

MCNER Webinar Series



Applications of Artificial Intelligence in Healthcare and Nutrition Research

Wednesday, February 26, 2025



Moderator:
Lisa Diewald, MS, RDN, LDN
Associate Director
MacDonald Center for Nutrition Education and Research
Villanova University M. Louise Fitzpatrick College of Nursing

Today's webinar objectives



- Learn key applications of artificial intelligence in obesity and nutrition research.
- Gain an understanding of the main challenges and ethical considerations associated with AI in health research.
- Learn the importance of leveraging interdisciplinary collaboration to advance AI-driven healthcare solutions.

Finding Slides for Today's Webinar



- Slides are posted at villanova.edu/cope
- From right menu→ Webinars
- Go to 2/26/25 webinar presented by Mengmeng Ji, PhD, MBBS

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- Villanova University M. Louise Fitzpatrick College of Nursing is accredited as a provider of nursing continuing professional development by the American Nurses Credentialing Center's Commission on Accreditation. This activity awards 1 contact hour for nursing professionals.
- This activity awards 1 CPEU in accordance with the Commission on Dietetic Registration's CPEU Prior Approval Program
 - Level 2 activity
 - Suggested CDR Performance Indicators: 7.3.1, 9.1.1, 10.4.6, 13.3.3
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- Questions are welcome!
- Please send through the Q&A Box during the presentation.
- Q&A session will follow the program.

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Introducing our Speaker

Applications of Artificial Intelligence in Healthcare and Nutrition Research



Mengmeng Ji, PhD, MBBS
Faculty
Division of Public Health Sciences
Washington University School of
Medicine at St. Louis



Applications of Artificial Intelligence in Healthcare and Nutrition Research

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February 26, 2025



Outline



- Introduction to AI, machine learning, and deep learning
- Applications of AI in healthcare and nutrition research
- Case Study: automated machine learning and diabetes
- Challenges and ethical considerations
- Q&A

What is Artificial Intelligence (AI)?

Artificial Intelligence: The theory and development of computer systems capable of performing tasks that historically required human intelligence.

“The science and engineering of making intelligent machines, especially intelligent computer programs”.
(John McCarthy, 1956)

White House. Executive Order on the Safe, Secure, and Trustworthy Development and Use of Artificial Intelligence. <https://www.federalregister.gov/documents/2023/11/01/2023-24283/safe-secure-and-trustworthy-development-and-use-of-artificial-intelligence>

The history of AI

1940s-1950s

Foundations of AI

In the 1940s, the first artificial neurons were conceptualised. The 1950s introduced us to the Turing Test and the term “Artificial Intelligence”.



1960s-1970s

Early Development

The 60s and 70s brought the birth of ELIZA, simulating human conversation, and Dendral, the first expert system, showcasing the early potentials of AI.



1980s

AI Winter & Expert Systems

The 80s faced reduced AI funding but saw the inaugural National Conference on AI. The backpropagation concept rejuvenated neural networks.



1990s

Revival & Emergence of ML

The 90s witnessed IBM's Deep Blue defeating chess champion Garry Kasparov and the inception of the LOOM project, laying the foundations for GenAI.



2000s

The Genesis of Generative AI

Geoffrey Hinton propelled deep learning into the limelight, steering AI toward relentless growth and innovation.



2010s

Rise of AI

In 2011, IBM Watson won “Jeopardy!”, highlighting AI’s language skills. The 2010s marked major AI milestones, including pioneering work in image recognition and the birth of GANs in 2014, followed by OpenAI’s founding in 2015.



2020s

GenAI Reaches New Horizons

At the start of this decade, we've seen significant strides in GenAI, notably with OpenAI's GPT-3 and DALL-E. 2023 welcomed advanced tools like ChatGPT-4 and Google's Bard, alongside Microsoft's Bing AI, enhancing accessibility and reliability of information.



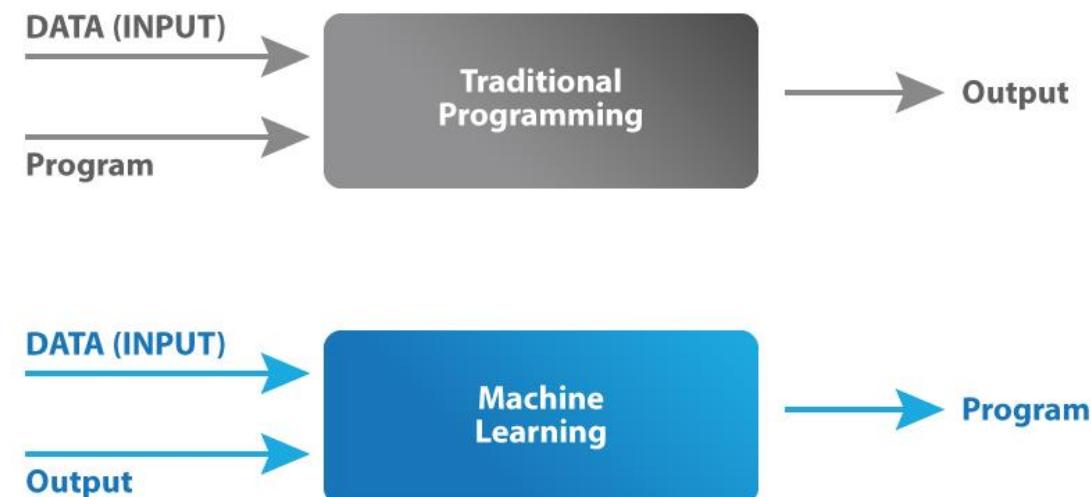


What is Machine Learning (ML)?

Machine Learning: The practice of learning from examples seen in data without being explicitly programmed.

“The scientific discipline that explores the construction and study of algorithms that can learn from data.”
(wikipedia)

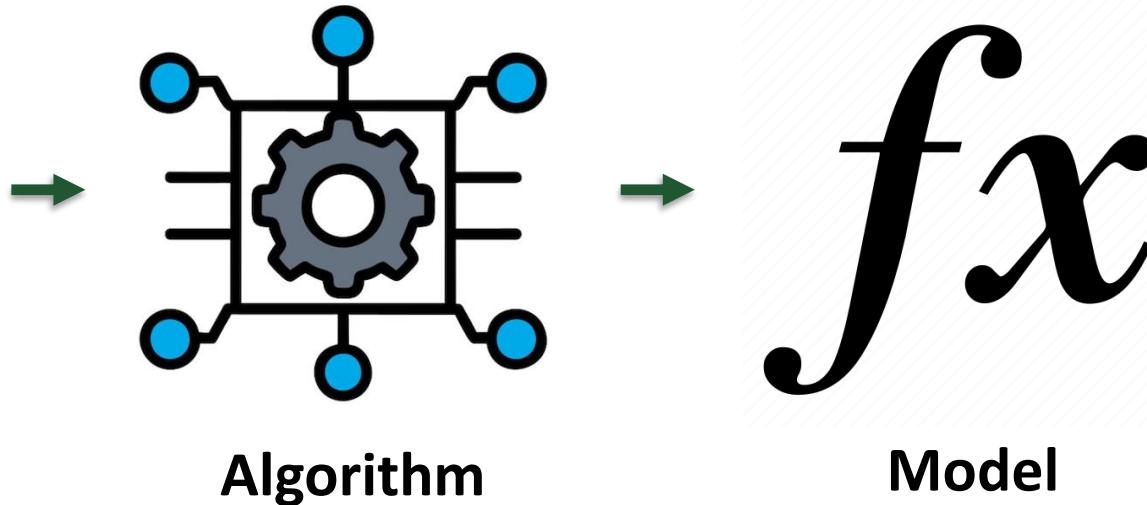
“The science of getting computers to act without being explicitly programmed.”
(Arthur Lee Samuel)



Machine Learning as a data driven approach



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- Acquire the knowledge automatically from data
- Discover patterns in complex data
- Generate predictions based on previous experience

