

# Emerging Avenue of Artificial Intelligence and Ethical Considerations

**Bishesh Khanal, PhD**

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<https://bishesh.github.io>

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Symposium on Responsible Conduct of Research in Nepal:  
Identifying and Addressing Gaps

Hotel Himalaya, Lalitpur, Nepal

18 July, 2023



# Outline

Very Brief Intro to NAAMII

Brief Intro to AI

AI avenues in Nepal's Healthcare

Ethical Considerations



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AI avenues in Nepal's Healthcare

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# NAAMII

**Vision: A world class center of excellence to improve the lives of people through scientific research**

A pioneering and leading research institute in Artificial Intelligence (AI) in Nepal

Scientific Research

Education & Outreach

Industry & Innovation



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# Research Groups at NAAMII

Autonomous Research Groups

Theoretical and Applied Research in Applied Maths, Informatics, AI

Diverse Application Domains: Healthcare, Agriculture, Finance, Education, AI-Society ...



# Research Groups at NAAMII

## Autonomous Research Groups

Theoretical and Applied Research in Applied Maths, Informatics, AI

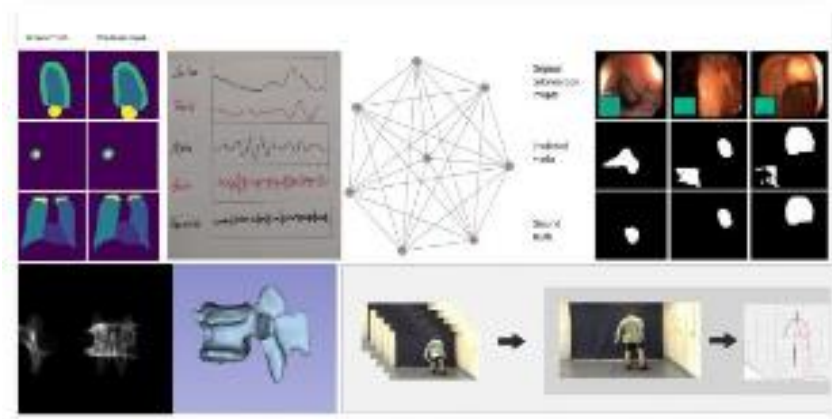
Diverse Application Domains: Healthcare, Agriculture, Finance, Education, AI-Society ...

Published >30 research articles (MICCAI, CVPR, ICML/ICLR workshops, MeDIA, TMI, ...)

Trained > 36 research staffs, supervised 5 Msc theses, 3 undergraduate final year projects, Full-time faculty co-supervising 2 PhD students in Korea and the USA



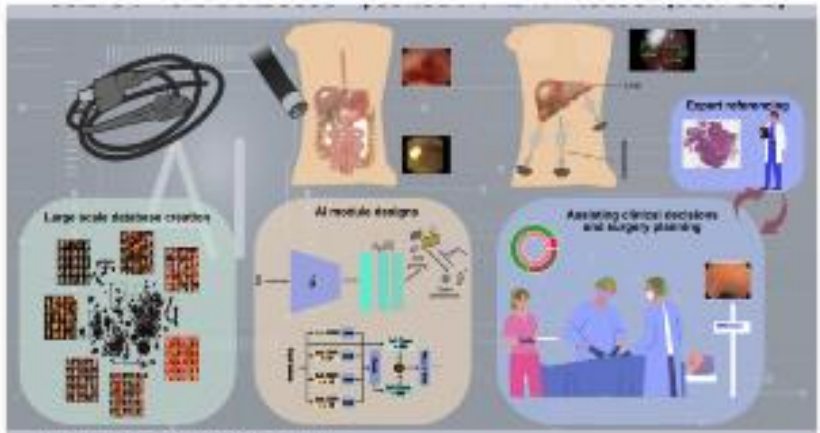
# Research Groups at NAAMII



Transforming Global health with AI (TOGAI)



B Bhattarai MultiModal Learning Lab (MMLL)



Computational Endoscopy, Surgery & Pathology (CESP)



Computational Genomics Lab (CGL)



Research using Artificial Intelligence in Neuroscience (RAIN)



AI & Society



# Research Groups at NAAMII

## TRANSFORMING GLOBAL HEALTH WITH AI (TOGAI)



Led by [Dr. Bishesh Khanal](#),  
Research Scientist at NAAMII

## B BHATTARAI MULTIMODAL LEARNING LAB (MMLL)



Led by [Dr. Binod Bhattarai](#),  
Adj. Research Scientist at NAAMII  
& Assitant Professor at University of Aberdeen

## COMPUTATIONAL GENOMICS LAB



Led by [Dr. Raunak Shrestha](#),  
Adj. Research Scientist at NAAMII  
& Research Scholar at UCSF

## RESEARCH USING ARTIFICIAL INTELLIGENCE IN NEUROSCIENCE (RAIN)



Led by [Dr. Nabin Koirala](#),  
Adj. Research Scientist at NAAMII  
& Associate Research Scientist at Yale University

## COMPUTATION ENDOSCOPY, SURGERY AND PATHOLOGY GROUP



Led by [Dr. Sharib Ali](#),  
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& Research Scholar at UCSF

## Other Adjunct Faculties



[Dr. Ajad Chhatkuli](#),  
Adj. Research Scientist at NAAMII  
& Postdoc Researcher at ETH Zürich



[Shreyasha Paudel](#),  
Research Fellow/Alumni at NAAMII  
& PhD Scholar at University of Toronto



[Dr. Danda Pani Paudel](#),  
Adj. Research Scientist at NAAMII  
& Researcher at Computer Vision Lab, ETH Zü



[Dr. Suman Raj Bista](#),  
Adj. Research Scientist at NAAMII

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& Lecturer at University of Leeds



[Dr. Taman Upadhaya](#),  
Adj. Research Scientist at NAAMII  
& Associate Researcher at UCSF



[Prof. Sabita Maharjan](#),  
Adj. Research Scientist at NAAMII  
& Associate Professor at University of Oslo

Transforming Global Health with AI  
(ToGAI)

IMAGING

NATURAL  
LANGUAGE  
PROCESSING (NLP)

EPIDEMIC AND  
GENOMICS  
INTELLIGENCE

Smartphone

Ultrasound

X-Ray

Conversational AI

Internet Content  
Analytics

Genomics

Epidemiology

Epidemic  
Intelligence

Diarrhea  
Parasite  
Detection

AI-Assisted  
Sonography

Scoliosis

COVID-19 Chatbot

COVID-19 Social  
Listening

Multi-drug  
Resistant  
Tuberculosis

Epidemiology  
of  
SARS-CoV-2

COVID-19  
Epidemiology  
Intelligence

Pesticide  
Detection

2D 3D  
Reconstruction

Intimate Partner  
Violence (IPV)

