a generation from now...

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A WIZARD OF EARTHSEA by URSULA K. LE GUIN





provide flute music to accompany this poem



Creating flute music to match specific lyrics can be a complex process that involves understanding the rhythm, melody, and mood of the piece. While I can't create a precise musical score in text form without knowing more specific details about your musical preferences, I can provide a general guideline that might help a musician craft something to go along with the poem.



Can you compose a short tune (say four to eight bars) using ABC notation

Rubik's Cube Master



a cheerful, uplifting tune that could work as a motif for your poem. It's

in 4/4 time and the key of C major.

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: Trajectory for simulant 5733: a 65.5 year-old Female

Simulation based on GBD is my main focus covered (pdc) 3.8 0.5 2 disease events dr visits double Δ single none 2026 2027 2028 2029 2030 2021 2022 2023 2024 2025 2031 Treated LDL-C CVD event visit LP LD statin HP LD statin Fibrates Untreated LDL-C in single pill 29 Ezetimibe HP HD statin Maintenance visit not in single pill Background visit LDL-C measurement Target LP HD statin

proportion da

LDL-C mmol/L

In work on COVID modeling, I have become convinced that modelin both "mass behavior" and "elite behavior" are an important area for future innovation



Simulation with Generative AI Agents

From Williams et al, Epidemic Modeling with Generative Agents



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Simulation with Generative AI Agents

Verbal Autopsy example continues --- agents can experience symptoms based on disease, talk about them with others, remember learning about experiences of confidantes, etc

Results of a simulation like this could include some quantification of how much evidence a VA surveillance system is likely to provide. Could even include "elite decision makers" in sim to respond to VA-based evidence (?)



Simulation with Generative AI Agents

What to teach epistudent this fall in Introduction to Epidemic Modeling??

Which also goes back to my interest in AI-Assisted programming...



AI Opportunities and Threats

- Natural Language Processing
- AI-Assisted Data Analysis
- Simulation with Generative Alagents

Bake in biases: e.g. LLM approach to VA might rely on inaccurate stereotypes of disease presentation

Lack of understanding: e.g. analyst being assisted does not know what they have done; or AI is too opaque and it is not clear to where simulation results are generalizable.

Thank you!

Abraham D. Flaxman

abie@uw.edu