Machine Learning vs Statistical Models



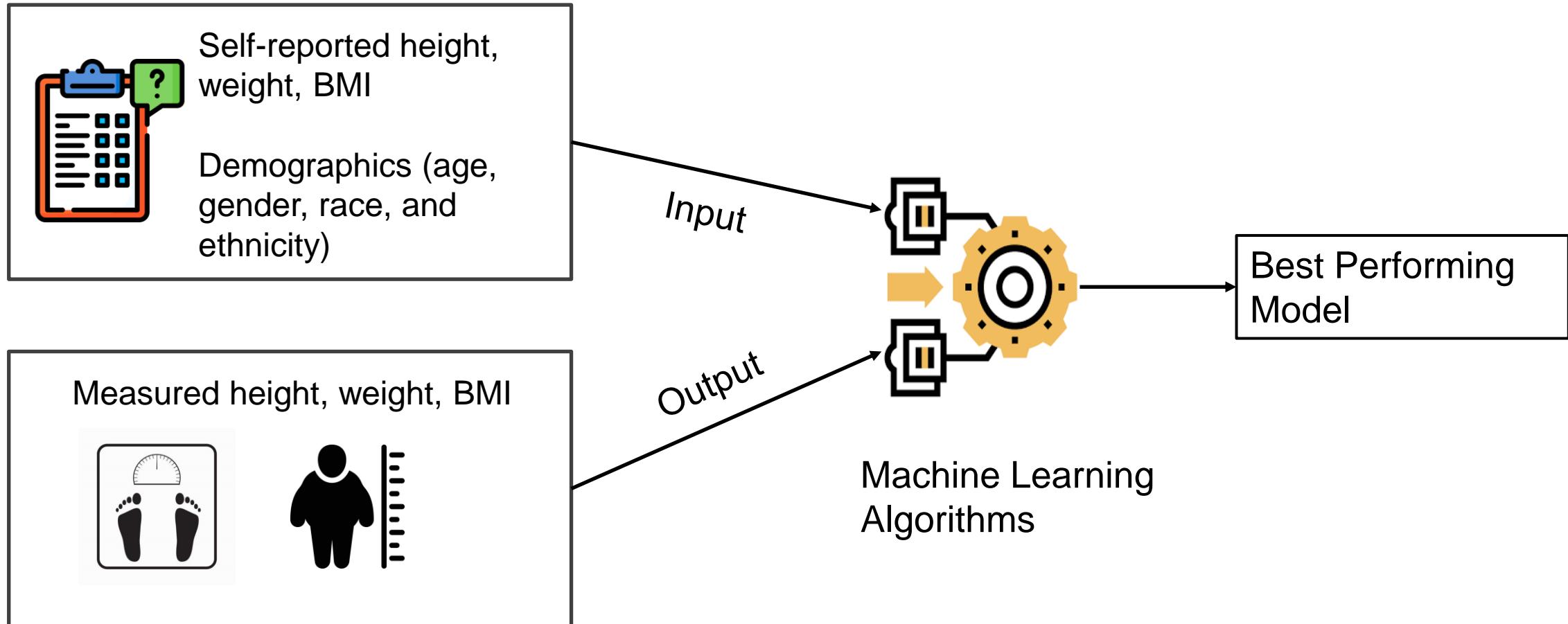
Machine Learning

- prediction and pattern recognition
- Complex, capturing non-linear relationships
- Makes much fewer a priori assumptions about the data distribution
- Lower interpretability, “black box”.

Statistical Models

- inference and explanation
- Simpler, parametric models with a predefined structure
- Relies on specific assumptions to ensure model validity.
- Offers greater interpretability

Machine Learning project example

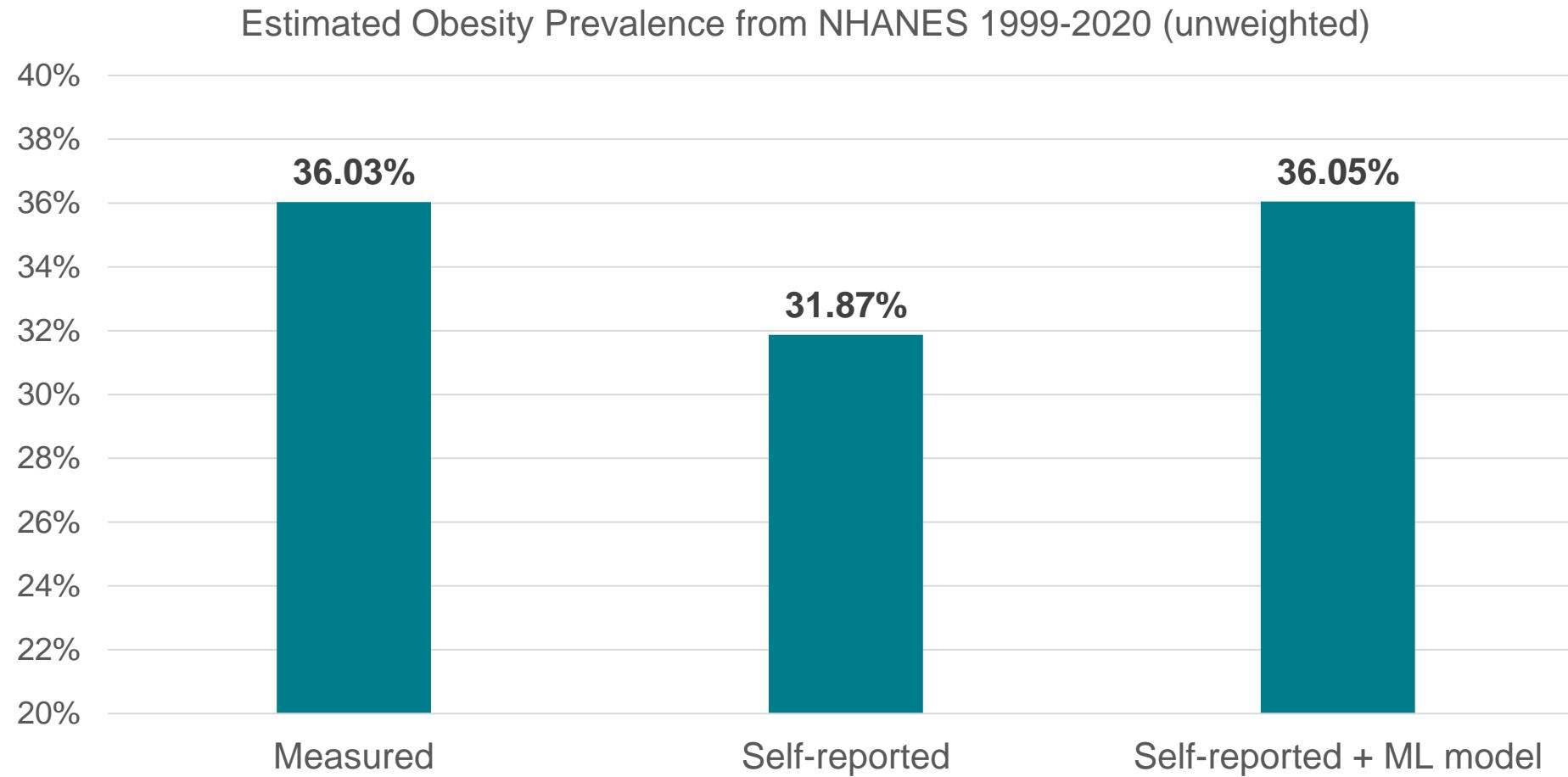


Data Source: NHANES 1999-2020

An R & Ji M. J Public Health Manag Pract. 2023;29(5):671-674.



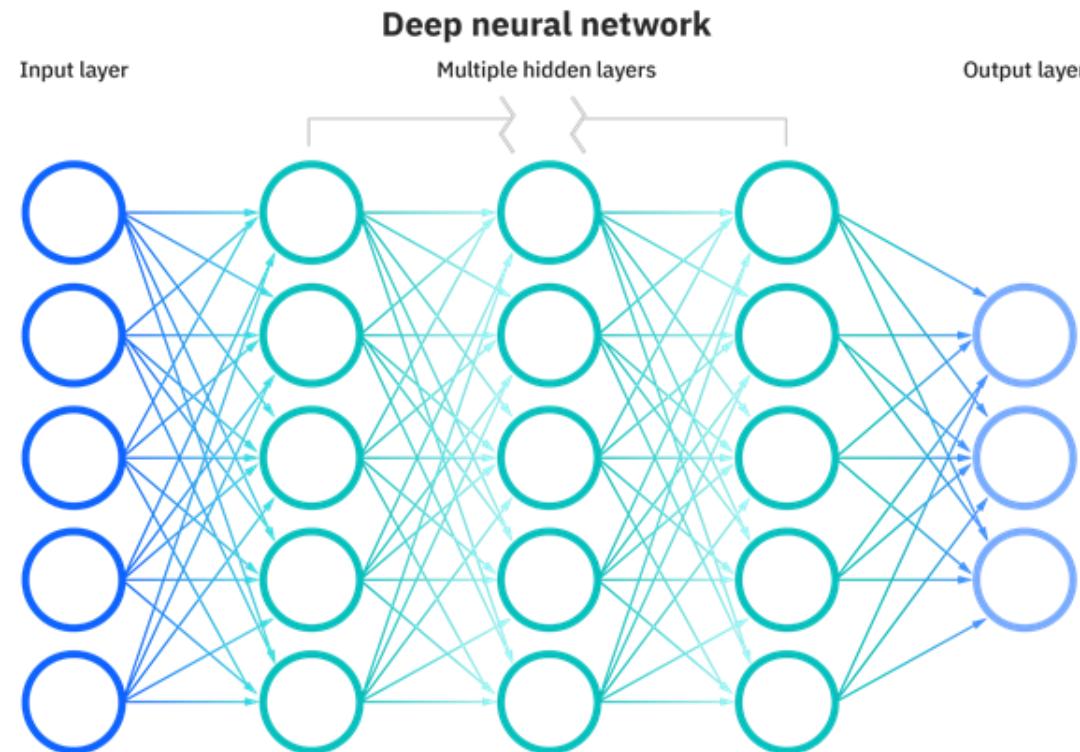
ML project example: model performance



What is Deep Learning?