COVID-19  
Social  
Listening

Cervical  
Cancer VIA  
Screening



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Very Brief Intro to NAAMII

**Brief Intro to AI**

AI avenues in Nepal's Healthcare

Ethical Considerations

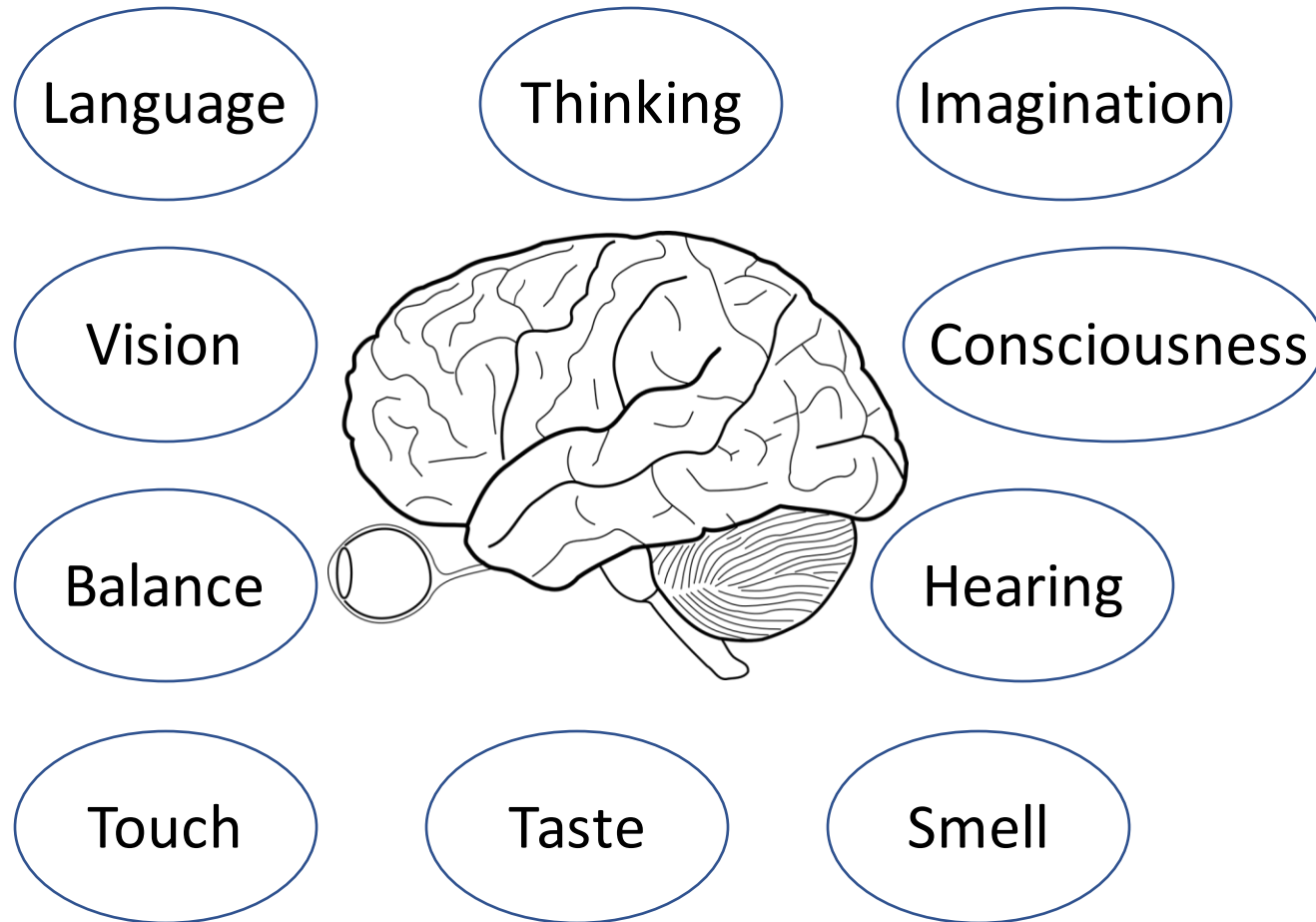


# Artificial Intelligence (AI)

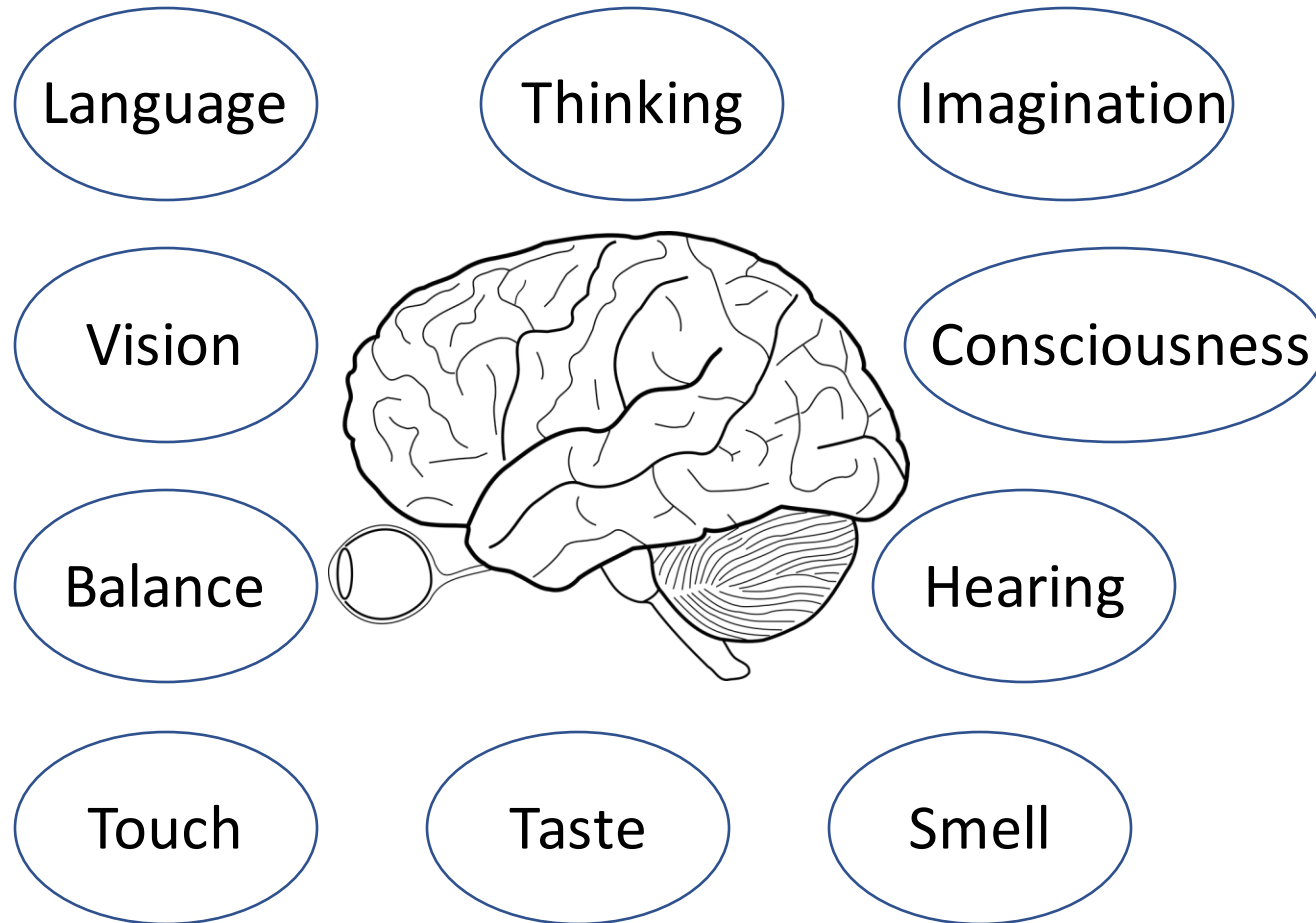
What is it?



# Artificial Intelligence: From "Natural Intelligence"?



# Artificial Intelligence: From "Natural Intelligence"?



Emulating all these:  
Artificial General Intelligence

What's AI then?

# Artificial Intelligence (AI)

Ability of man-made agents to do tasks traditionally attributed to “intelligence” of some level





# Artificial Intelligence (AI)

NEPALI - DETECTED

ENGLISH

SPANISH

FRENCH



ENGLISH

FILIPINO

NEPALI



आज म महाराजगंज टिचिङ अस्पतालमा पढाउँदै छु ।।

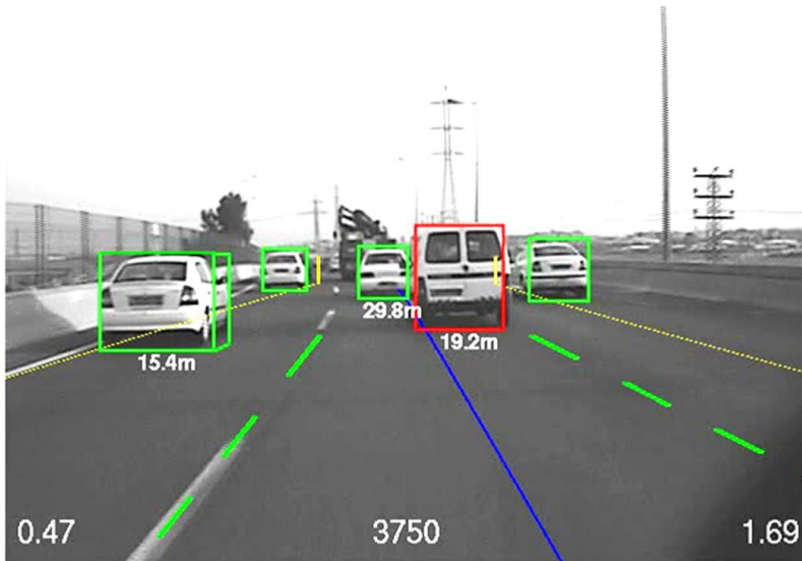


Today I am teaching at Maharajganj Teaching Hospital.

Āja ma mahārājagañja ṭiciṅa aspatālamā paḍhā'uṃdai chu.

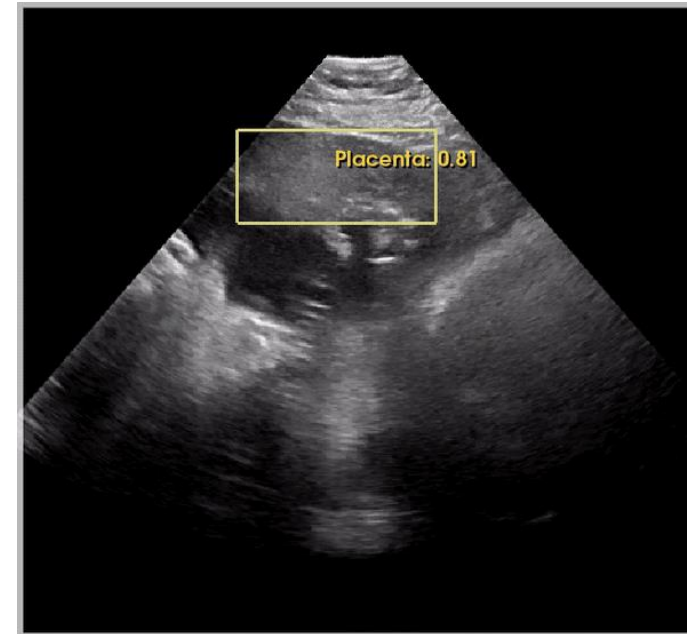


43 / 5,000

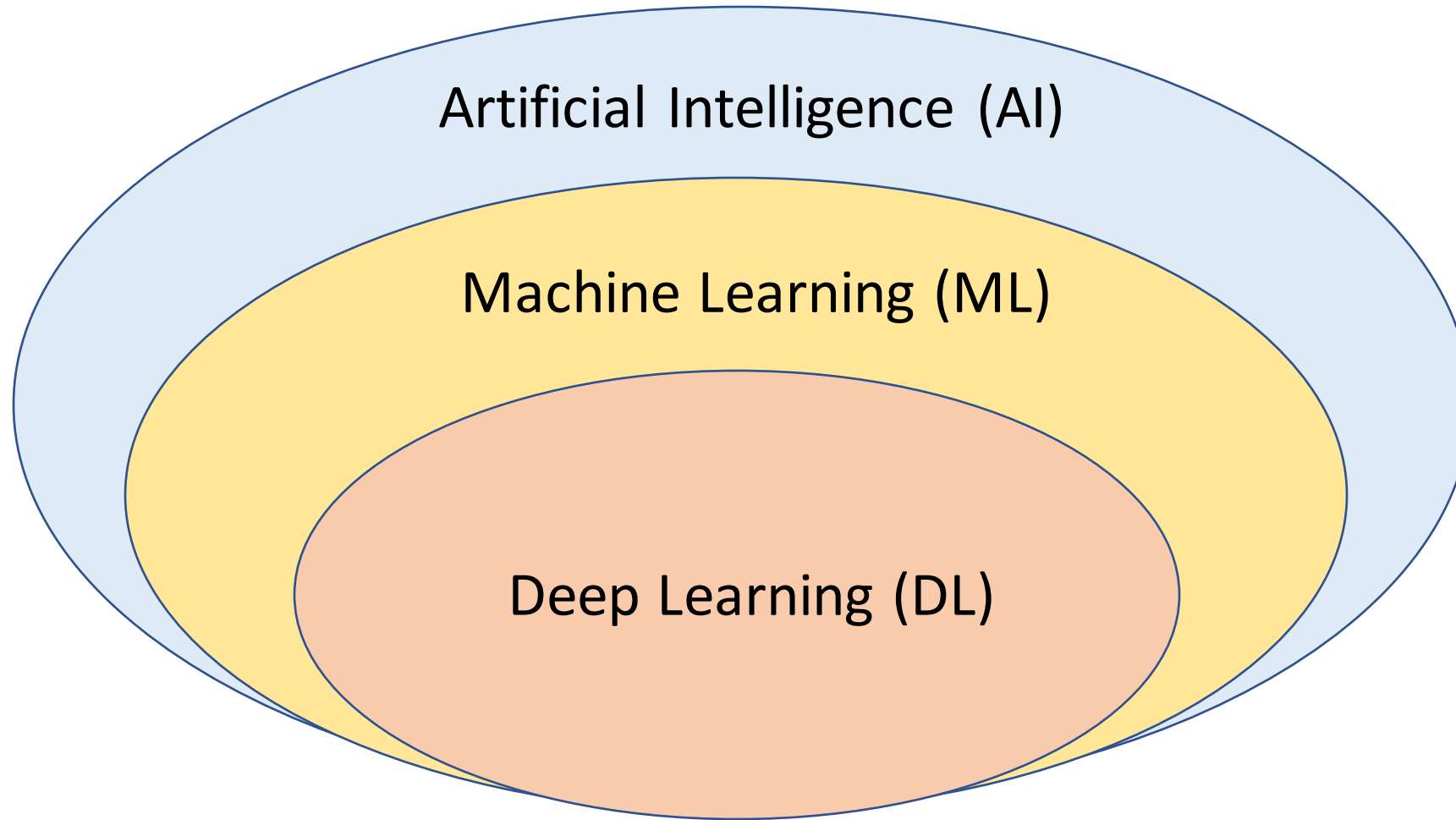


- Facebook Ads
- Facebook auto tagging
- Twitter Ads

- Netflix recommendation
- Amazon recommendation
- Youtube recommendation



# AI, Machine Learning and Deep Learning



# Supervised Deep Learning

Collect a large number of "annotated data"

