



# AI Foundation Models in Surgery: Future Directions and Global Impact

Intelligent Health 2024

Basel, Switzerland – September 12, 2024

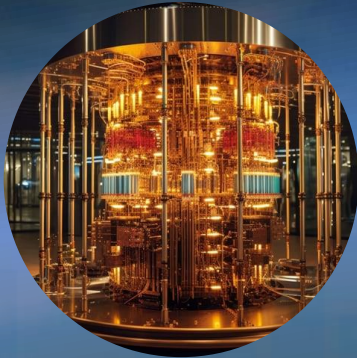
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Vice Chair for Innovation, Department of Surgery

Surgical Director of Duke AI Health

# The Era of Technological Disruption



# ChatGPT



### Capabilities



### Limitations

uting in

Remembers what user said earlier in the conversation

May occasionally generate incorrect information

10

Allows user to provide follow-up corrections

May occasionally produce harmful instructions or biased content

Trained to decline inappropriate requests

Limited knowledge of w events after

## Innovation and Safety

“Technology is altering the landscape across various sectors. and the future is unfolding before our eyes.”

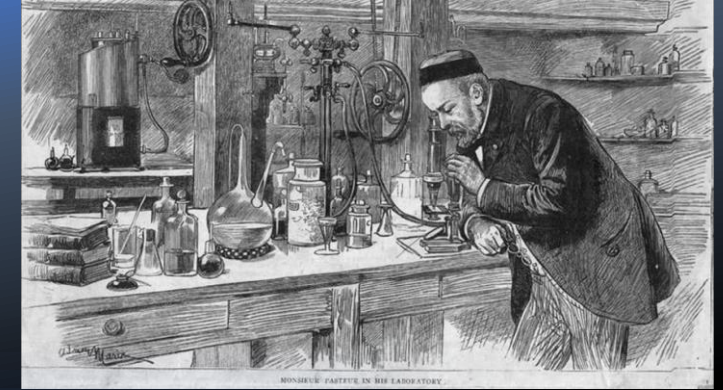


# What is happening in the Operating Room ?



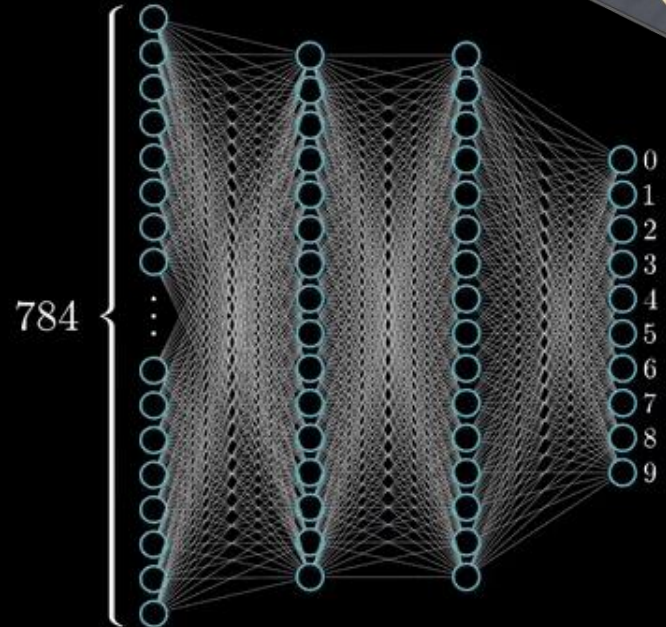
# Surgical Revolutions (in the last 200 years)

- General Anesthesia – 1840s
- Antiseptic Surgery – 1860s
- Endoscopic Procedures – 1960s
- Cognitive Computing – 2010s



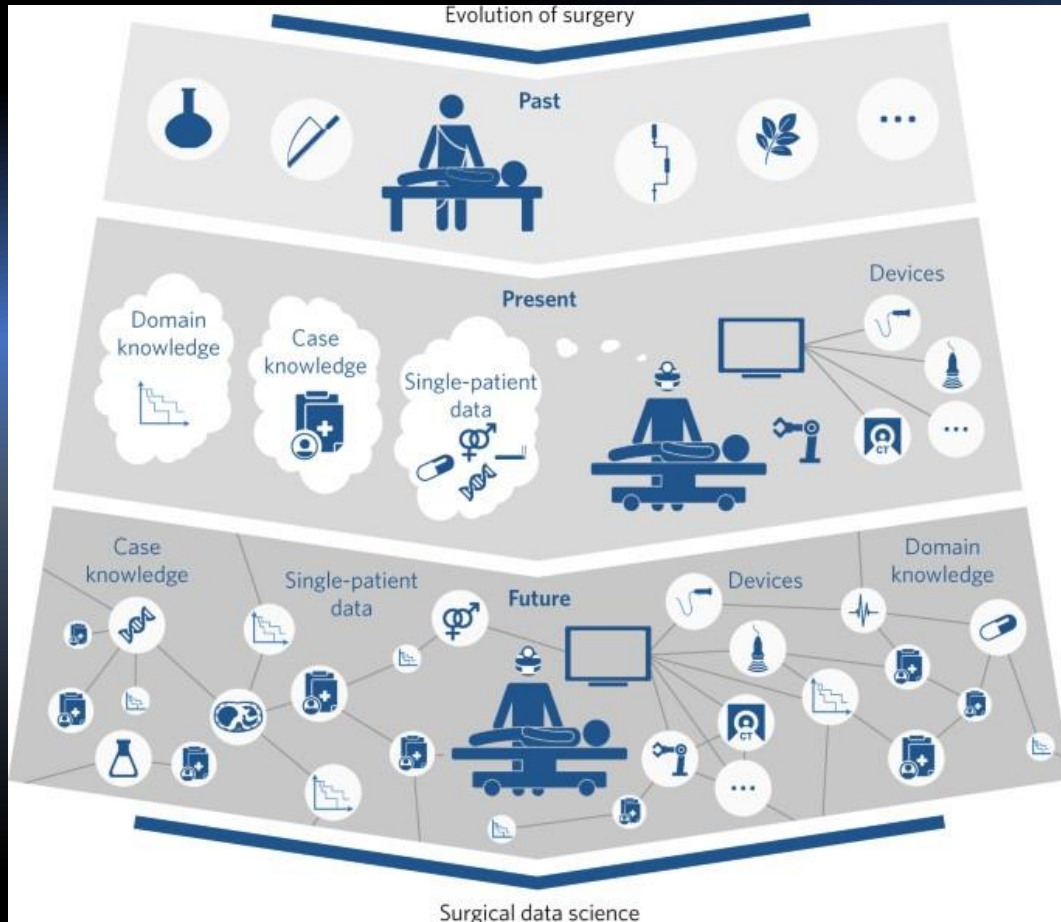
# Video DATA

# More computing power



## More powerful/efficient techniques

# Large amount of DATA



The average hospital produces roughly 50 petabytes of data every year. That's more than twice the amount of data housed in the Library of Congress, and it amounts to **137 terabytes per day**.

## BIG DATA IN FORMULA ONE



Formula One cars generate **terabytes of data** during a race. Dozens of engineers at the track and as far away as the U.K. comb over the data during a race in near real-time, looking for any adjustment that could **win or lose** a race.

RACE TEAMS COMBINED TO GENERATE **243 TERABYTES** OF DATA FROM THEIR VEHICLES AT THE 2014 U.S. GRAND PRIX IN AUSTIN, TX.