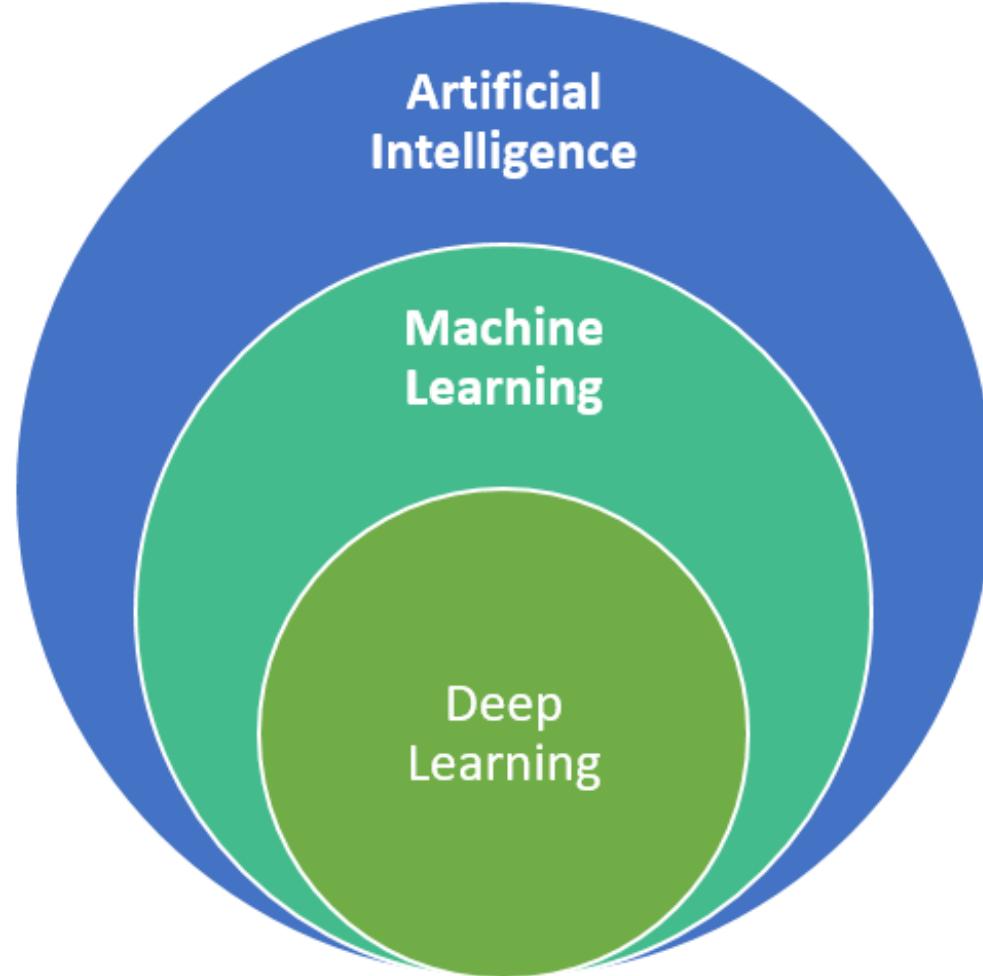


- Deep learning (DL) is a type of machine learning that uses artificial neural networks to learn from data.

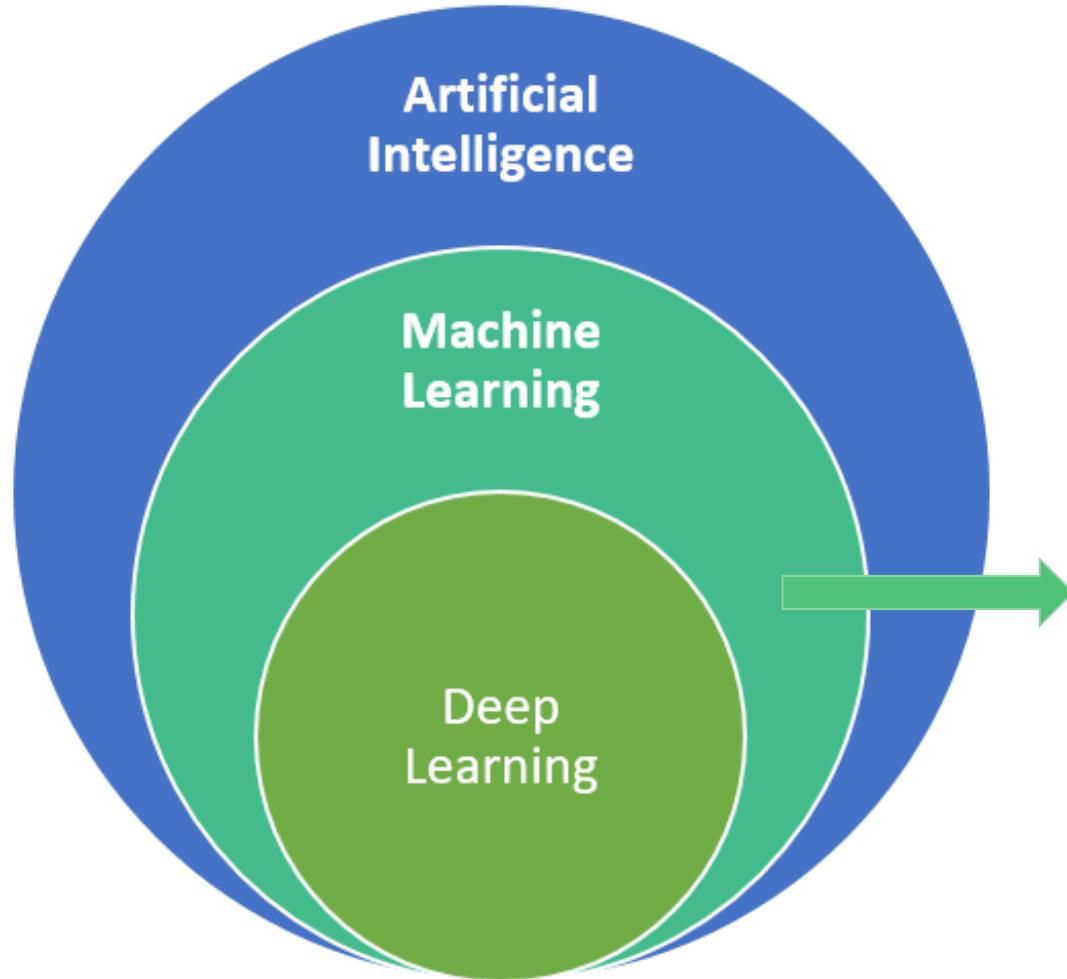


- image recognition
- natural language processing
- speech recognition
- ...