- E.g. 10 lakhs chest X-ray images with diagnosis on each image



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- Train AI model (DL model) to predict diagnosis from 9 lakhs images
- Validate (Evaluate) the performance of AI model on 1 lakh images



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Evaluation metrics

- Accuracy, Sensitivity, Specificity



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**Generalization**

**Can we now use this AI model anywhere in the world?**

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Empowering general-public for prevention and early detection

Broadening the scope of health services in primary and community health centers

Improving quality of health service delivery in overburdened tertiary centers

Understanding biology and human body for better health care





# Key Themes in Nepal's Context

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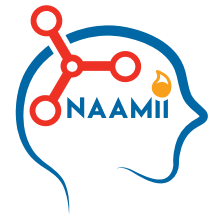
# Pesticide Estimation from Low-Cost Paper Device & AI App

Pesticides in Vegetables & Fruits

- **Big problem in Nepal**

**What if:**

- we go to pharmacy & buy a kit
- Grind vegetables
- See results of kit on Smartphone



Khanal, B., Pokhrel, P., **Khanal, B.**, & Giri, B. (2021). *Machine-Learning-Assisted Analysis of Colorimetric Assays on Paper Analytical Devices*. ACS omega, 6(49), 33837-33845.

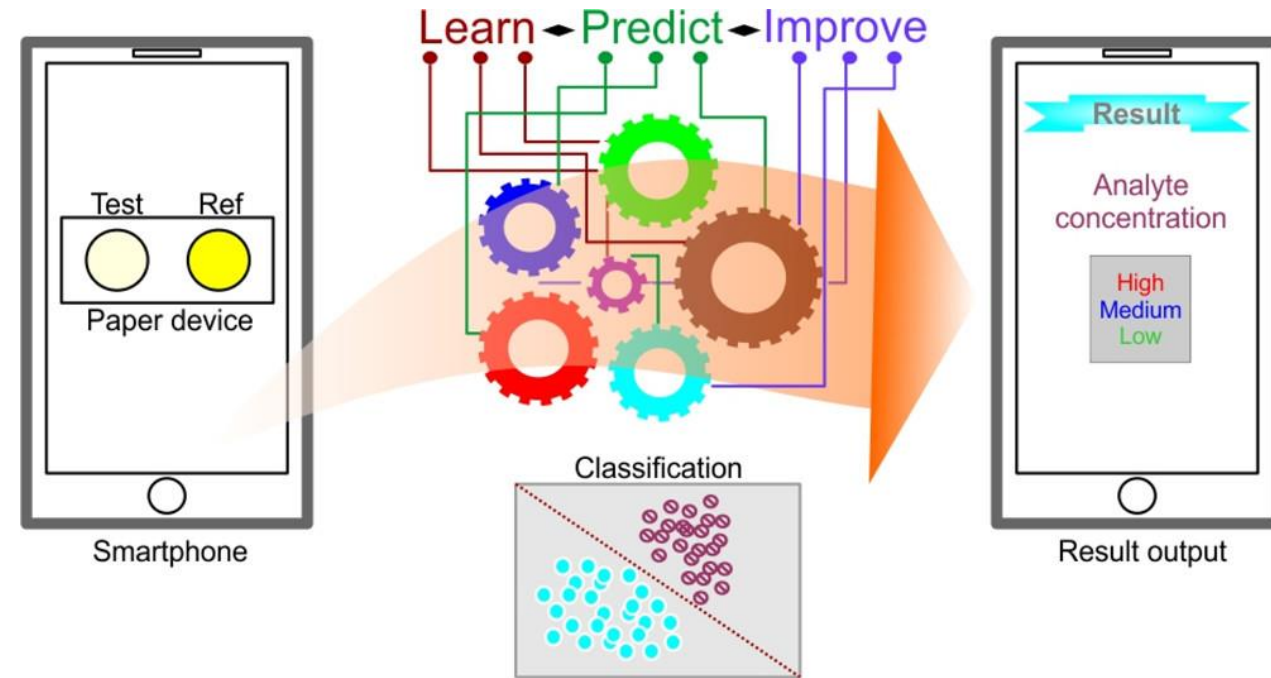
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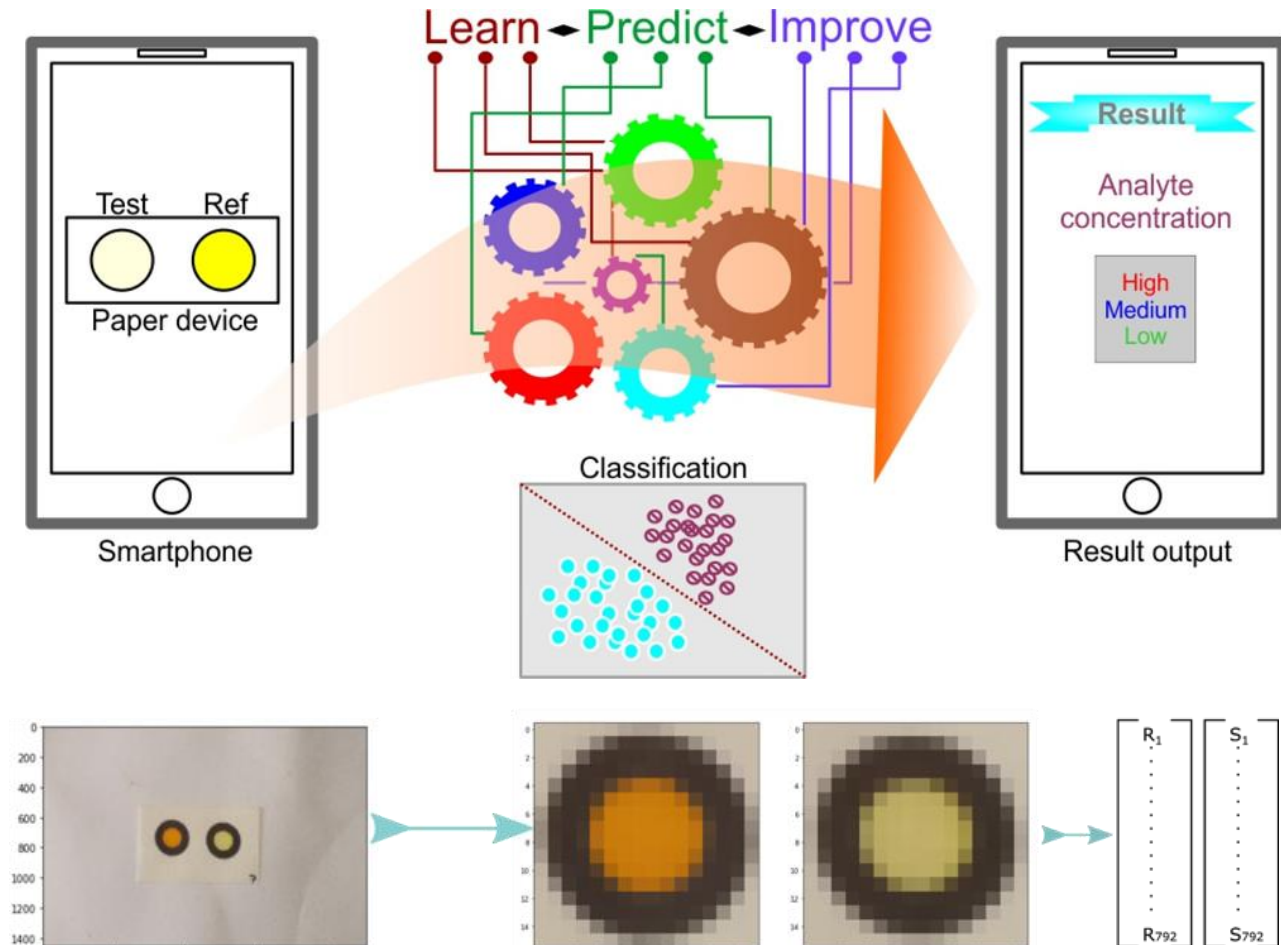
Pesticides in Vegetables & Fruits

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***We showed that this could be possible for major pesticides.***



Khanal, B., Pokhrel, P., Khanal, B., & Giri, B. (2021). *Machine-Learning-Assisted Analysis of Colorimetric Assays on Paper Analytical Devices*. ACS omega, 6(49), 33837-33845.

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# AI-assisted VIA for Cervical Cancer Screening

## Cervical Cancer

- Major type of cancer for women in Nepal (& many LMICs)
- Preventable if identified in pre-cancerous stage
- Can be cured if in early stage
- Visual Inspection with Acetic acid (VIA) most common screening in low-resource

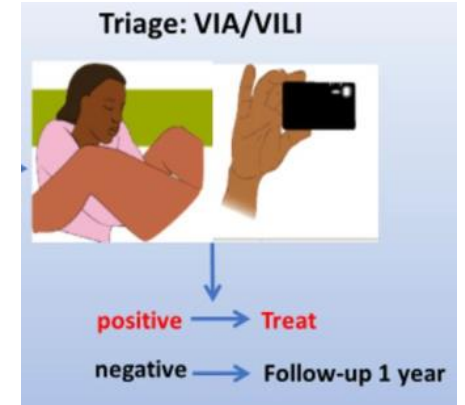
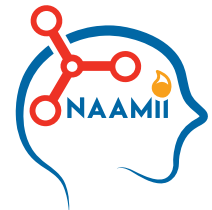


Image Source: Grohar D, et al. Scaling up community-based cervical cancer screening in Cameroon employing a single visit approach. IJGC 2020.