243 TERABYTES OF DATA COMPARED TO ...



EQUIPPED WITH **HUNDREDS OF SENSORS**, F1 CARS PROVIDE A STREAM OF DATA THAT'S ANALYZED **THOUSANDS OF MILES** AWAY IN NEAR REAL-TIME



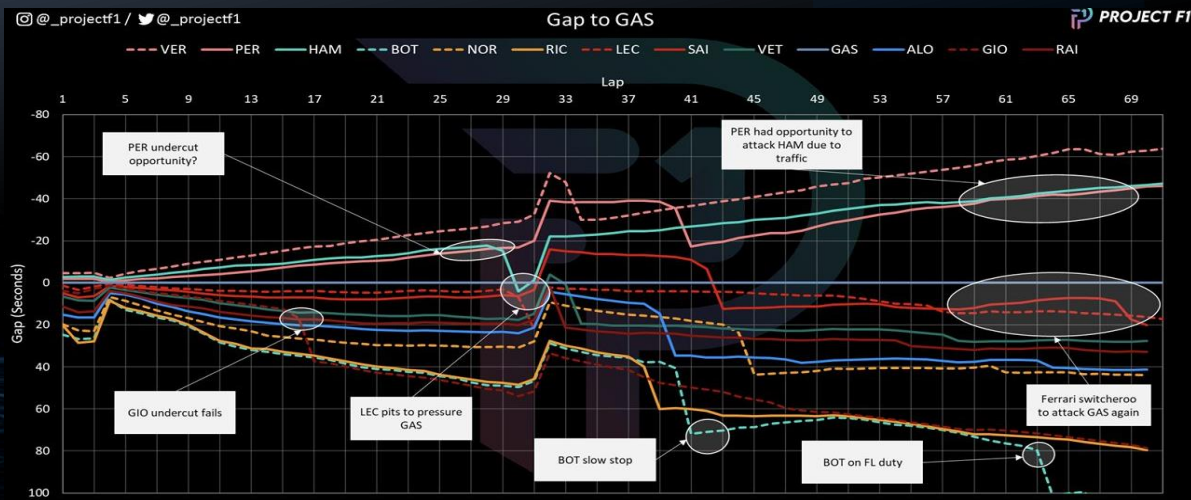
RACE FANS GENERATED MORE THAN **2.3 TERABYTES** OF AT&T MOBILE DATA DURING THE U.S. GRAND PRIX BY SHARING PHOTOS AND SENDING TWEETS, LESS THAN 1% COMPARED TO THE RACING TEAMS.