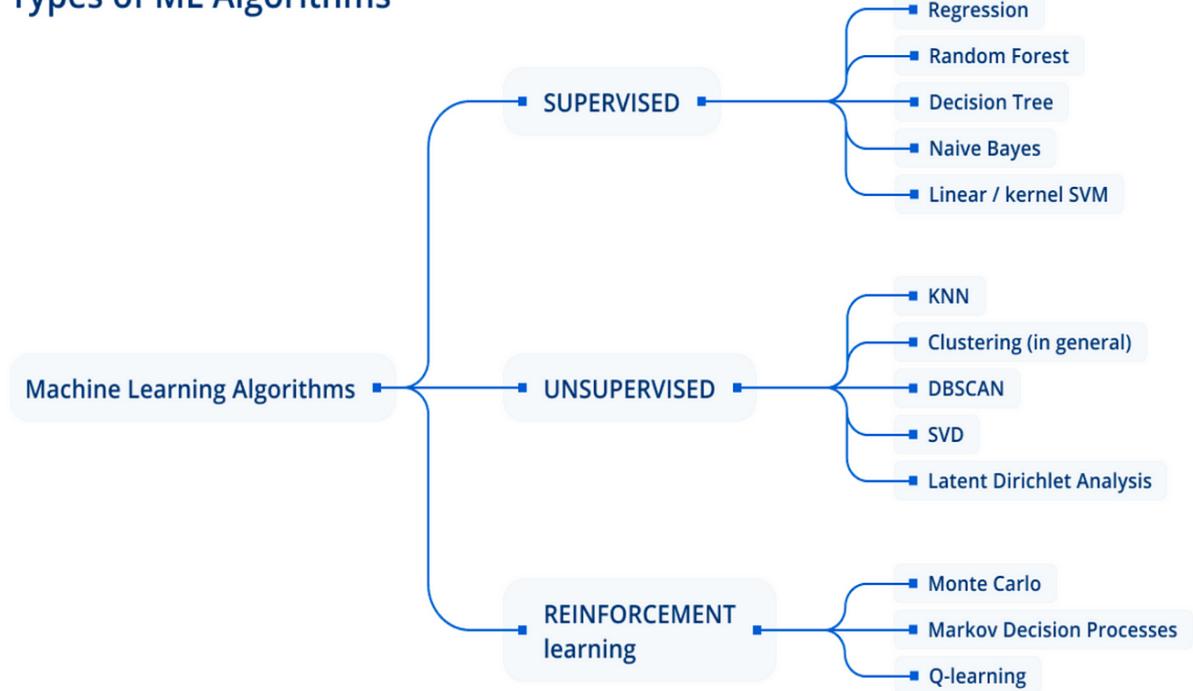
The relationship between AI, ML and DL



The relationship between AI, ML and DL

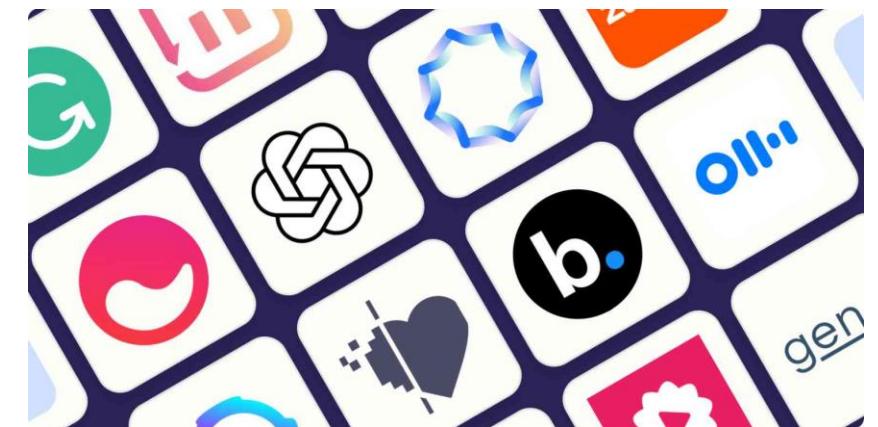
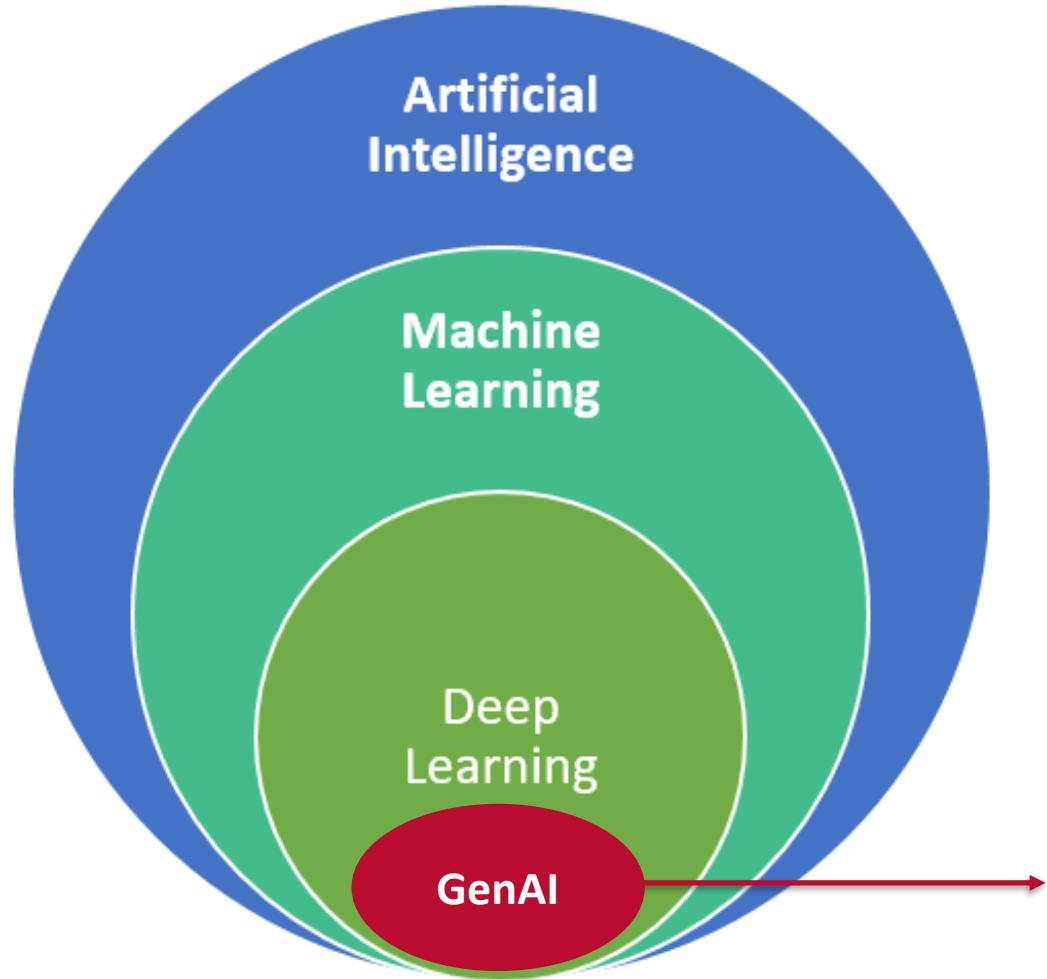


Types of ML Algorithms





The relationship between AI, ML and DL



Generative AI



AI in healthcare: medical imaging

- Among the most promising clinical applications of AI is diagnostic imaging.
- Over 70% of FDA-approved AI medical devices are in radiology.



Health

AI is already outperforming human doctors, with more advancements expected

Artificial intelligence is being developed to analyse images from a host of medical tests such as retina scans and mammograms