© Bishesh Khanal, NAAMII



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## Problem

- Must screen large fraction of the women population but lack of experts

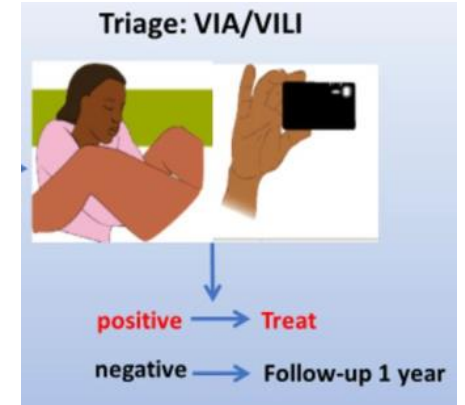


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## One Part of the Solution

- AI to assist non-experts (e.g. FCHVs, nurses) to do VIA more objectively & accurately

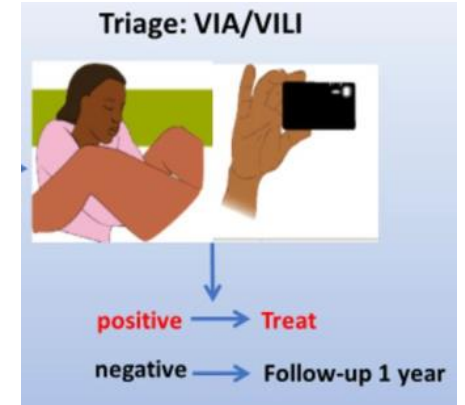
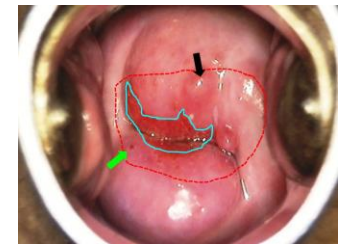


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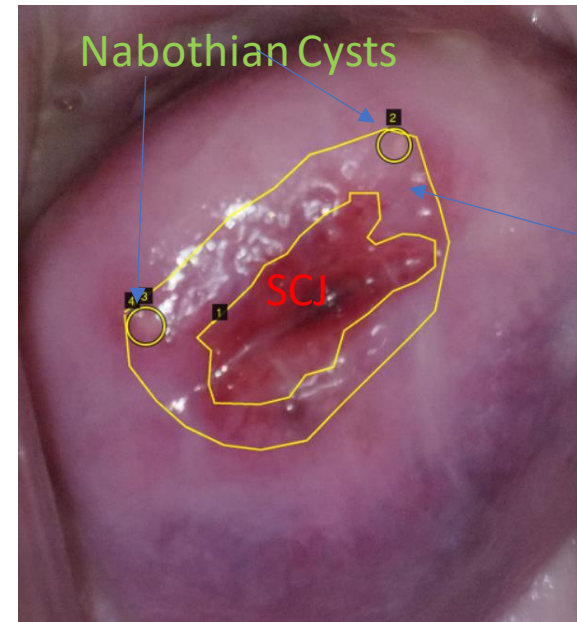




# AI-assisted VIA for Cervical Cancer Screening

## Anatomy of Cervical Cancer

- Most cancers and precancers originate from the squamous epithelium in the Transformation Zone(TZ) of the cervix



Transformation Zone (TZ)

# AI-assisted VIA for Cervical Cancer Screening

## Anatomy of Cervical Cancer

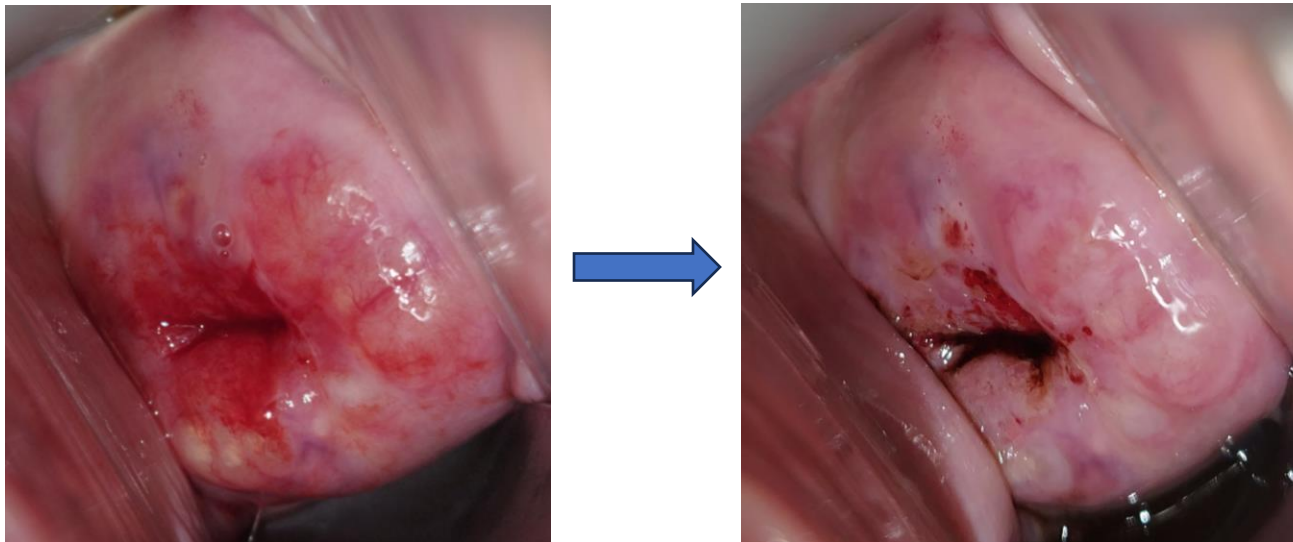
- But very difficult to locate these landmarks in most of the cases!



# AI-assisted VIA for Cervical Cancer Screening

## Morphology of Cervical Cancer

- Precancerous lesions take whitish color on application of acetic acid



Positive case from our camp verified as precancer by biopsy

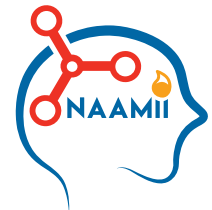
# AI-assisted VIA for Cervical Cancer Screening

## Morphology of Cervical Cancer


- Precancerous lesions take whitish color on application of acetic acid
- But, not every acetowhite change in cervix is a precancer or a cancer!



Positive cases from our camps, diagnosed as Infections and negative for cancer/pre-cancer

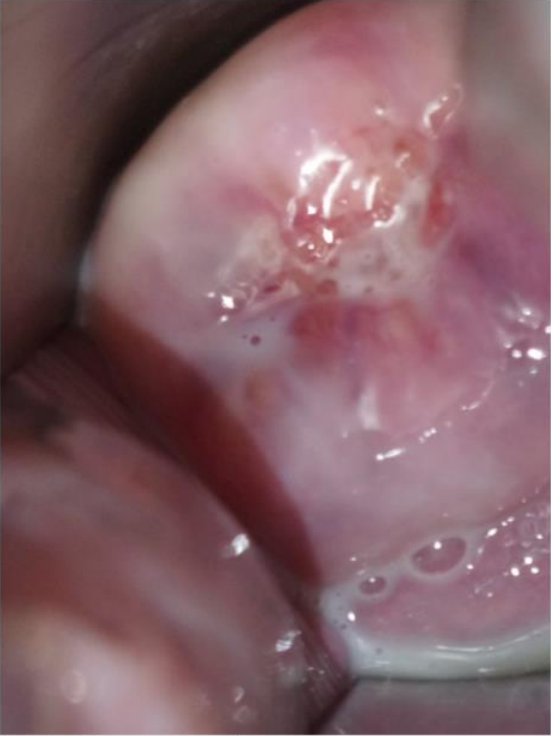


# AI-assisted VIA for Cervical Cancer Screening

 KANCHAN 🏠 Cervical Cancer Classification and Annotation Tool STUDY 894 [SIGN OUT](#) ↗


### Pre-Acetic Acid Image

[Zoom In](#) [Reset](#) [Zoom Out](#)




### Post-Acetic Acid Image 1 of 6


[Zoom In](#) [Reset](#) [Zoom Out](#)



[← PRE-ACETIC ACID IMAGE](#) [NEXT IMAGE →](#)

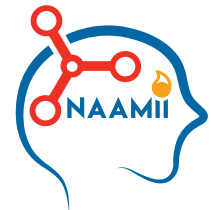
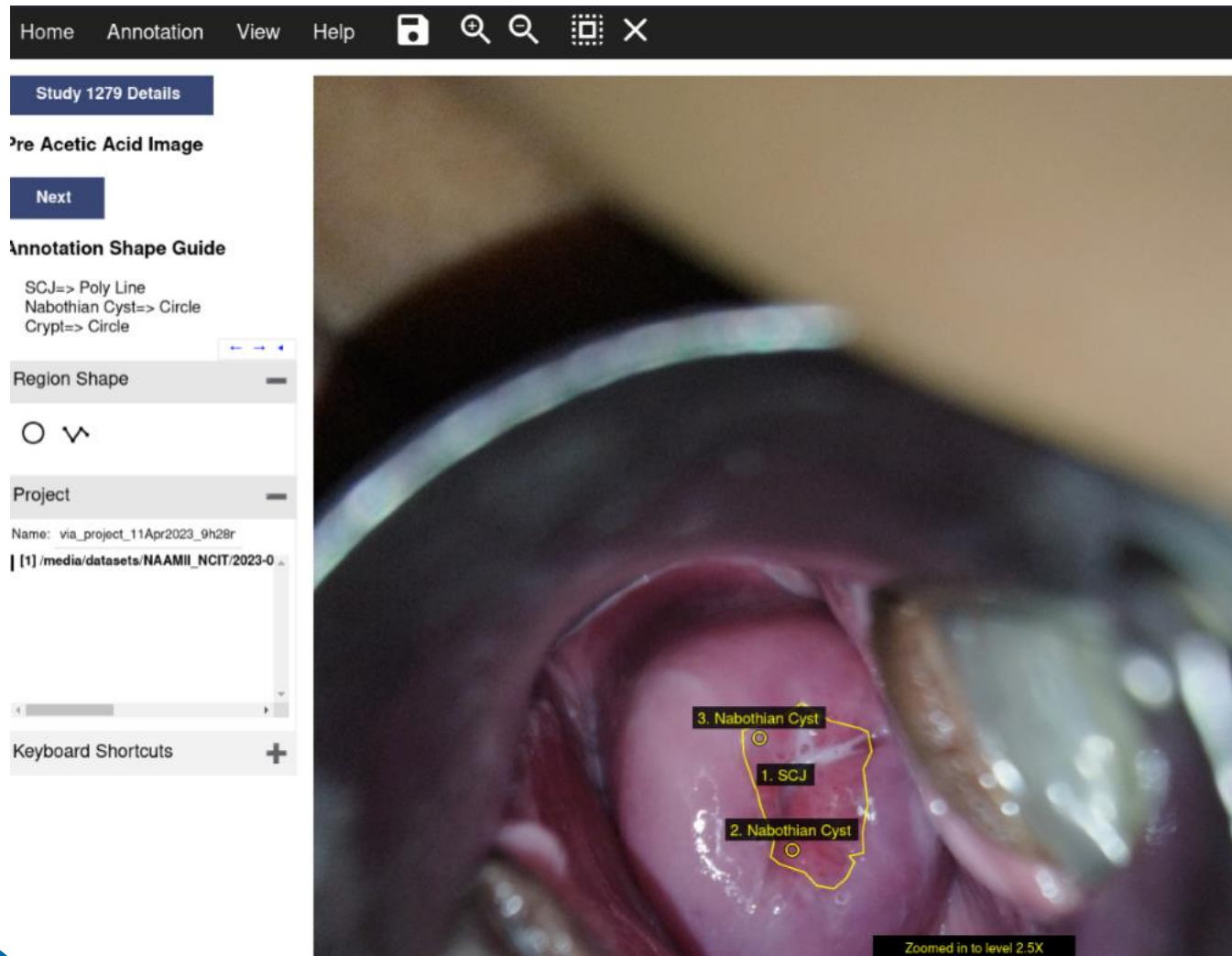
Time after acid application: **01 min, 00 sec**