SOURCES: Infiniti Red Bull Racing, AT&T, Library of Congress, Twitter

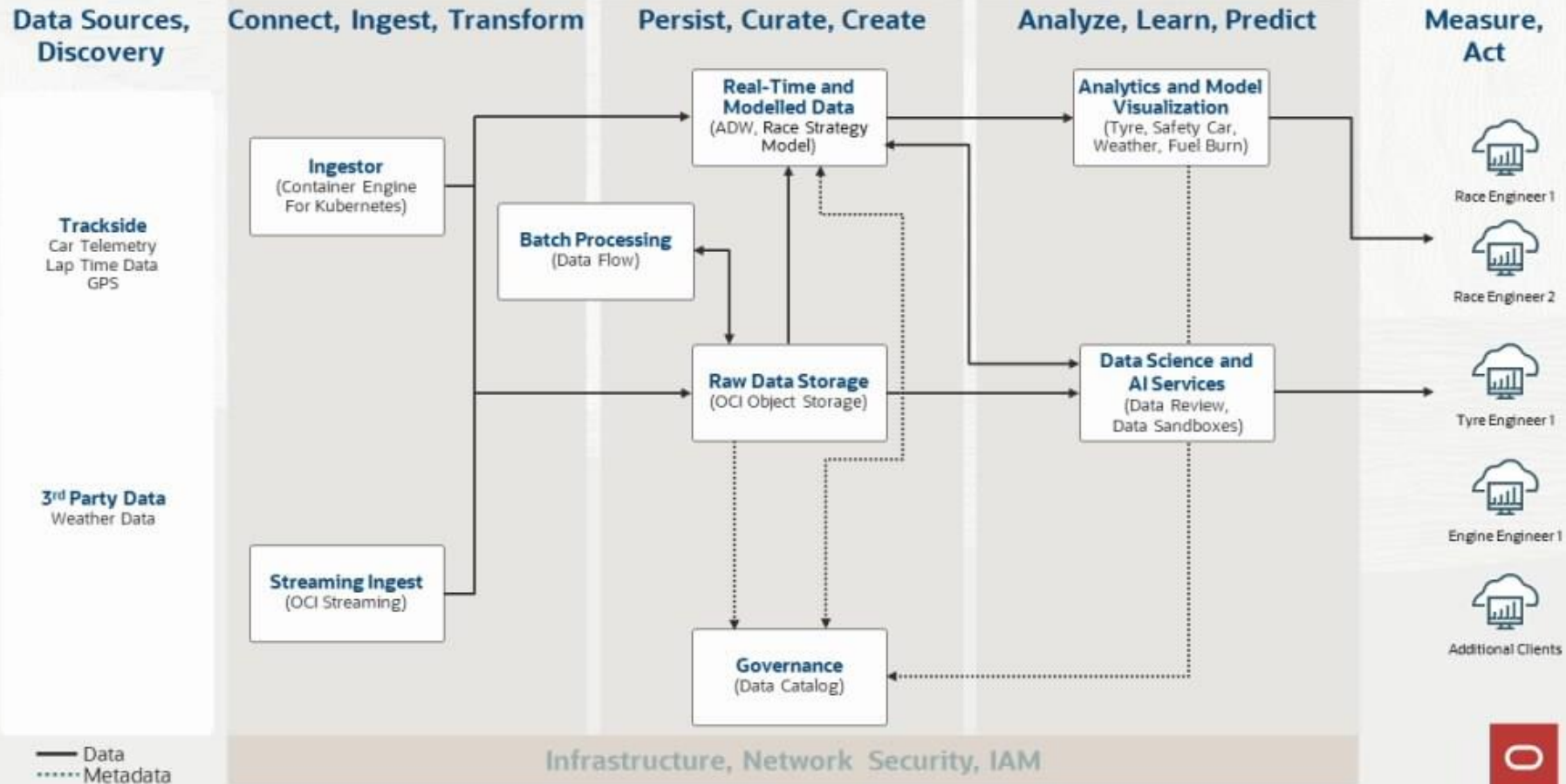
FORBES MEDIA





# Error Handling and Recovery

## Oracle Data Platform – Race Strategy





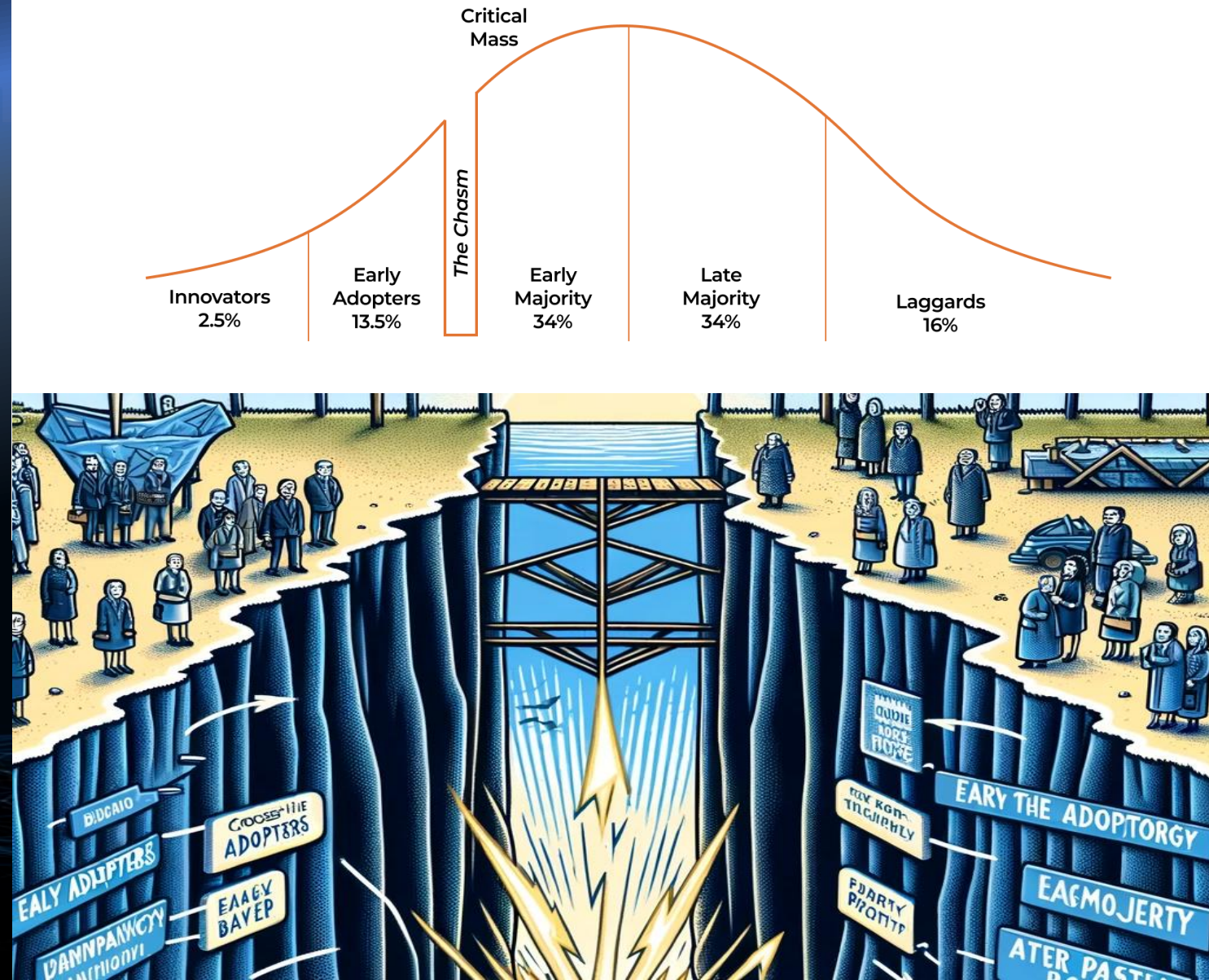
Healthcare, specially in surgery is a sector where technology's promise has not been fully realized.



“We have advanced machinery and data analytics, but the operating room still relies heavily on human judgment, sometimes leading to inefficiencies”

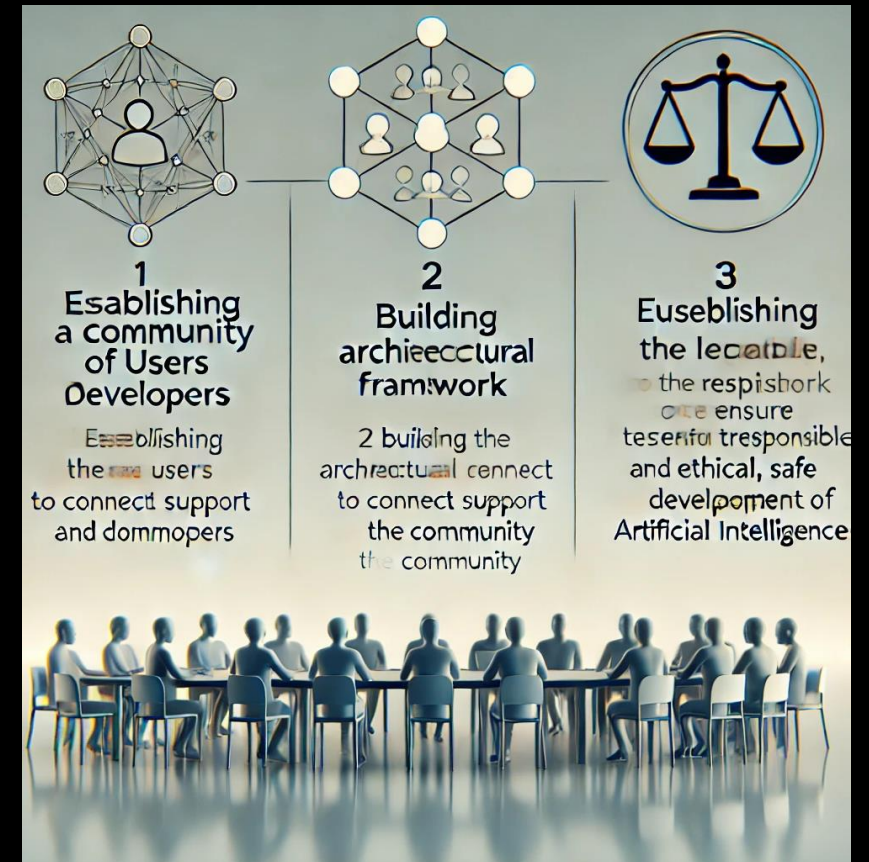
# Diffusion of Innovation Theory

- Relative Advantage
- Compatibility to Workflow
- Complexity of Use
- Triability of models
- Observability of Results