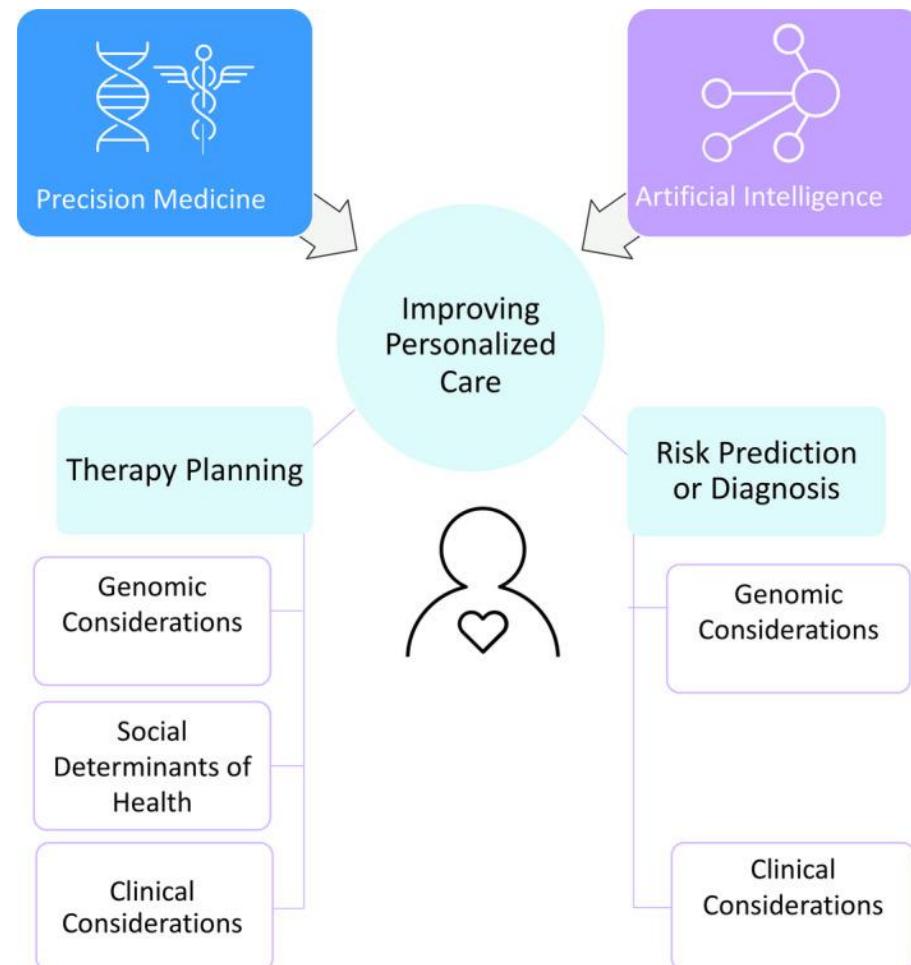


Oren et al., The Lancet Digital Health, 2020, 2(9), e486 - e488

AI in healthcare: personalized medicine



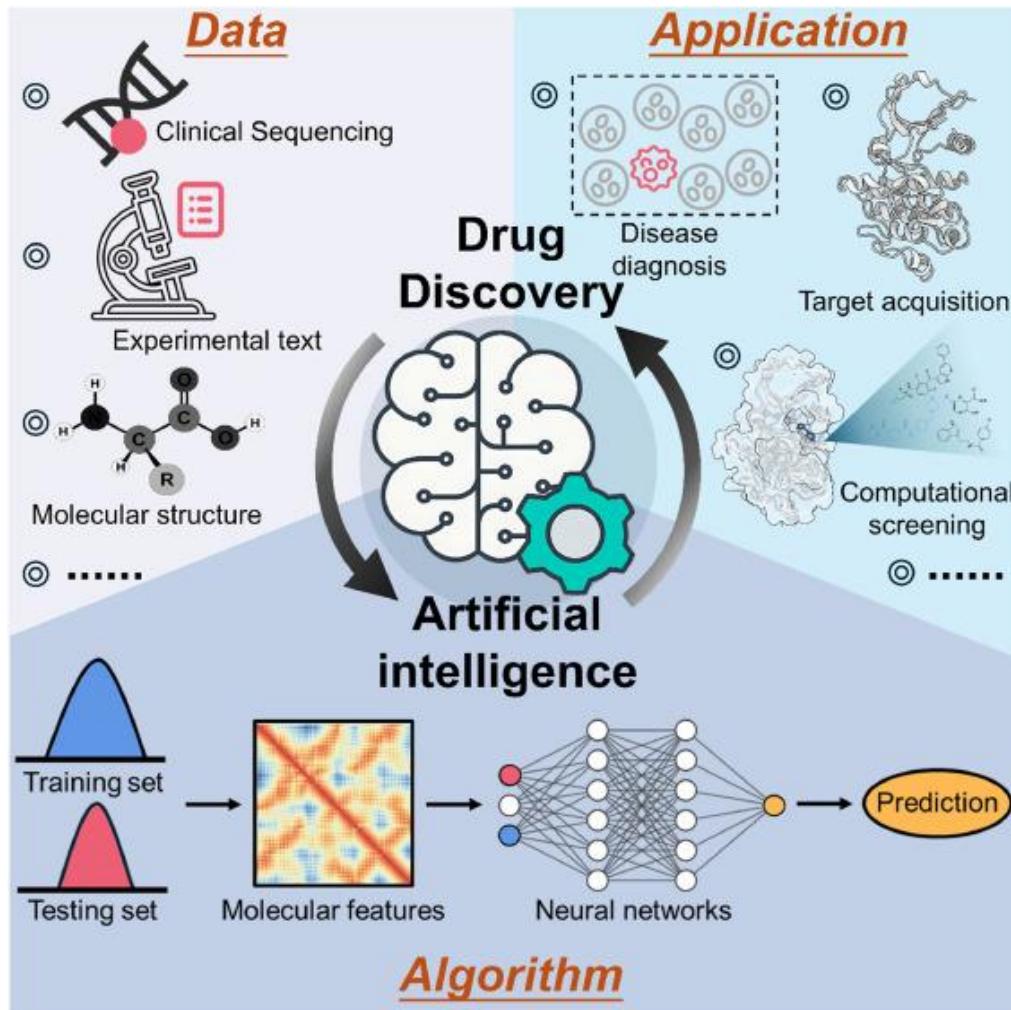
- AI can process vast amounts of patient data (genomics, medical records, etc.) to recommend custom treatment plans.
 - Multi-omics data integration
 - Tailored disease risk assessment
 - Precision treatment and drug Selection
 - Real-time patient monitoring and Feedback



AI in healthcare: drug discovery



- AI has emerged as a transformative force in drug discovery, revolutionizing every stage of the process.
 - Target identification
 - Protein structure prediction
 - Accelerate clinical trials
 - Data integration
 - Post-marketing surveillance



Other applications of AI in healthcare



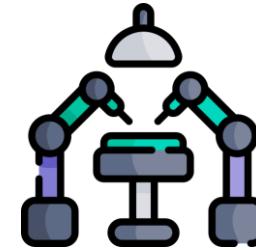
Administrative Automation



Patient Support



Predictive Analytics



Robotic Surgery



The applications of AI in nutrition research

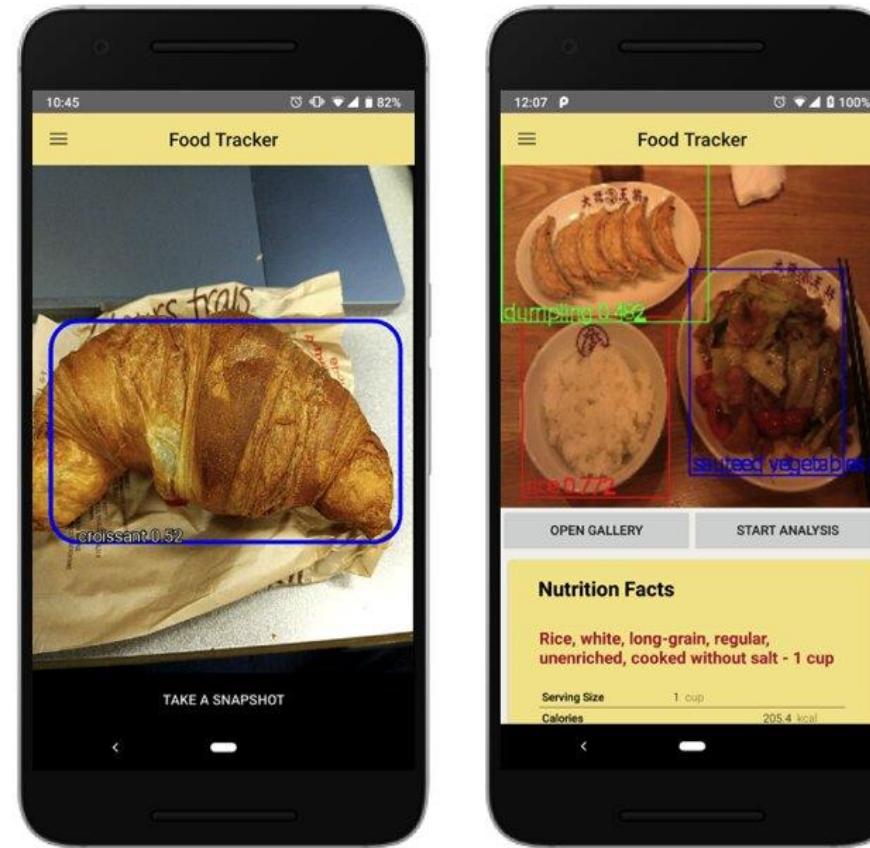
- AI is evolving quickly and has the potential to revolutionize the nutrition field.
 - Dietary assessment
 - Lifestyle intervention
 - Obesity management
 - Nutrition epidemiology
 - Food estimation
 - Malnutrition prediction



Dietary Assessment and Food Recognition



- AI can estimate calories in food using tools like apps and online tools that analyze images of food.



Dietary Assessment and Food Recognition



- AI can estimate calories in food using tools like apps and online tools that analyze images of food.
- Speech recognition as new way for dietary assessment.