Last Updated at	April 13, 2023, 5:30 a.m.
Cervix Image	Yes
Image Quality Issues	<a href="#">Obstruction in the view of the Cervix</a>
Adequate Quality for Decision	Yes
Visibility of the SCJ	Fully visible
Percentage of the Cervix Visible	>75%
Image Orientation	
Image Quadrants fully visible	[1, 2, 3, 4]
Outcome of this image	Negative
Annotated Labels	

[EDIT CLASSIFICATION ✓](#) [ANNOTATE THIS IMAGE](#) 



# AI-assisted VIA for Cervical Cancer Screening



# AI-assisted VIA for Cervical Cancer Screening

Pre-Acetic Acid Image

Zoom In Reset Zoom Out



Kanchan

Cervix Image	Yes
Image Quality Issues	
Adequate Quality for Decision	Yes
Visibility of the SCJ	Fully visible
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Image Orientation	
Image Quadrants fully visible	[1, 2, 3, 4]

Dr.Kumari

Cervix Image	Yes
Image Quality Issues	
Adequate Quality for Decision	Yes
Visibility of the SCJ	Fully visible
Percentage of the Cervix Visible	>75%
Image Orientation	
Image Quadrants fully visible	[1, 2, 3, 4]

Dr.Sabina

Cervix Image	Yes
Image Quality Issues	
Adequate Quality for Decision	Yes
Visibility of the SCJ	Fully visible
Percentage of the Cervix Visible	>75%
Image Orientation	
Image Quadrants fully visible	[2]

Annotations

Dr.Sabina



Dr.Kumari



Kanchan



# Key Themes in Nepal's Context

Empowering general-public for prevention and early detection

Broadening the scope of health services in primary and community health centers

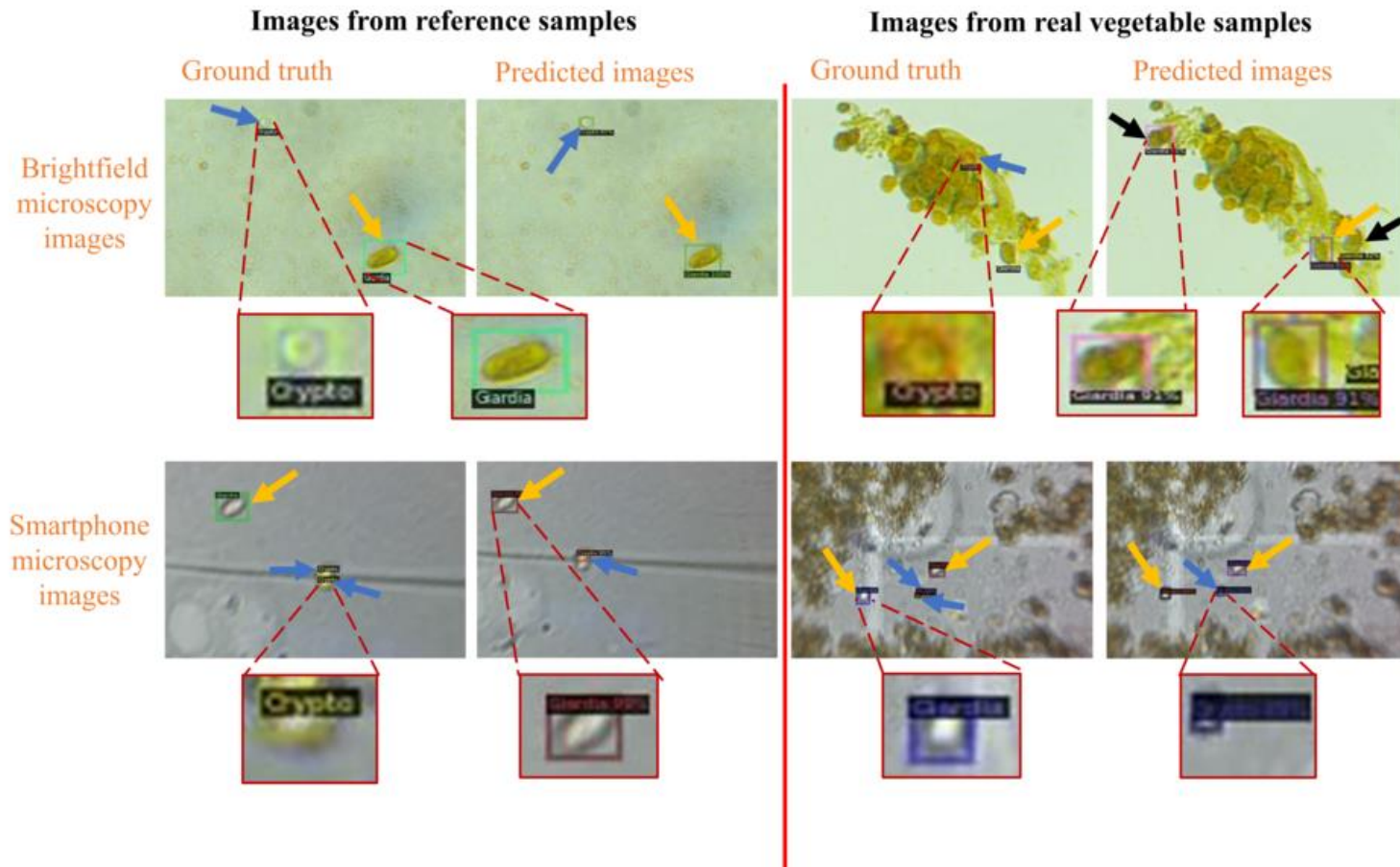
Improving quality of health service delivery in overburdened tertiary centers

Understanding biology and human body for better health care





# AI-Assisted Smartphone Microscopy



Diarrhea causing parasites

Rapid easy test in

Water, vegetables, stools

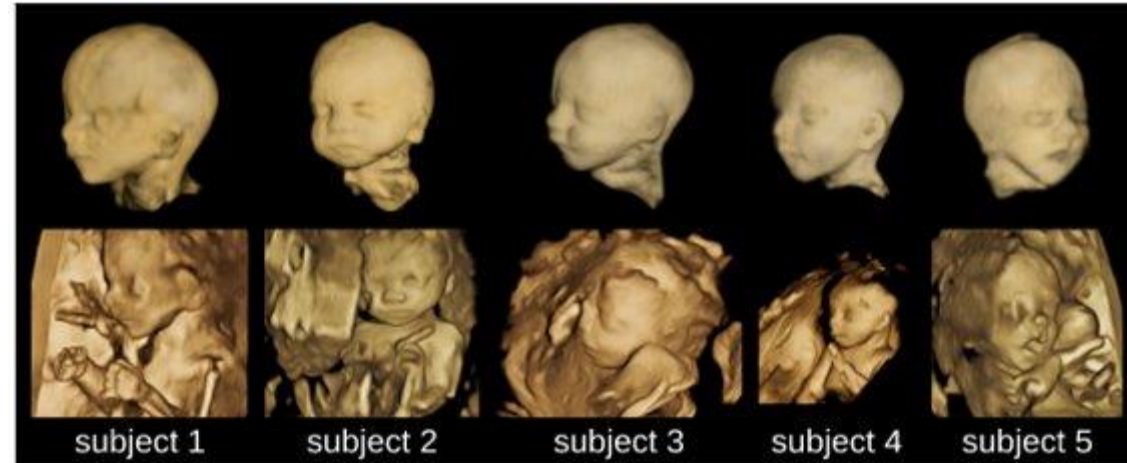
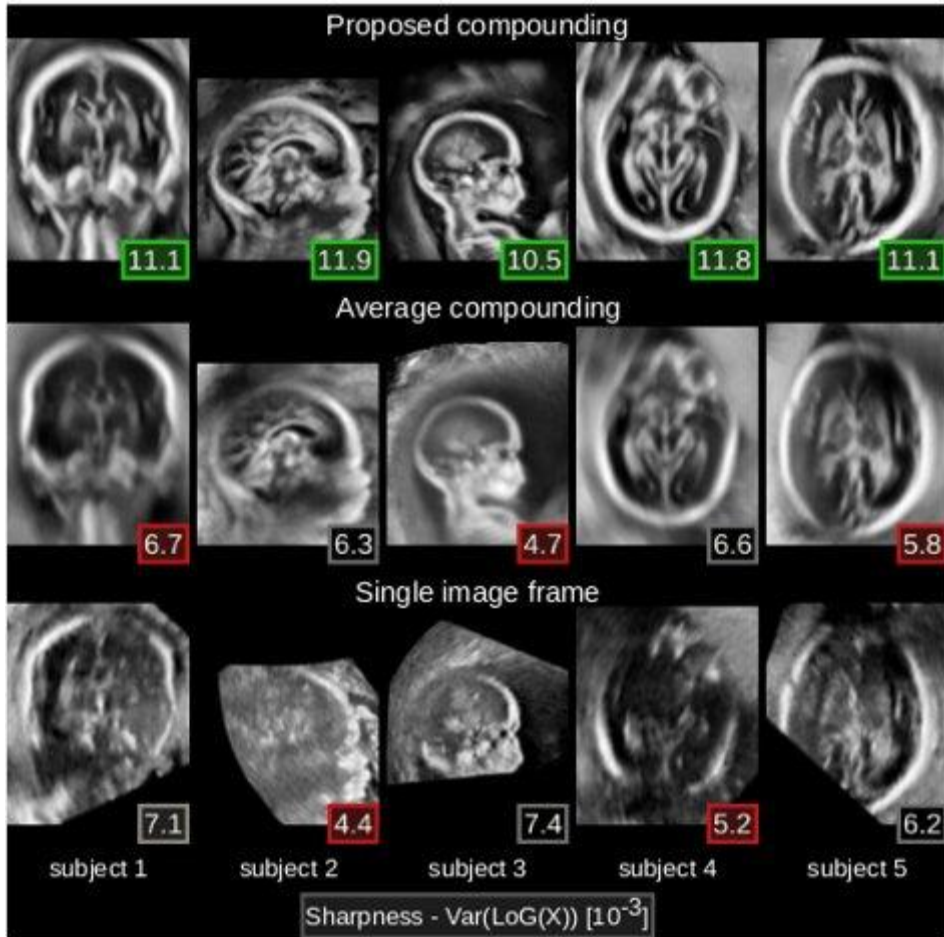
More objective & consistent detection