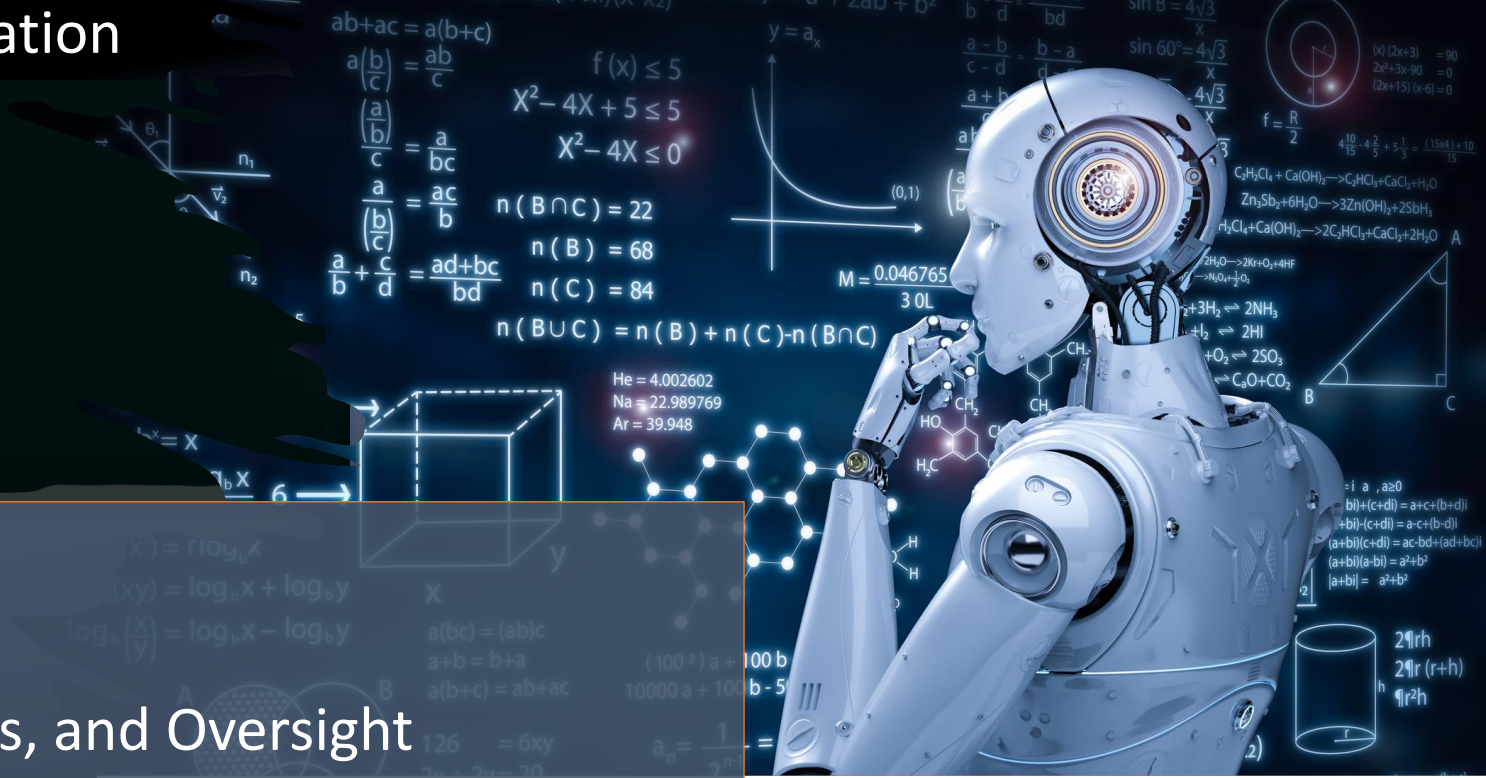


# Three major pillars of development

- Establishing a community of users and developers
- Building the architectural framework to connect and support the community.
- Establishing a legal framework to ensure trustworthy, responsible, ethical, and safe development of artificial intelligence.



# Preparation, Creation, and Implementation



## Foundational work

- Annotation ✓
- Data Structure and Use ✓
- Governance Policies, Regulations, and Oversight

## Structural needs

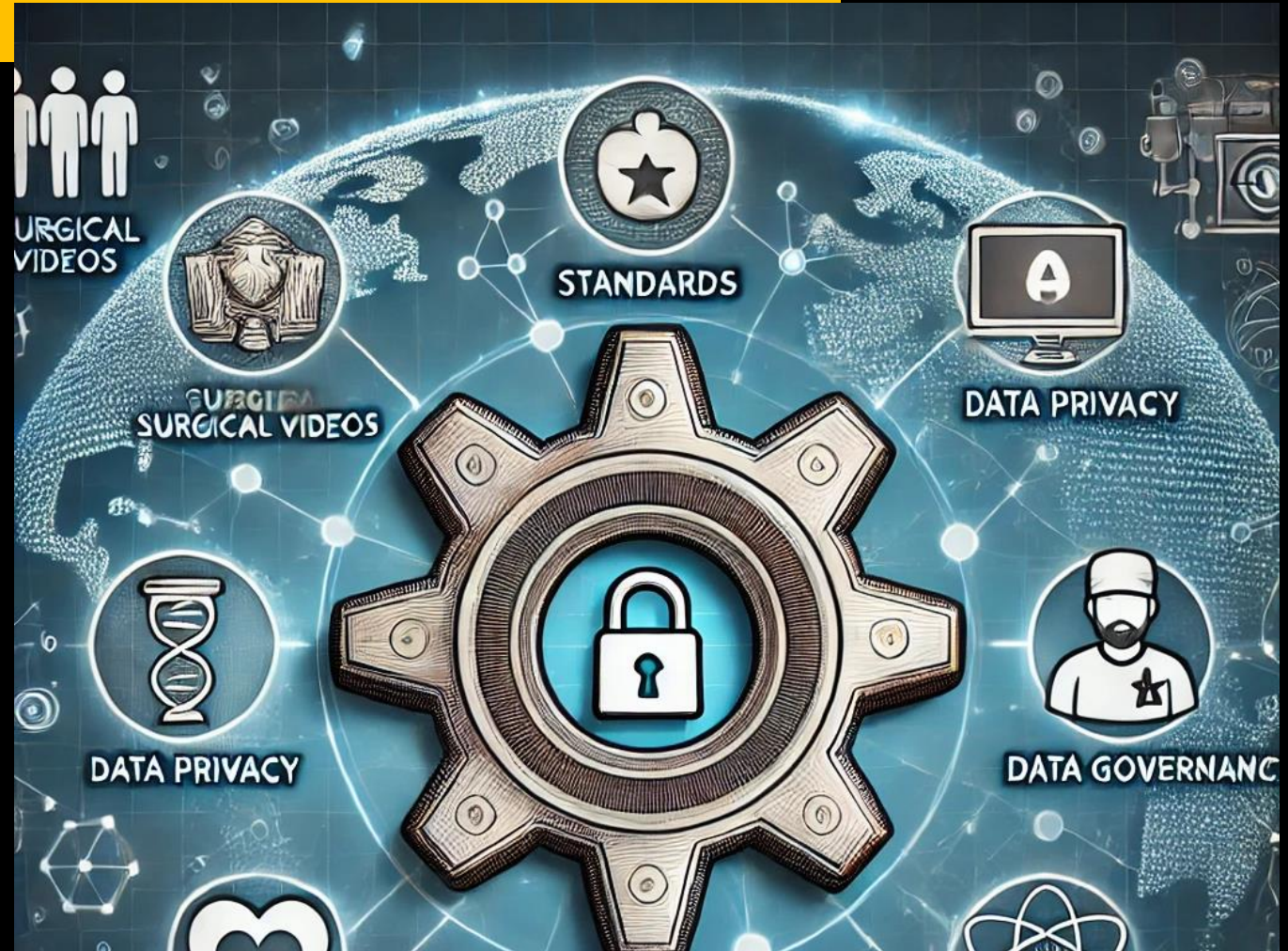
- Video Data Acquisition Framework ✓
- Creation of a Community ✓
- Management through Data Lifecycle

## Knowledge creation and dissemination

- Scientific Research ✓
- Education ✓
- Cultural Transformation



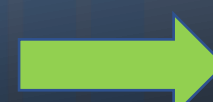
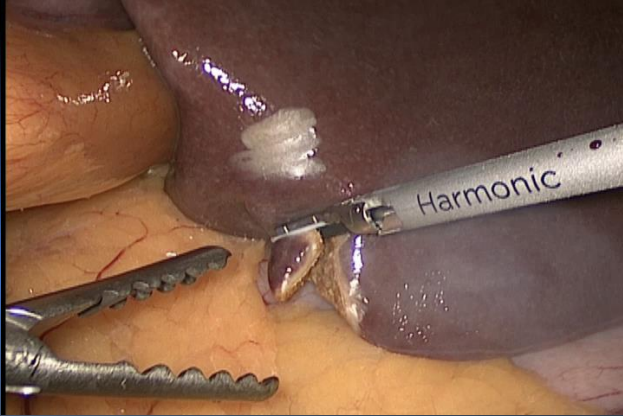
# Surgical AI Standards