JOURNAL OF MEDICAL INTERNET RESEARCH

Taylor et al

Original Paper

Use of Natural Spoken Language With Automated Mapping of Self-reported Food Intake to Food Composition Data for Low-Burden Real-time Dietary Assessment: Method Comparison Study

Salima Taylor^{1*}, MS; Mandy Korpusik^{2*}, PhD; Sai Das¹, PhD; Cheryl Gilhooly¹, PhD; Ryan Simpson³, MS; James Glass², PhD; Susan Roberts¹, PhD

¹Jean Mayer United States Department of Agriculture Human Nutrition Research Center on Aging, Tufts University, Boston, MA, United States

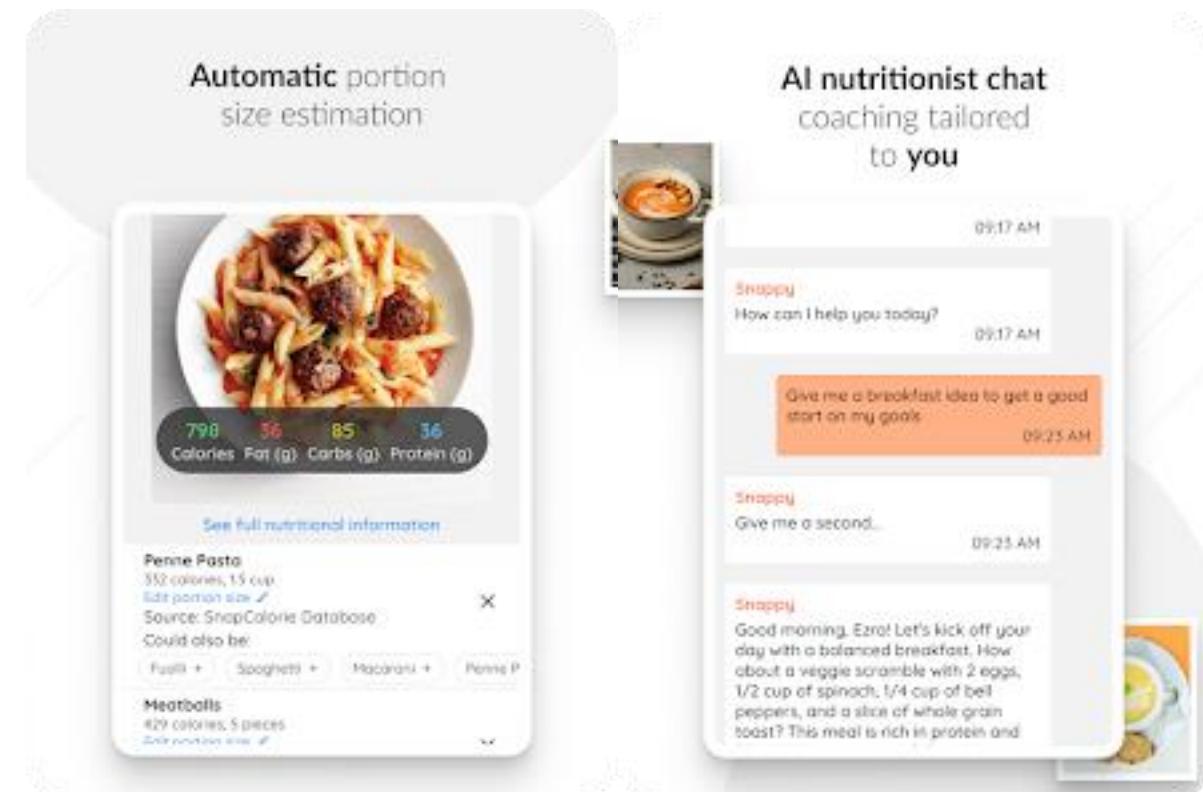
²Computer Science and Artificial Intelligence Laboratory, Massachusetts Institute of Technology, Cambridge, MA, United States

The 24 h recall is considered to be the gold standard for food intake assessment; the AI-based app was as accurate as the gold standard itself, showing no significant difference between the two tools.

Personalized Diet Recommendations



- AI can gather various data sources, such as genetic information, lifestyle choices, and health metrics, to offer nutrition advice that is specifically designed for each individual.
- Integration with wearables to provide real-time dietary recommendations.



APP: SnapCalorie AI Calorie Counter

Nutritional Deficiency Detection and Monitoring



- AI integrates clinical data (blood tests, symptom records, body composition) to assess the likelihood of deficiencies in vitamins, minerals, and other nutrients.

RESEARCH ARTICLE

Predicting malnutrition from longitudinal patient trajectories with deep learning

Boyang Tom Jin¹, Mi Hyun Choi², Meagan F. Moyer¹, David A. Kim^{1,4*}

1 Department of Computer Science, Stanford University, Stanford, California, United States of America,

2 Department of Bioengineering, Stanford University, Stanford, California, United States of America,

3 Department of Digital Health Care Integration, Stanford Health Care, Stanford, California, United States of America, **4** Department of Emergency Medicine, Stanford University, Stanford, California, United States of America

* davidak@stanford.edu

Deep learning models can reliably predict malnutrition from existing longitudinal patient records, with better predictive performance and lower data-collection requirements than existing instruments.

Jin et al., PLoS One. 2022;17(7):e0271487.

Lifestyle interventions



- AI has demonstrated encouraging outcomes in increasing adherence to a healthier diet and physical activity.

JMIR MHEALTH AND UHEALTH

Maher et al

Original Paper

A Physical Activity and Diet Program Delivered by Artificially Intelligent Virtual Health Coach: Proof-of-Concept Study

Carol Ann Maher¹, PhD; Courtney Rose Davis¹, PhD; Rachel Grace Curtis¹, PhD; Camille Elizabeth Short², PhD; Karen Joy Murphy³, PhD

¹Alliance for Research in Exercise, Nutrition and Activity, Allied Health and Human Performance, University of South Australia, Adelaide, Australia

²Melbourne Centre for Behaviour Change, School of Psychological Sciences and School of Health Sciences, University of Melbourne, Melbourne, Australia

³Alliance for Research in Exercise, Nutrition and Activity, Clinical and Health Sciences, University of South Australia, Adelaide, Australia

An artificially intelligent virtual assistant-led lifestyle-modification intervention was feasible and achieved measurable improvements in physical activity, diet, and body composition at 12 weeks.

Maher et al, JMIR Mhealth Uhealth. 2020;8:e17558.

Implications for Nutrition Research

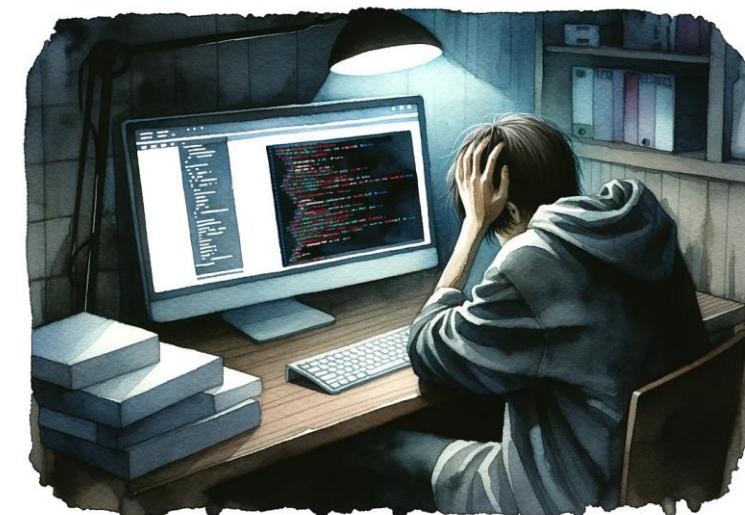


- AI can transform the accuracy of results by eliminating biases from self-reported data.
- AI can make nutrition information and recommendations available to a greater extent.
- AI can increase dietary and physical activity adherence.
- AI enables real-time monitoring and timely feedback and interventions.
- AI has the potential to analyze genetics, culture, lifestyle, and health conditions at once, and enable precision nutrition.