S. Nakarmi, ..., B. Khanal. Deep-learning assisted detection and quantification of (oo)cysts of Giardia and Cryptosporidium on smartphone microscopy images. ICLR Workshop 2021



# Intelligent Fetal Imaging and Diagnostics



## Addresses

- Limited Field of View
- Consistency across different operators
- Less expertise required

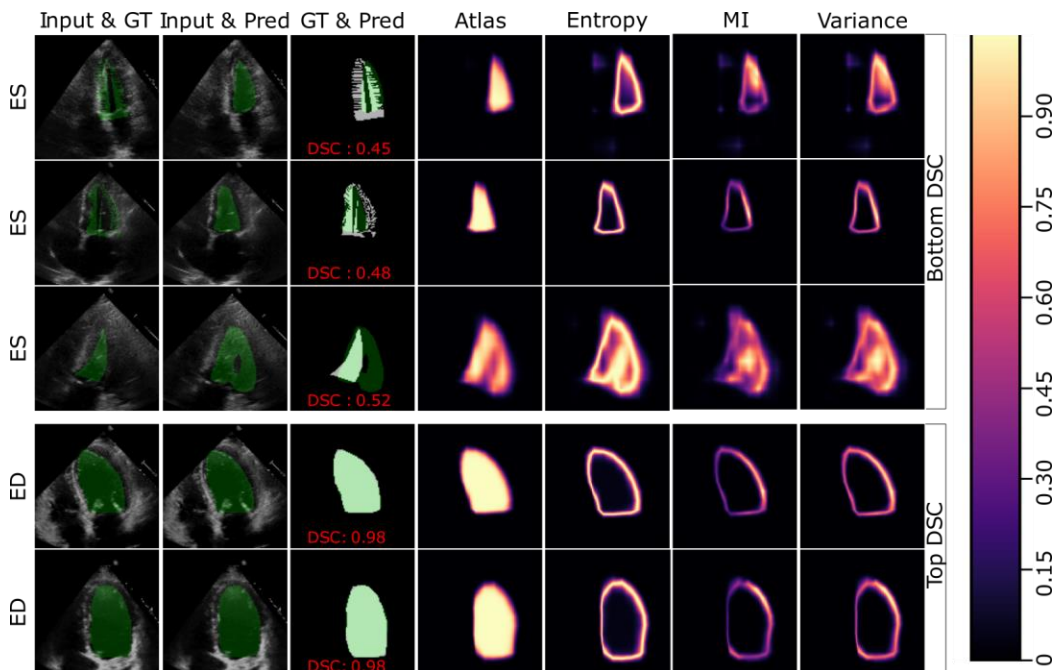
R. Wright, ..., B. Khanal, et al. Complete Fetal Head Compounding from Multi-View 3D Ultrasound... MICCAI 2019

R. Wright, ..., B. KHanal, et al. Fast fetal head compounding from multi-view 3D ultrasound. Media 2023.

# Uncertainty Based Quality Control in Biometrics Estimation

Automated cardiac structures segmentation at End Systole and End Diastole

Volume estimation, Ejection fraction; Important markers for Cardiovascular Disease



Dice scores taking most certain predictions

Test Set	First 20%	First 40%	First 60%	First 80%	Full Dataset (100%)	Current SOA
CAMUS-ED	0.953	0.946	0.944	0.935	0.932	0.939*
CAMUS-ES	0.944	0.936	0.928	0.923	0.911	0.916*
Dynamic-ED	0.946	0.942	0.939	0.936	0.930	0.927
Dynamic-ES	0.929	0.921	0.914	0.909	0.899	0.903

\* Current SOA result with manual cleaning of bad quality images, while ours is automated based on uncertainty output



# Key Themes in Nepal's Context

Empowering general-public for prevention and early detection

Broadening the scope of health services in primary and community health centers

- Inventing new technology when needed!

Improving quality of health service delivery in overburdened tertiary centers

Understanding biology and human body for better health care



# Biplanar X-ray to 3D reconstruction



Rural Trauma

- No CT scan available
- Generate 3D from X-ray



Shakya, M, and **Khanal, B.** Benchmarking Encoder-Decoder Architectures for Biplanar X-ray to 3D Bone Shape Reconstruction.  
Under review



# Benchmarking Biplanar X-ray to 3D Bone Reconstruction

Mahesh Shakya, Bishesh Khanal

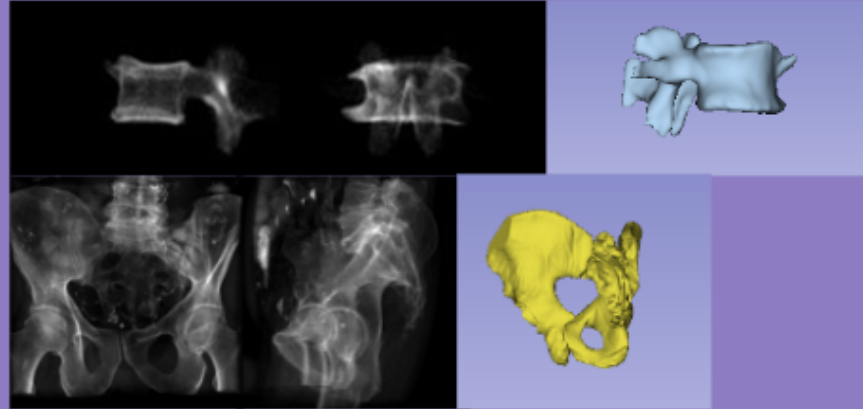
## Why X-ray to 3D Reconstruction?

- 3D segmentations useful in surgery planning but not always accessible
- **X-ray to 3D: high availability, lower cost.**

## Approaches

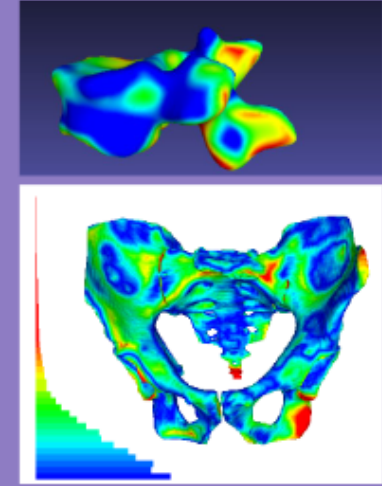
- Statistical Shape Model (poor out-of-training performance)
- 2D to 3D Registration (slow)
- Deep learning approach (data-hungry)

How can we improve performance on clinically relevant parameters of interest?



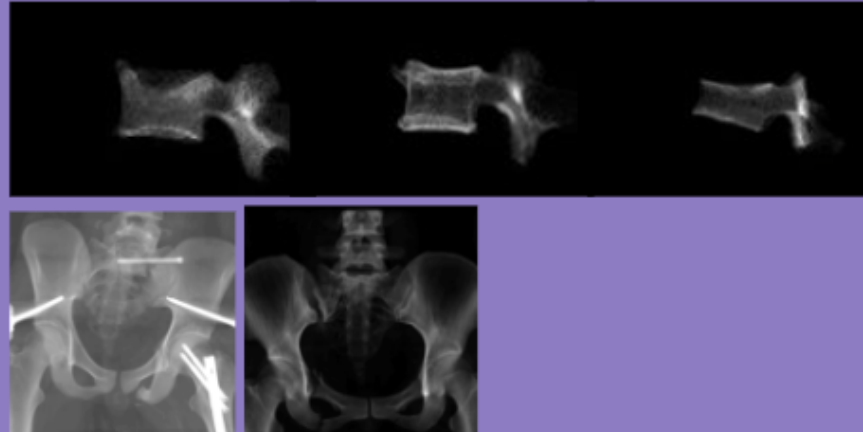
**Gap:** Medical Datasets are skewed. They usually contain fewer abnormal samples than healthy ones. AI models performance is biased towards majority class but results on these abnormal samples may be clinically more important.

How do current state of the art perform in public dataset? (Not as well as reported.)



Error: Red, Green, Blue color denote positive, zero and negative distance to ground truth. Avg reconstruction error 2.5 mm Trained on the challenging public Vertebra Dataset (VerSe<sup>1</sup>).