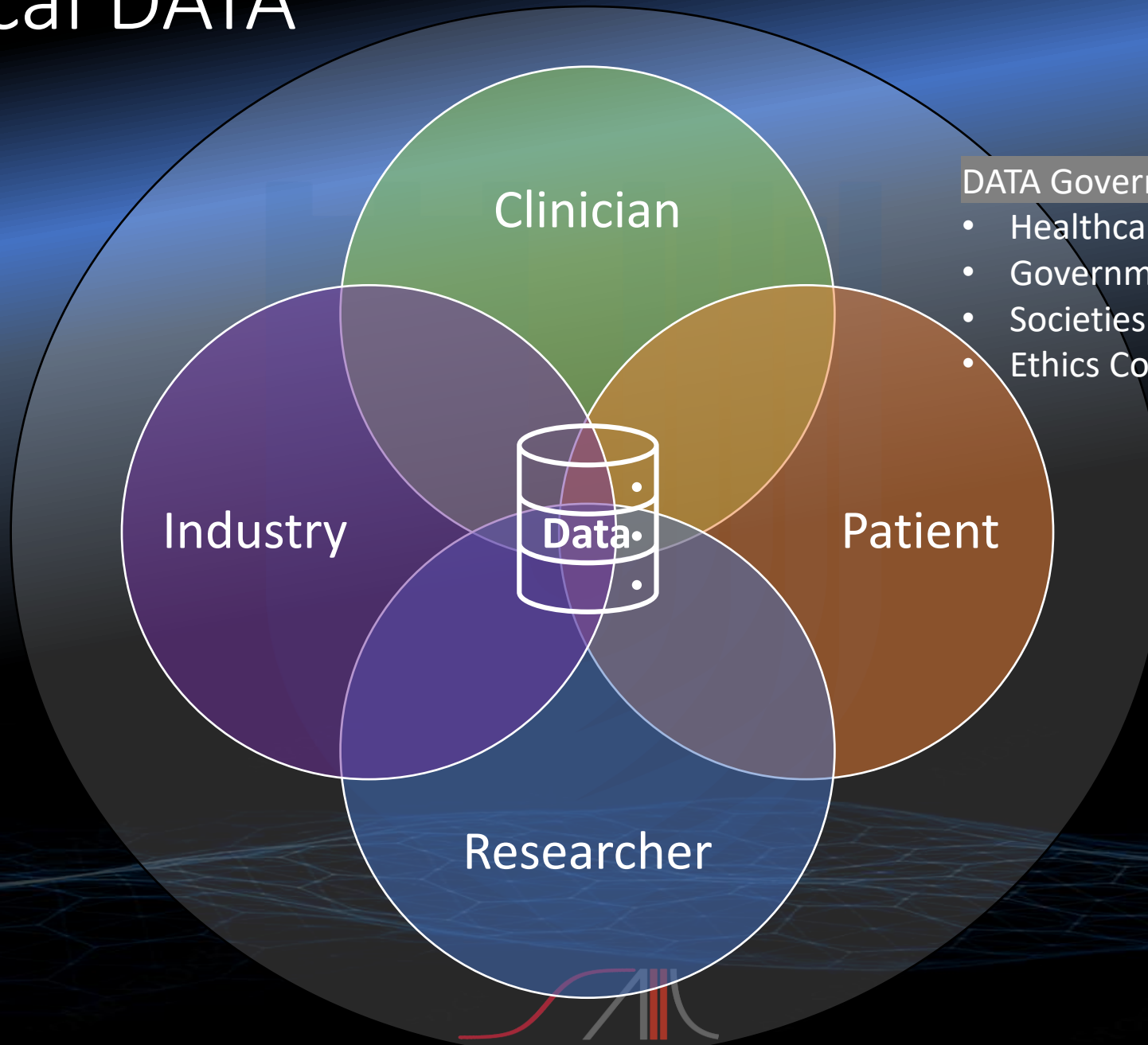
## Foundational work

- Annotation ✓
- Data Structure and Use ✓
- Governance Policies, Regulations, and Oversight

# DATA collection



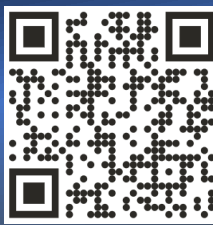
# Surgical DATA



## DATA Governance, Policies and Oversight

- Healthcare Systems
- Governments
- Societies
- Ethics Committees

# Consensus Recommendations on an Annotation Framework for Surgical Video



Surgical Endoscopy (2021) 35:4918–4929  
<https://doi.org/10.1007/s00464-021-08578-9>



## CONSENSUS STATEMENT



## SAGES consensus recommendations on an annotation framework for surgical video

Ozanan R. Meireles<sup>1</sup> · Guy Rosman<sup>1,2</sup> · Maria S. Altieri<sup>3</sup> · Lawrence Carin<sup>4</sup> · Gregory Hager<sup>5</sup> · Amin Madani<sup>6</sup> · Nicolas Padoy<sup>7,8</sup> · Carla M. Pugh<sup>9</sup> · Patricia Sylla<sup>10</sup> · Thomas M. Ward<sup>1</sup> · Daniel A. Hashimoto<sup>1</sup> · the SAGES Video Annotation for AI Working Groups

Received: 25 April 2021 / Accepted: 26 May 2021 / Published online: 6 July 2021  
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### Abstract

**Background** The growing interest in analysis of surgical video through machine learning has led to increased research efforts; however, common methods of annotating video data are lacking. There is a need to establish recommendations on the annotation of surgical video data to enable assessment of algorithms and multi-institutional collaboration.

**Methods** Four working groups were formed from a pool of participants that included clinicians, engineers, and data scientists. The working groups were focused on four themes: (1) temporal models, (2) actions and tasks, (3) tissue characteristics and general anatomy, and (4) software and data structure. A modified Delphi process was utilized to create a consensus survey based on suggested recommendations from each of the working groups.

**Results** After three Delphi rounds, consensus was reached on recommendations for annotation within each of these domains. A hierarchy for annotation of temporal events in surgery was established.

**Conclusions** While additional work remains to achieve accepted standards for video annotation in surgery, the consensus recommendations on a general framework for annotation presented here lay the foundation for standardization. This type of framework is critical to enabling diverse datasets, performance benchmarks, and collaboration.





# Annotation Framework

## Hierarchical Structure with Expandable Granularity



### Temporal Events

Phase (generic)

Step (procedure- specific)

Task (generic)

Action (generic)