Barriers to use ML/DL for nutrition research



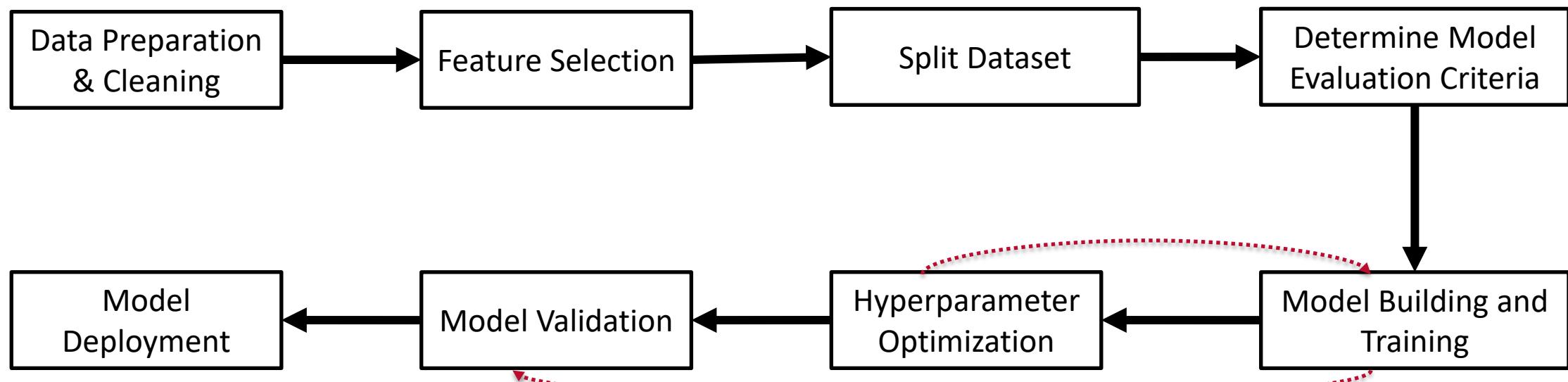
- As AI becomes more implemented into health research, there are also notable challenges, particularly the accessibility of AI and the data science involved.
- Automated machine learning (AutoML) could simplify AI development for both experts and non-experts.



Machine learning pipeline



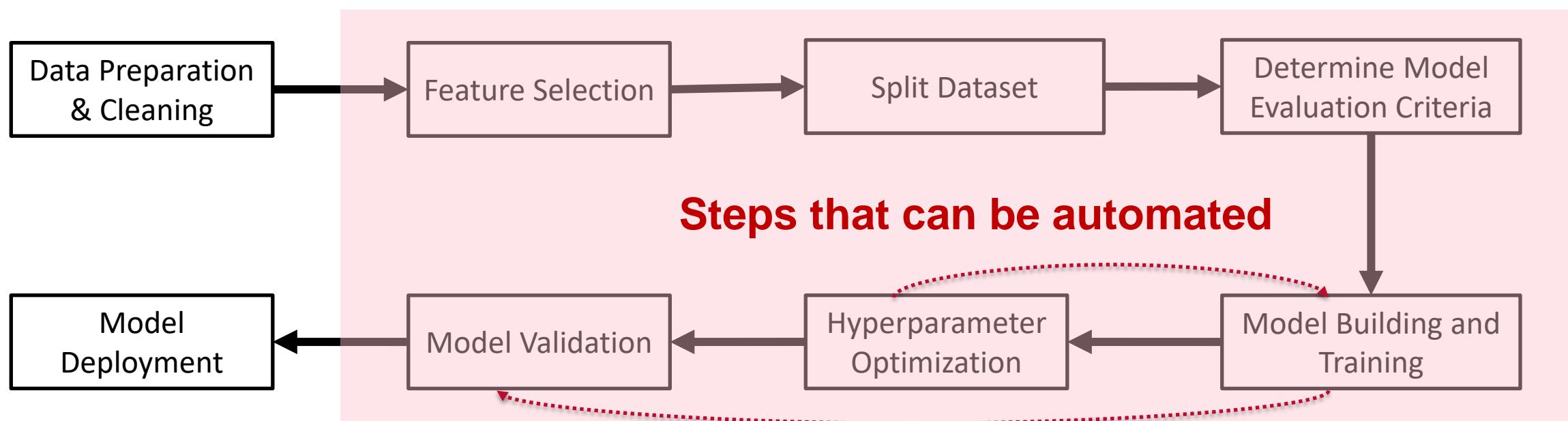
- The steps for machine learning include data collection, preparation, model selection, training, evaluation, tuning, deployment, and prediction.



Automated machine learning



- Automated machine learning (AutoML) is the process of automating the tasks of applying machine learning to real-world problems. It is the combination of automation and machine learning.





Case Study

**Use of Automated Machine Learning to Detect
Undiagnosed Diabetes in U.S. Adults**

Diabetes screening



- In the US, an estimated 37 million adults aged 18 years or older have diabetes, with 23% of them remaining undiagnosed.
- Screening relies on blood testing is not widely followed, especially among individuals with lower education, income, and limited healthcare access.
- We aim to develop an AI-powered screening tool for undiagnosed diabetes, utilizing self-reported anthropometric measures and health behavior data.

Data and participants



- National Health and Nutrition Examination Survey (NHANES) 1999-2020

NHANES 1999-2020
N = 112,502



9,559 without diabetes

A1C <6.5% **AND**
FPG < 126 mg/dL **AND**
2-h PG < 200 mg/dL

Exclusion Criteria:

- Diagnosed diabetes
- Missing self-reported diabetes status
- Missing lab results for diabetes diagnosis
- Age < 20 years
- Pregnant females

2,256 with undiagnosed diabetes

A1C \geq 6.5% **OR**
FPG \geq 126 mg/dL **OR**
2-h PG \geq 200 mg/dL

Study design: H2O AutoML



- H2O AutoML was used for automating the machine learning workflow, which includes automatic training and tuning of many models within a user-specified time-limit.

```
# Load your dataset
data = h2o.import_file("your_data.csv")

# Define target and predictors
y = "target"                      # Replace with your target column name
x = data.columns
x.remove(y)

# Split the dataset into training and test sets
train, test = data.split_frame(ratios=[0.8], seed=42)

# Run AutoML for a maximum of 60 seconds
aml = H2OAutoML(max_runtime_secs=60, seed=42)
aml.train(x=x, y=y, training_frame=train)
```

Feature selection



Category	Features (n=28)
Demographics	age, gender, race/ethnicity, education, marital status, income
Anthropometric Measures	weight, height, BMI, waist circumference
Family History	family history of diabetes
Lifestyle Factors	smoking, drinking, physical activity, frequency of dining out per week
Dietary Intake	daily total energy, daily total fat, daily total cholesterol, daily total sugar
Chronic Conditions	hypertension, myocardial infarction, coronary heart disease, weak/failing kidney, stroke, congestive heart failure, liver disease, cancer, arthritis

Performance of AutoML Model



Undiagnosed Diabetes (≥ 1 test+) vs No Diabetes ^a										
	Automated ML		Logistic Regression		Support Vector Machines		Random Forest		XGBoost	
	Train	Test	Train	Test	Train	Test	Train	Test	Train	Test
Accuracy	89.5%	86.5%	72.4%	71.6%	79.6%	73.3%	96.5%	80.9%	98.3%	80.6%
AUC	0.940	0.909	0.800	0.801	0.879	0.766	0.991	0.795	0.998	0.776

^a "No diabetes" group includes normoglycemia and prediabetes.