Dice Score grouped by vertebra pathology grade



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Empowering general-public for prevention and early detection

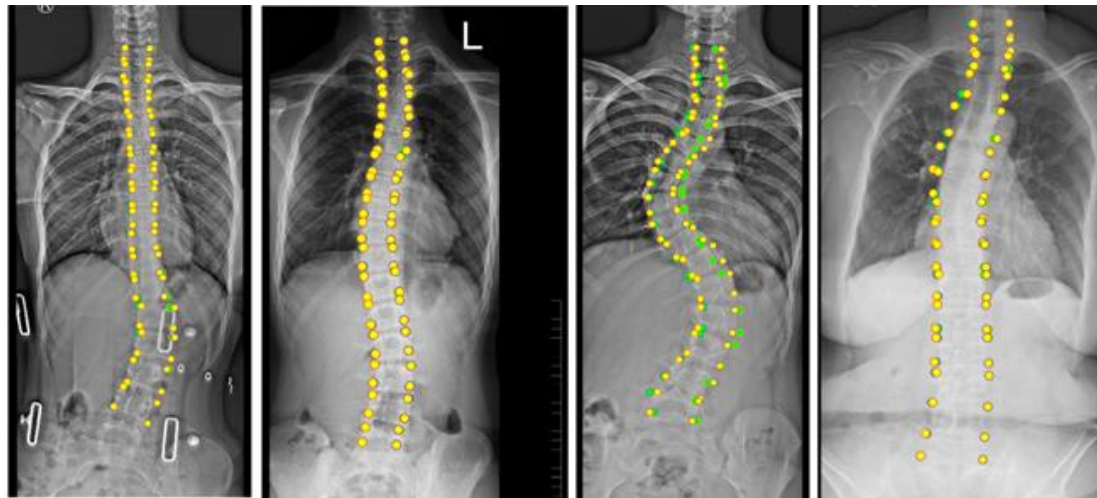
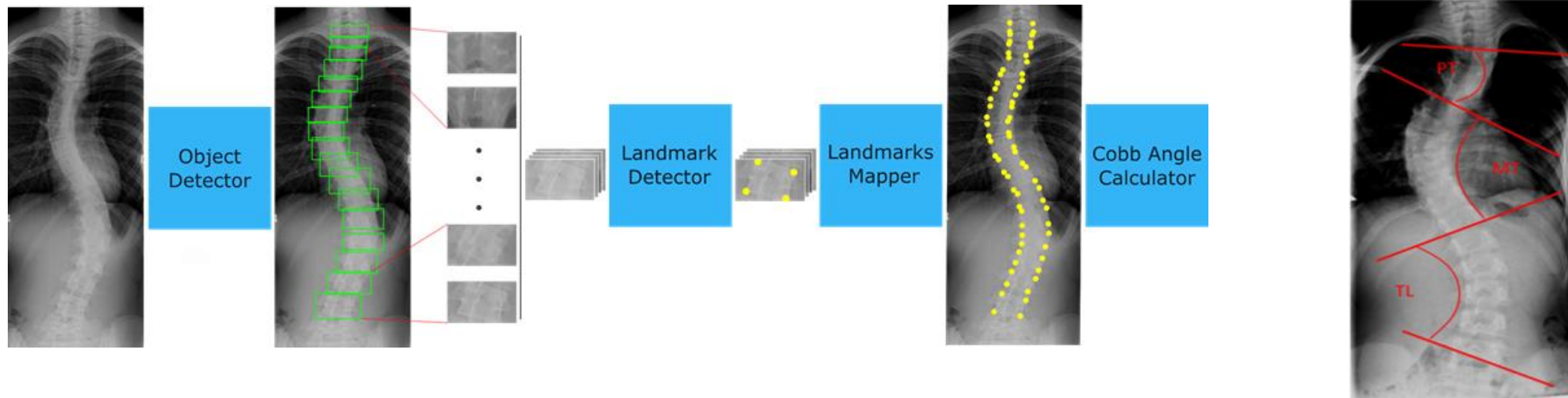
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# Automatic Cobb Angle Detection



## Scoliosis

- Common in adolescents
- Permanent disability or fatal if untreated
- Cobb angle an important marker
  - Tedious for surgeons
  - High inter-rater variability





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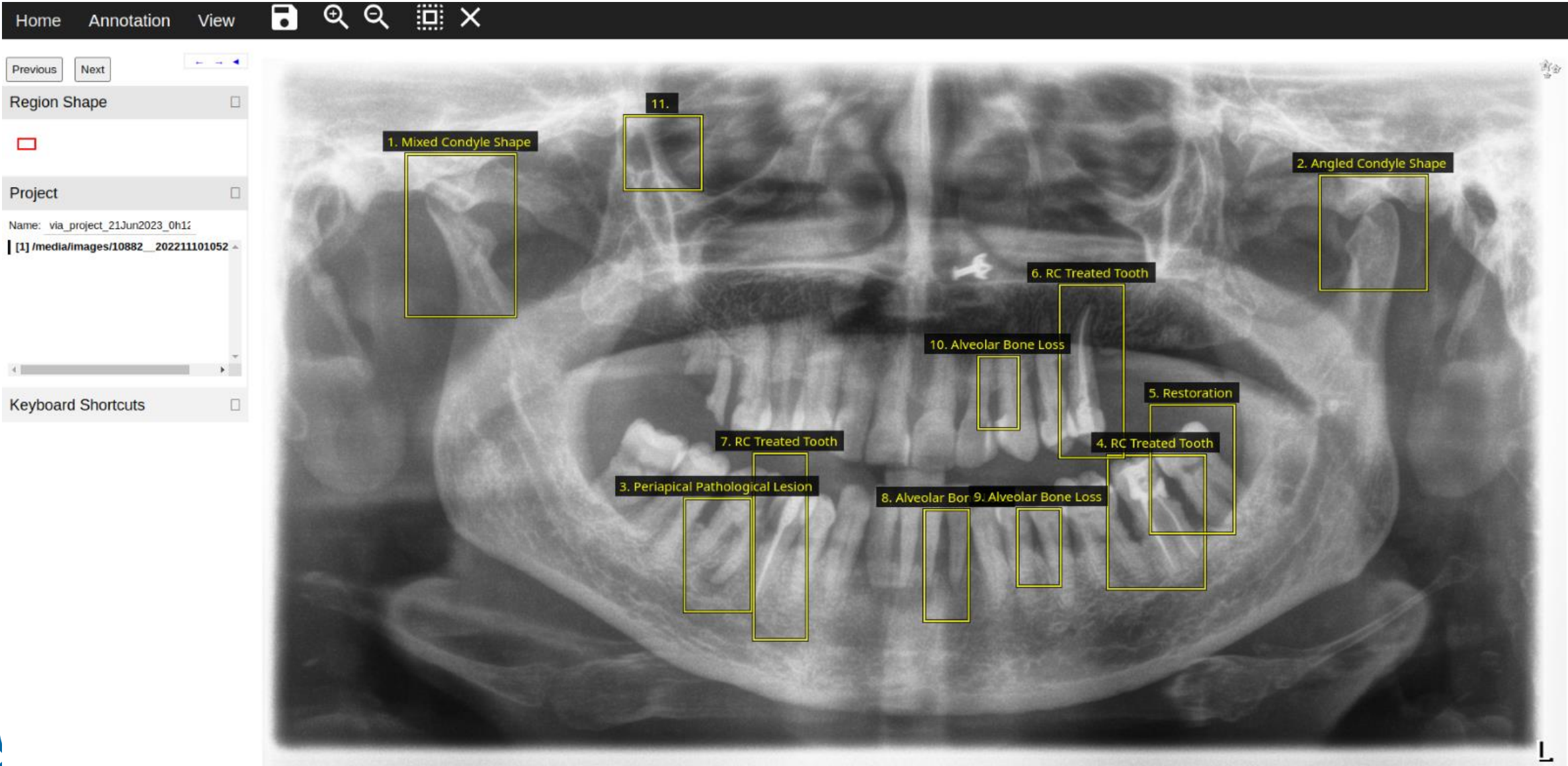
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# Dental Structures and Anomalies Detection



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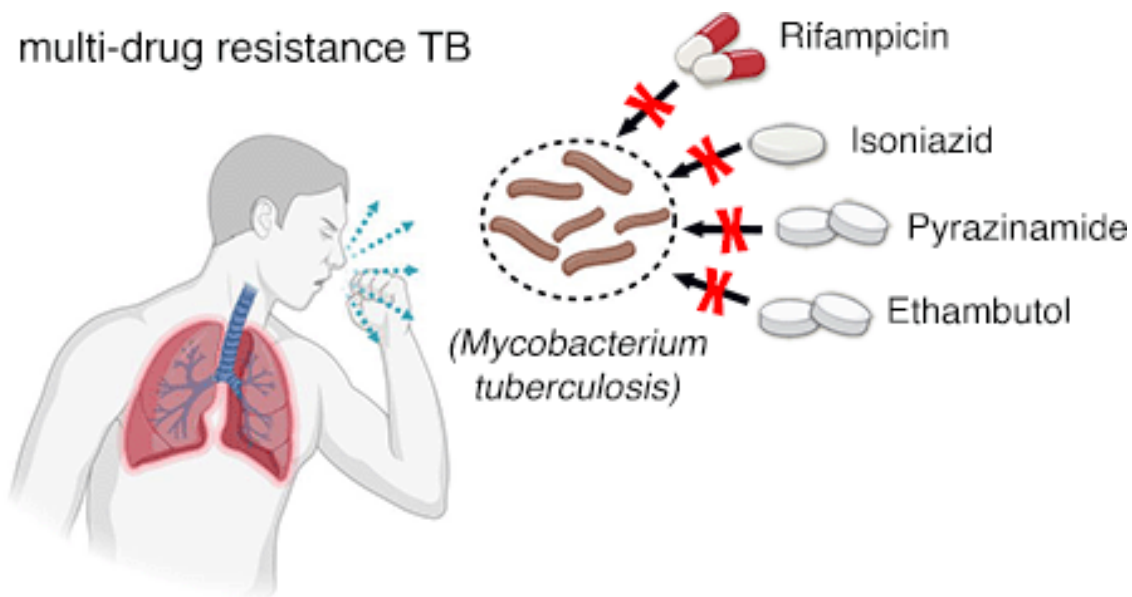
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Understanding biology and human body for better health care



# Tuberculosis (TB): a major infectious disease

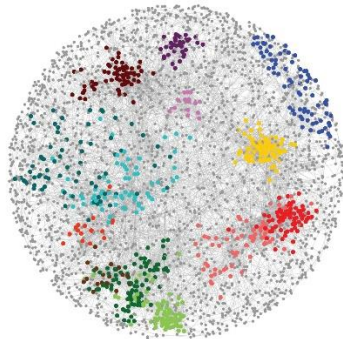


- Nepal has high TB burden
- TB bacteria are spread from person to person through the air
- Multi-drug resistant TB is biggest bottleneck in its eradication
- Gene mutations leading to multi-drug resistant TB are largely unknown

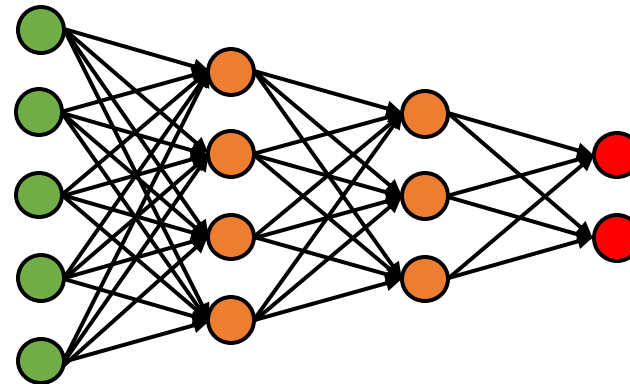
# Prediction of multi-drug resistant Tuberculosis