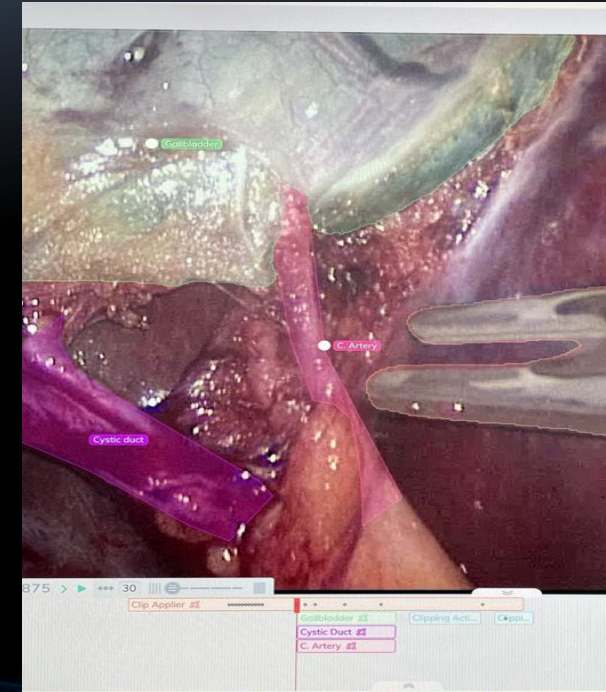
### Spatial Events

Anatomic region

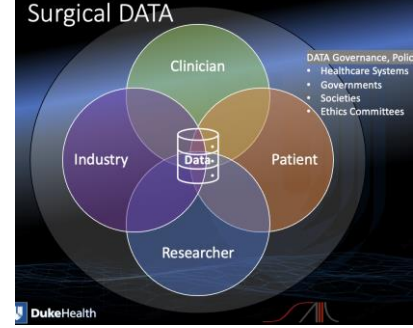
Specific anatomy

General anatomy

Tissue characteristics

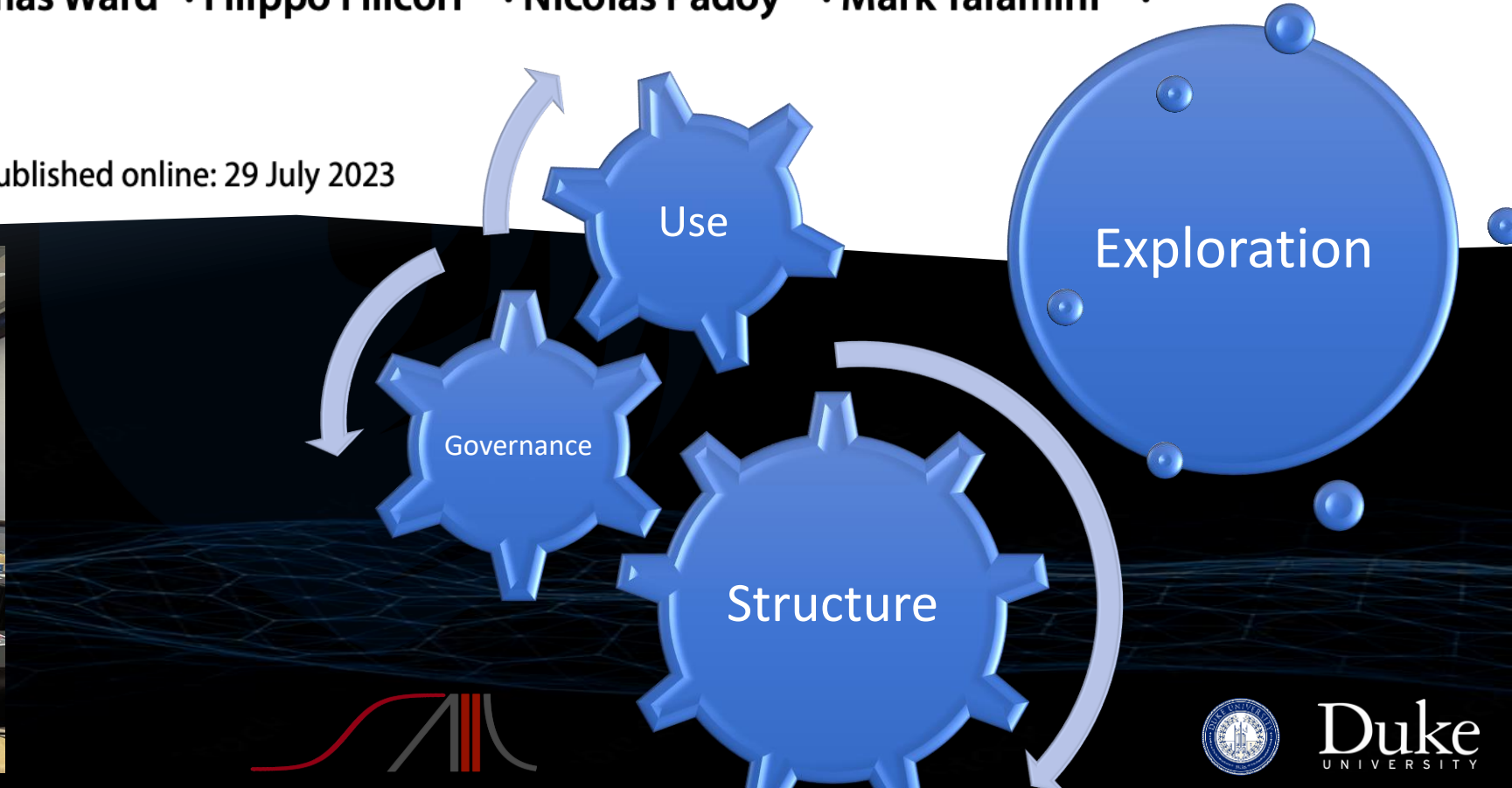


# SAGES consensus recommendations on surgical video data use, structure, and exploration (for research in artificial intelligence, clinical quality improvement, and surgical education)



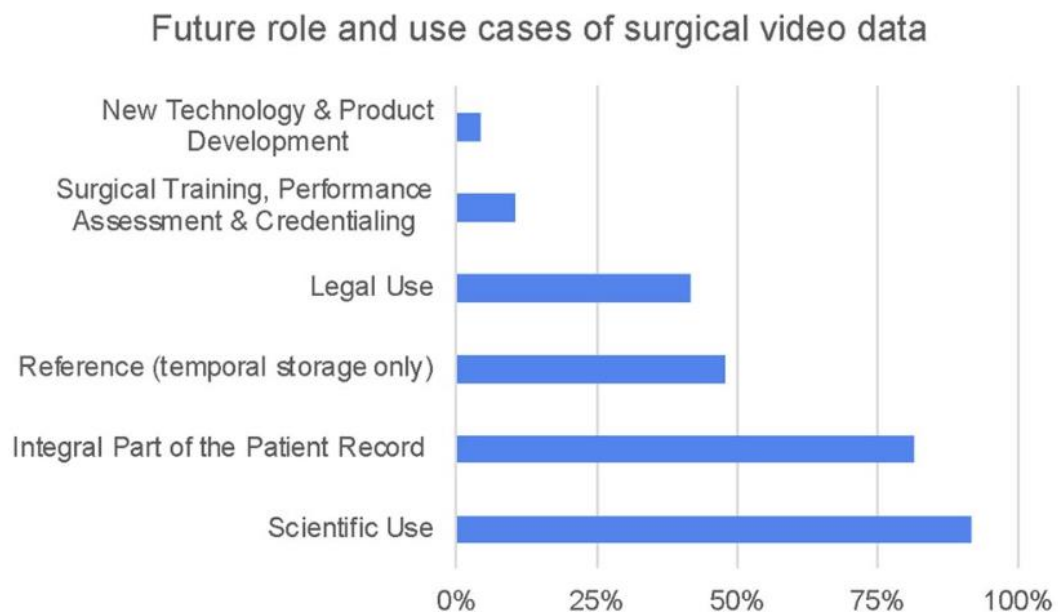
Jennifer A. Eckhoff<sup>1,2</sup>  · Guy Rosman<sup>1,3</sup> · Maria S. Altieri<sup>4</sup> · Stefanie Speidel<sup>5</sup> · Danail Stoyanov<sup>6</sup> · Mehran Anvari<sup>7</sup> · Lena Meier-Hein<sup>8</sup> · Keno März<sup>8</sup> · Pierre Jannin<sup>9</sup> · Carla Pugh<sup>10</sup> · Martin Wagner<sup>11</sup> · Elan Witkowski<sup>1</sup> · Paresh Shaw<sup>12</sup> · Amin Madani<sup>13</sup> · Yutong Ban<sup>1,3</sup> · Thomas Ward<sup>1</sup> · Filippo Filicori<sup>14</sup> · Nicolas Padoy<sup>15</sup> · Mark Talamini<sup>16</sup> · Ozanan R. Meireles<sup>1</sup>