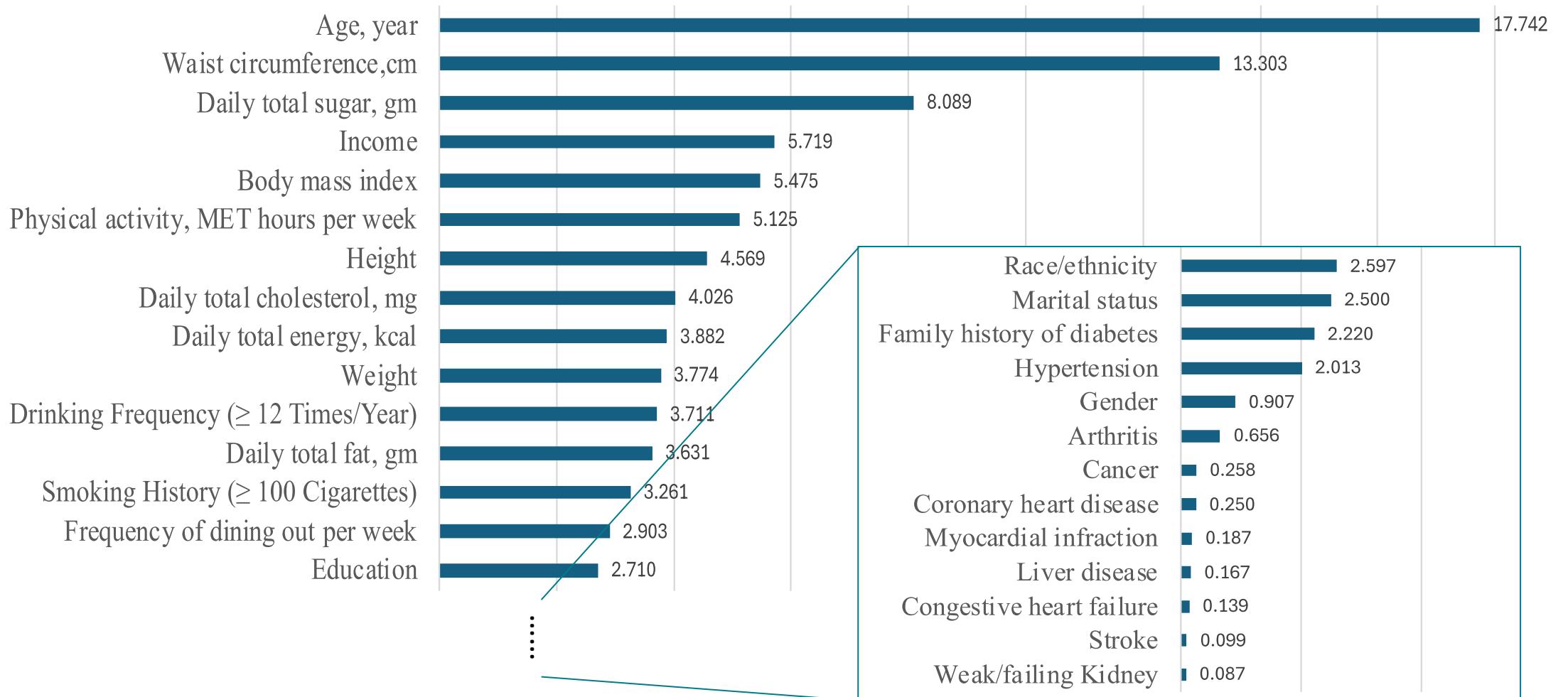
Sensitivity: 70.3%
Specificity: 90.5%

Positive predictive value: 64.1%
Negative predictive value: 92.6%

Model interpretation



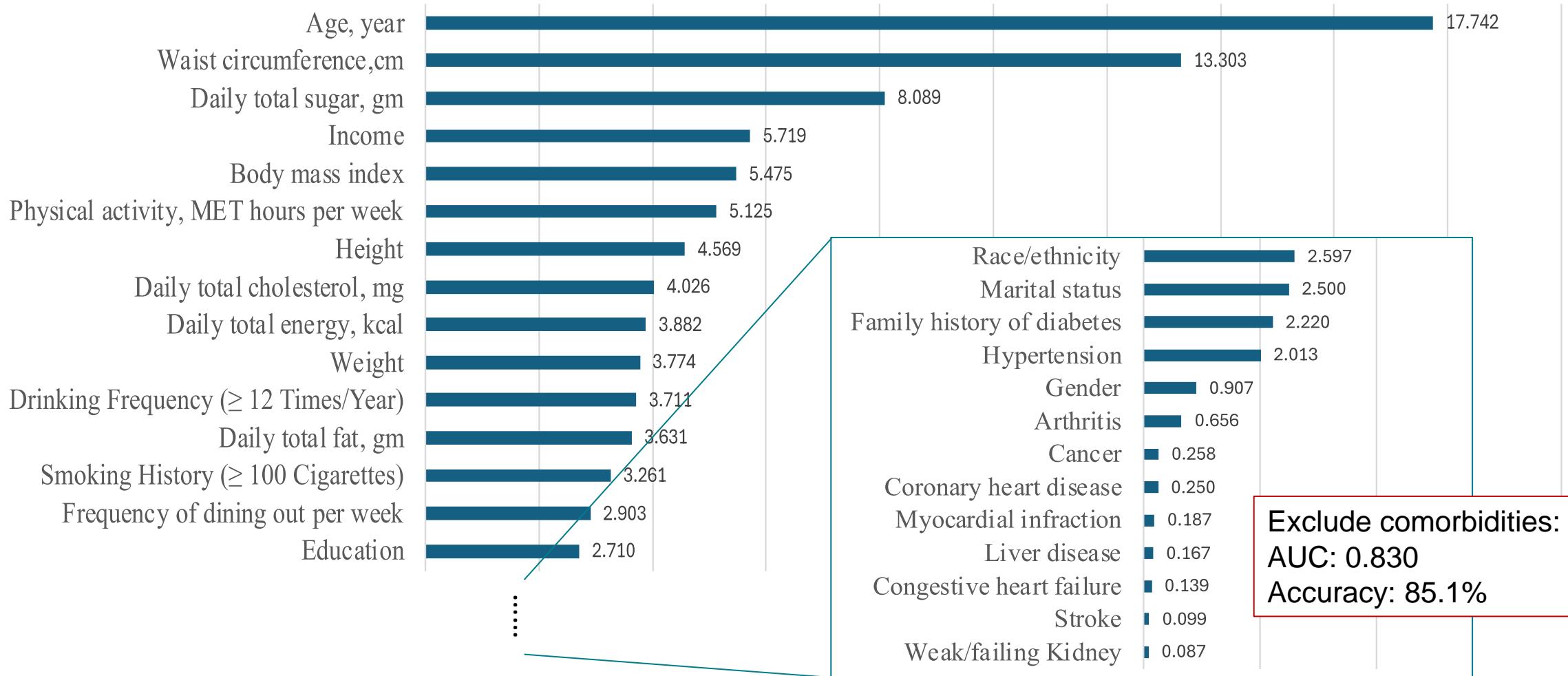
Figure: Scaled Feature Importance (% of Total)



Model interpretation



Figure: Scaled Feature Importance (% of Total)



Exclude comorbidities:
AUC: 0.830
Accuracy: 85.1%

Discussion of case study

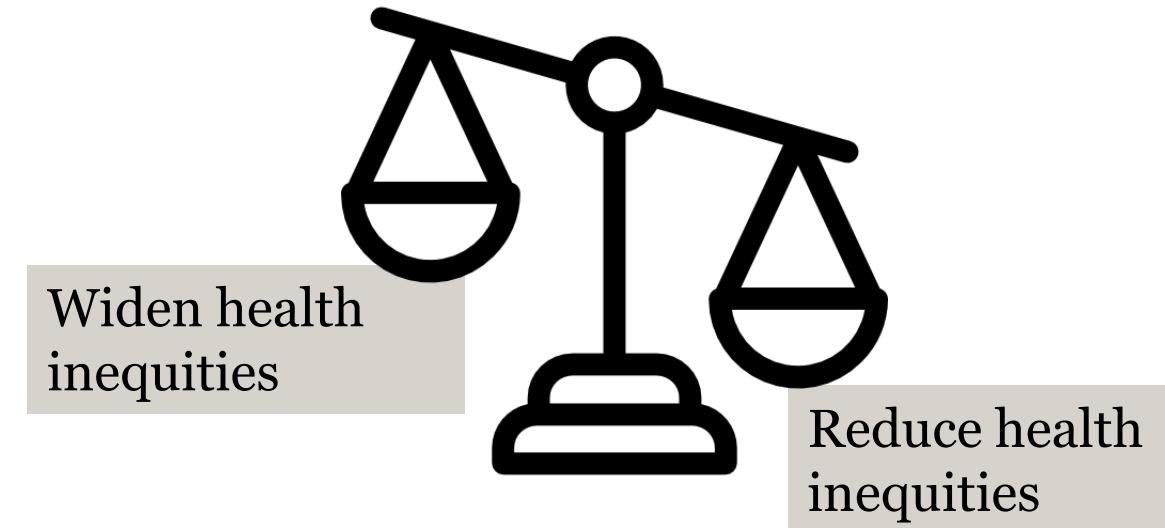


- This study showed that AutoML models outperformed traditional machine learning models in detecting undiagnosed diabetes.
- AutoML streamlines complex machine learning processes, enabling health researchers to leverage advanced technology without requiring extensive ML expertise.
- Given its high accuracy and adaptability to the broader U.S. population, this model has potential to support large-scale diabetes screening and early intervention strategies.

AI's potential for public health comes with risks and challenges



- Data quality and availability
- Bias and fairness
- Transparency and explainability
- Accountability and liability
- Privacy and security risks
- Ethical and legal considerations



Key takeaways



AI adds to traditional statistical modeling by capturing complex non-linear relationships and enhancing predictive accuracy.

AI broadens health research by enabling the analysis of multi-modal data, including text, images, audio, video, and other forms of digital content.

AI is becoming increasingly accessible to health researchers.

Addressing bias and ensuring fairness in AI algorithms is a critical ethical concern.



**Thank you!
Questions?**

Contact:

Mengmeng Ji, PhD, MBBS
j.mengmeng@wustl.edu

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MCNER Webinar Series for Health Professionals

Food Marketing and Children: Impact and New Directions

Jennifer L. Pomeranz, JD, MPH



Save the Date!

**Wednesday, 4/30/25
12-1 PM EST**

Registration Opens Soon!

Q&A



Moderator:
Lisa Diewald, MS, RDN, LDN
mcner@villanova.edu

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