## TB mutations

ACCTGCTACA GCGGCAGGCAACTATGCCCAACGG  
ATGACAGGACTGCT GACGGCAACGGATTCTG  
GTTAGCGGAATCATCACC GACTCTCGATCGCGTTTC  
GGAGGTTCCA TCGTCGGCA GGACTTCGACC CGA CCGGGCGAT  
GTTAGCGGAATCATCACC GACTCTCGATCGCGTTTC



Protein-Interaction Network



Predicting drug-resistant TB and drug-resistant mutations

Understand mechanisms and interactions of genes

(work under progress)



RSTMH  
DEDICATED TO GLOBAL HEALTH SINCE 1907

NIHR | National Institute for Health Research



# Outline

Very Brief Intro to NAAMII

Brief Intro to AI

AI avenues in Nepal's Healthcare

**Ethical Considerations**



# AI and Ethics

## AI Ethics Survey in Nepal

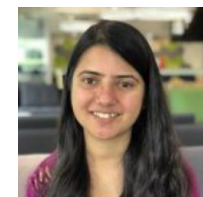
**Team:** [Shreyasha Paudel](#), [Aatiz Ghimire](#), [Bishesh Khanal](#)

### Project Goals

The main goal of this project is to do a landscape mapping of understanding and approaches to AI Ethics in Nepal among three key stakeholders: Technology Students, Technology Professionals, and Policymakers.

There is an immense potential of AI in helping positively transform the economic situation of countries like Nepal. However, there is also a big risk that the AI-driven development dramatically broadens the gap of the existing digital divide. In this context, there is a need for national conversation and shared vocabulary among different stakeholders to understand each other's concerns and challenges. A shared vocabulary is necessary between users, policy-makers, activists, and developers of AI to further the growth of the industry in an ethical and responsible manner. This project will fill the gap by doing a survey and publicizing challenges and gaps in understanding of issues related to AI ethics.

The survey design will be informed by state of the research and similar reports published in the global context and link it to gaps and challenges that may exist due to local context within Nepal. The main outcome of this project will be an informational resource that will contribute to an increased understanding and shared vocabulary among multiple stakeholders regarding ethical concerns with AI. We also believe that the insights from this project will be helpful for other developing countries with a rapidly growing AI ecosystem and provide opportunities and shared vocabulary to address ethical concerns among different stakeholders.



**Shreyasha Paudel**  
Adj. Research Fellow



# FUTURE-AI: Best practices for trustworthy AI in medicine

FUTURE-AI is an international, multi-stakeholder initiative for defining and maintaining concrete guidelines that will facilitate the design, development, validation and deployment of trustworthy AI solutions in medicine and healthcare based on six guiding principles: Fairness, Universality, Traceability, Usability, Robustness and Explainability.

1

## FAIRNESS

The Fairness principle states that medical AI tools should maintain the same performance across individuals and groups of indiv...

[Learn More >](#)

2

## UNIVERSALITY

The Universality principle states that a medical AI tool should be generalisable outside the controlled environment where it wa...

[Learn More >](#)

3

## TRACEABILITY

The Traceability principle states that medical AI tools should be developed together with mechanisms for documenting and monito...

[Learn More >](#)

4

## USABILITY

The Usability principle states that the end-users should be able to use a medical AI tool to achieve a clinical goal efficientl...

[Learn More >](#)

5

## ROBUSTNESS

The Robustness principle refers to the ability of a medical AI tool to maintain its performance and accuracy under expected or ...

[Learn More >](#)

6

## EXPLAINABILITY

The Explainability principle states that medical AI tools should provide clinically meaningful information about the logic behi...

[Learn More >](#)



# Researchers Involved



**FUTURE-AI: International consensus guideline for trustworthy and deployable artificial intelligence in healthcare. Under review**

# Researchers Involved



FUTURE-AI: International consensus guideline for trustworthy and deployable artificial intelligence in healthcare. Under review

# FUTURE-AI: Trustworthy and Deployable AI in Healthcare

## Problems

Errors and patient harm

Biases and increased health inequalities

Lack of transparency and accountability

Lack of data privacy and security breaches

## Status

Absence of clear, widely accepted guidelines on how medical AI tools should be designed, developed, evaluated and deployed to be trustworthy



FUTURE-AI: International consensus guideline for trustworthy and deployable artificial intelligence in healthcare. Under review

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# FUTURE-AI: Trustworthy and Deployable AI in Healthcare

## Problems

Errors and patient harm

Biases and increased health inequalities

Lack of transparency and accountability

Lack of data privacy and security breaches

## Needs

Technically robust, clinically safe, ethically sound and legally compliant

Robustness, safety, security, fairness, transparency, traceability, accountability, generalisability, explainability, usability and responsible AI

Trusted and accepted by patients, clinicians, health organisations and authorities



# FUTURE-AI: Trustworthy and Deployable AI in Healthcare

Guidance for constructing medical AI tools that will be trusted, deployed and adopted in real-world clinical practice

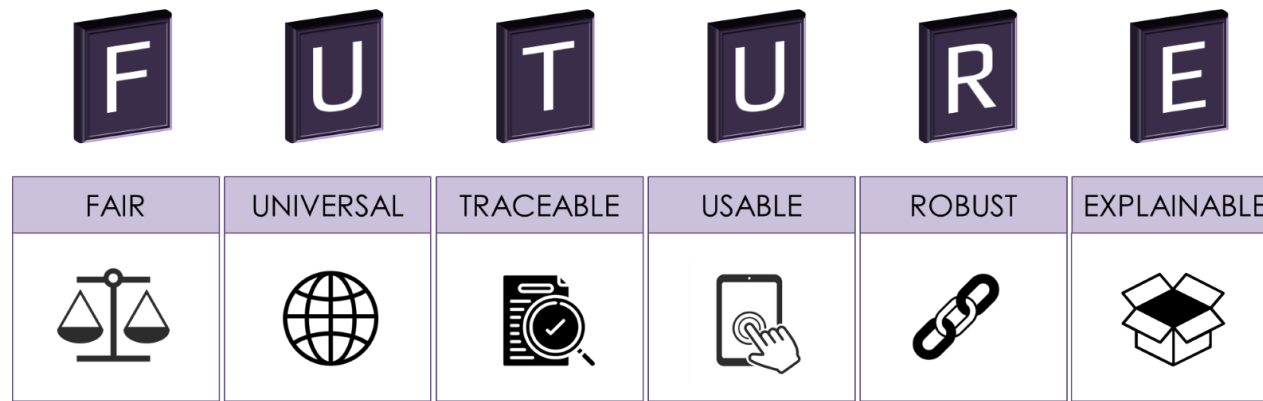
A set of 28 best practices addressing technical, clinical, legal and socio-ethical dimensions of trustworthy AI

Covers the entire lifecycle of medical AI, from design, development and validation to regulation, deployment, and monitoring



**FUTURE-AI: International consensus guideline for trustworthy and deployable artificial intelligence in healthcare. Under review**

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**Fair:** Same performance across individuals and groups of individuals (including under-represented and disadvantaged groups)

**Universal:** Generalisable outside the controlled environment where it was built; to new patients and new users (e.g. new clinicians), and when applicable, to new clinical sites

**Traceable:** Mechanisms for documenting and monitoring the complete trajectory of the AI tool, from development and validation to deployment and usage



<b>F</b>	<b>U</b>	<b>T</b>	<b>U</b>	<b>R</b>	<b>E</b>
FAIR	UNIVERSAL	TRACEABLE	USABLE	ROBUST	EXPLAINABLE
					

**Usable:** The end-users should be able to use a medical AI tool to achieve a clinical goal efficiently and safely in their real-world environment; easy functionalities and interfaces with minimal errors + clinically useful and safe

**Robust:** The ability of a medical AI tool to maintain its performance and accuracy under expected or unexpected variations in the input data. Existing research has shown that even small, imperceptible variations in the input data may lead AI models into incorrect decisions

**Explainable:** Medical AI tools should provide clinically meaningful information about the logic behind the AI decisions



**FUTURE-AI: International consensus guideline for trustworthy and deployable artificial intelligence in healthcare. Under review**

# Ethical Considerations in AI research and product development

**Problem Definition**

**Data Collection**

**Data Preprocessing**

**Training AI Model**

**Clinical Decision Making**

**Diagnosis and Treatment**





# Ethical Considerations in AI research and product development

## Problem Definition

Who chooses the problems to solve?

What problems to solve?

Capacity, funding, ...



# Ethical Considerations in AI research and product development

## Problem Definition

Who chooses the problems to solve?

What problems to solve?

Capacity, funding, ...

**Mindset?**



# Ethical Considerations in AI research and product development

**Data Collection**

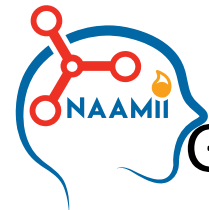
**Data Preprocessing**

**Training AI Model**

(Digital) Data infrastructure and culture of recording data

Privacy, anonymization, information security

Preprocessing for one database may not work for another database



GPU and computational power availability

# Ethical Considerations in AI research and product development

## Retrospective Data

- Informed consent? Waiver?
- E.g. 10 Lakhs X-ray images from different hospitals -> anonymized and shared



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## Personally Identifiable data vs not ?

- Cervix images
- Neuromuscular disease severity from facial deformities (Dystonia)
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## Sample size and statistical evaluation

- Sample size calculation during AI methods development is an open problem
- Model development and research is different from evaluating model's performance



# Ethical Considerations in AI research and product development

## Research and Product development

- Capacity development and building ability to build products in Nepal Vs
- Exploited and the status of perpetually being consumer from foreign companies Vs
- Providing patients the best that is available in the world to save lives or improve quality



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- Regulations in Nepal for AI Software? Accountability? DDA?





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## AI product regulations

- Evaluation: where and in what context?
- E.g. cervical pre-cancer detection accuracy: AI vs Gynecologist or AI vs nurses



# AI: An Opportunity to Democratize Innovation in Medicine

Recent advances in Artificial Intelligence (AI) provides us a unique opportunity

Cutting edge and frontier research in medicine possible

Many problems relevant to Nepal and global south **NEEDS** cutting edge research but are  
**GLOBALLY NEGLECTED**

**Identify REAL problems, Frontier research Needed & Possible!**

**Sustainability through innovation & entrepreneurship!!**

**Need to build facilitative contextualized regulations with no harm principles**