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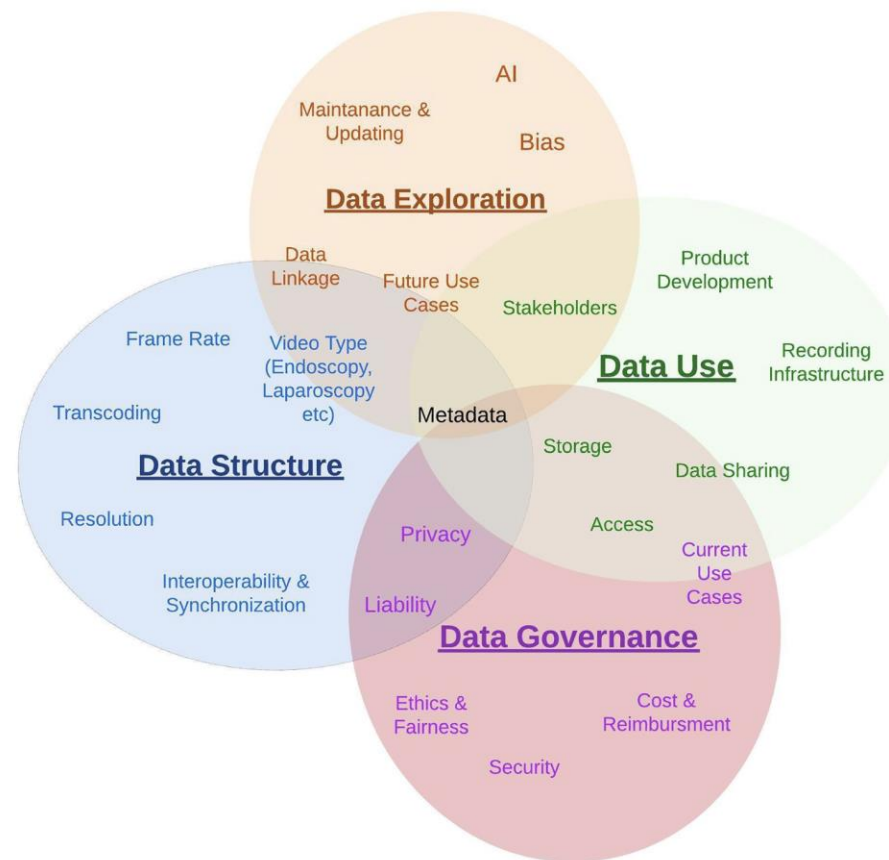


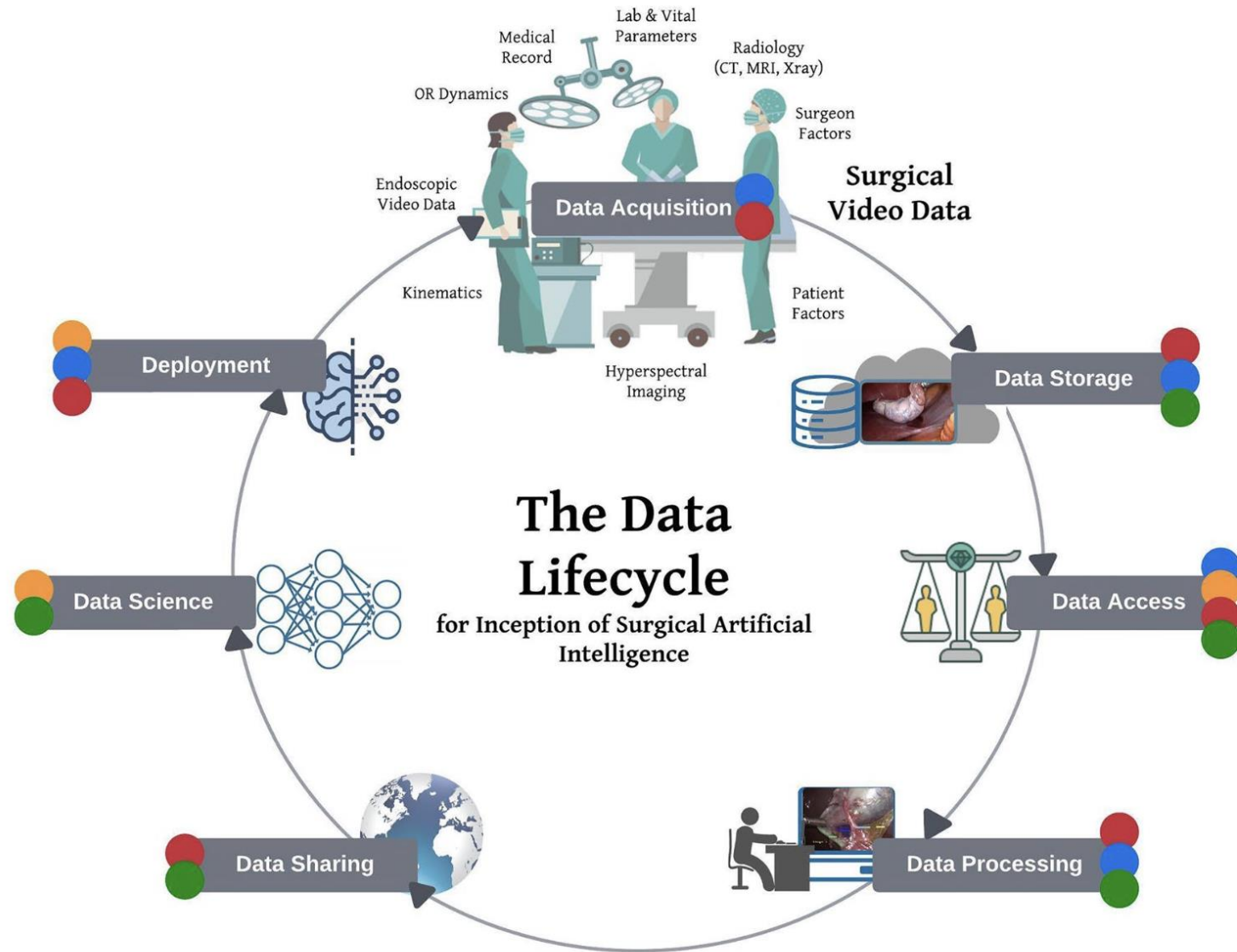
Duke UNIVERSITY

# Use Cases



**Fig. 7** Results of statement 8—future applications and use cases of surgical video data, identified by survey participants





**Fig. 1** The Data Lifecycle, highlighting stages of surgical video data en route to the creation of AI. Schematic outline of essential attributes of data architecture and infrastructure influencing current data use and future exploration and considerations for adequate governance



Community

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# Scientific efforts

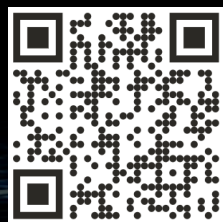
## Clinical Trials



Computer Vision Challenges



Multi-institutional collaborations



Academia and Industry partnership



Standards for Publications



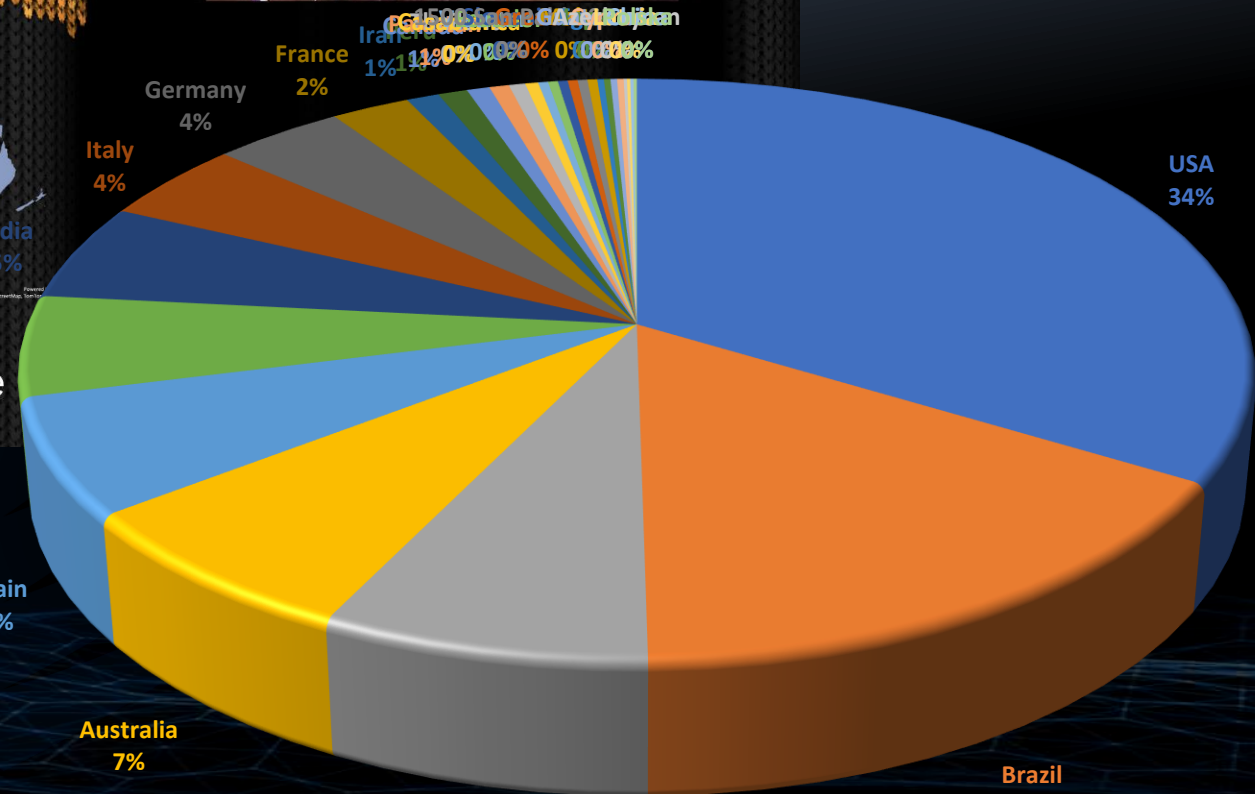
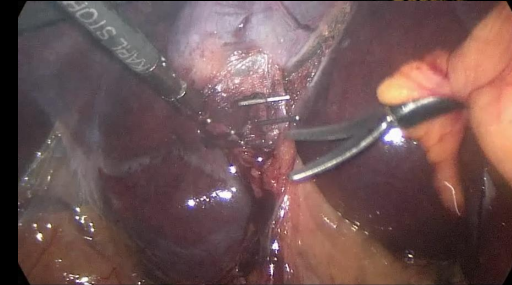
Validation Studies



Promote Diversity



# The Critical View of Safety Challenge



## DATA Donors

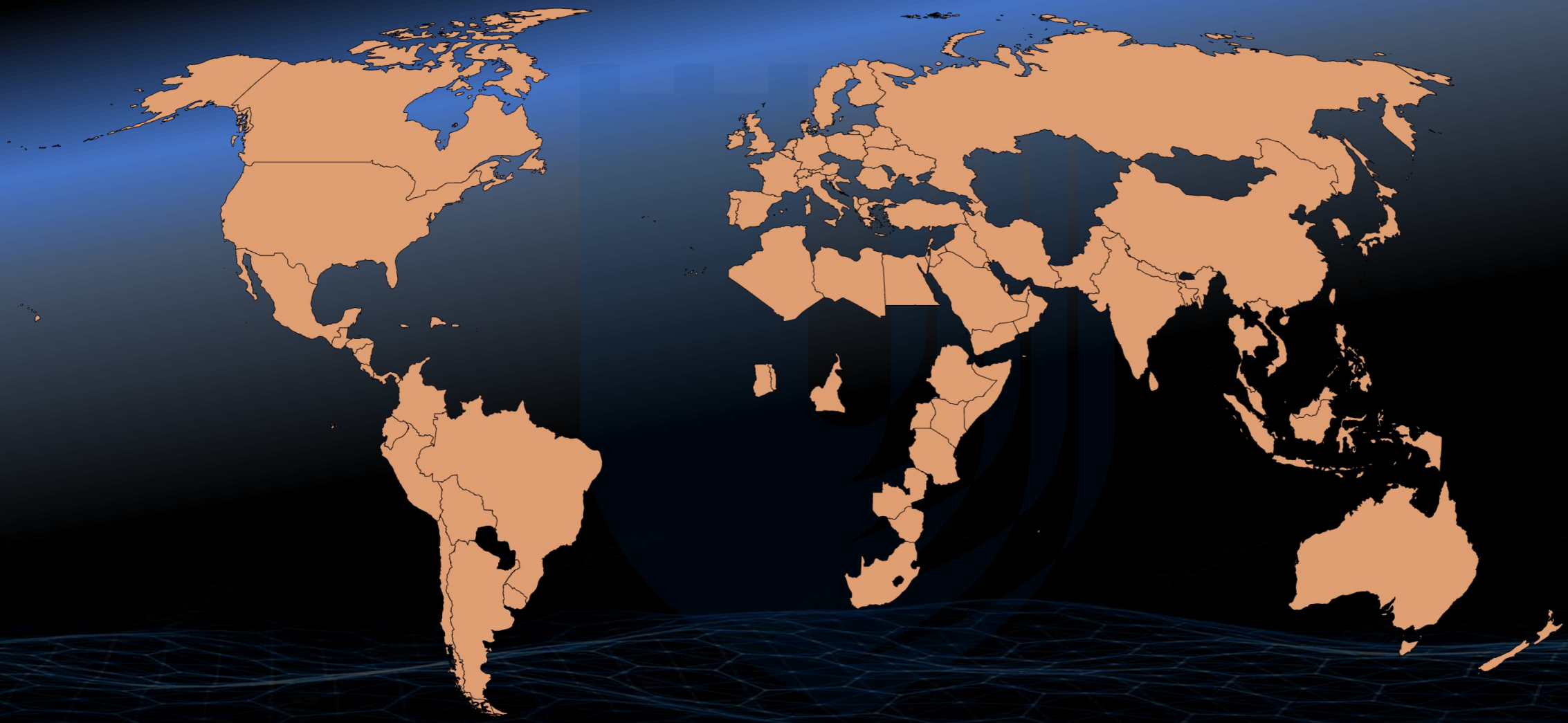
Accounts: 155  
Institutions: 64  
Countries: 57

# 1700 +

A SAGES Initiative

# 10.000 Unique Visitors

Series1





# Competitors

- 36 Teams
- 16 Countries

Grand Challenge Challenges Algorithms ... Help Sign In Register

Challenges / SAGES CVS Challenge (CVS-Challenge) / Home



## The Critical View of Safety Challenge

A SAGES Initiative

Info Statistics Join

Home The Society of American Gastrointestinal and Endoscopic Surgeons (SAGES) is hosting its first-ever biomedical challenge that aims to lay the foundations for the development of clinically meaningful and effective Artificial Intelligence (AI)



MICCAI 2024 Marrakesh MOROCCO

# Education and Training



Scientific Meetings



Dedicated Fellowships  
e.g. SAILL



Medical School  
Curriculum



Publications

[SAIIL-Net Login](#)

# Welcome to SAIIL Public

At the  
coll

At the Surgical Artificial Intelligence and Innovation Laboratory (SAIIL), we are committed to fostering a collaborative and open research community. We understand the value of sharing resources, datasets, tools, and insights with other researchers, students, and individuals interested in the field of surgical AI. To accelerate innovation and improve patient care worldwide, we are in the process of gradually making these resources available to the public.

earch  
duate

## Education and Training

Scientific Meetings

Dedicated Fellowships  
e.g. SAIIL

Medical School Curriculum

Publications



Surgery & AI

Projects

SAILL-Net

Work with us

Team

Sponsorship

Surgery & AI

Projects

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Team

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MASSACHUSETTS  
GENERAL HOSPITAL



DukeHealth

Artificial Intelligence and Innovation Laboratory



Welcome to SAILL Network



Surgery & AI

Projects

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Get Involved

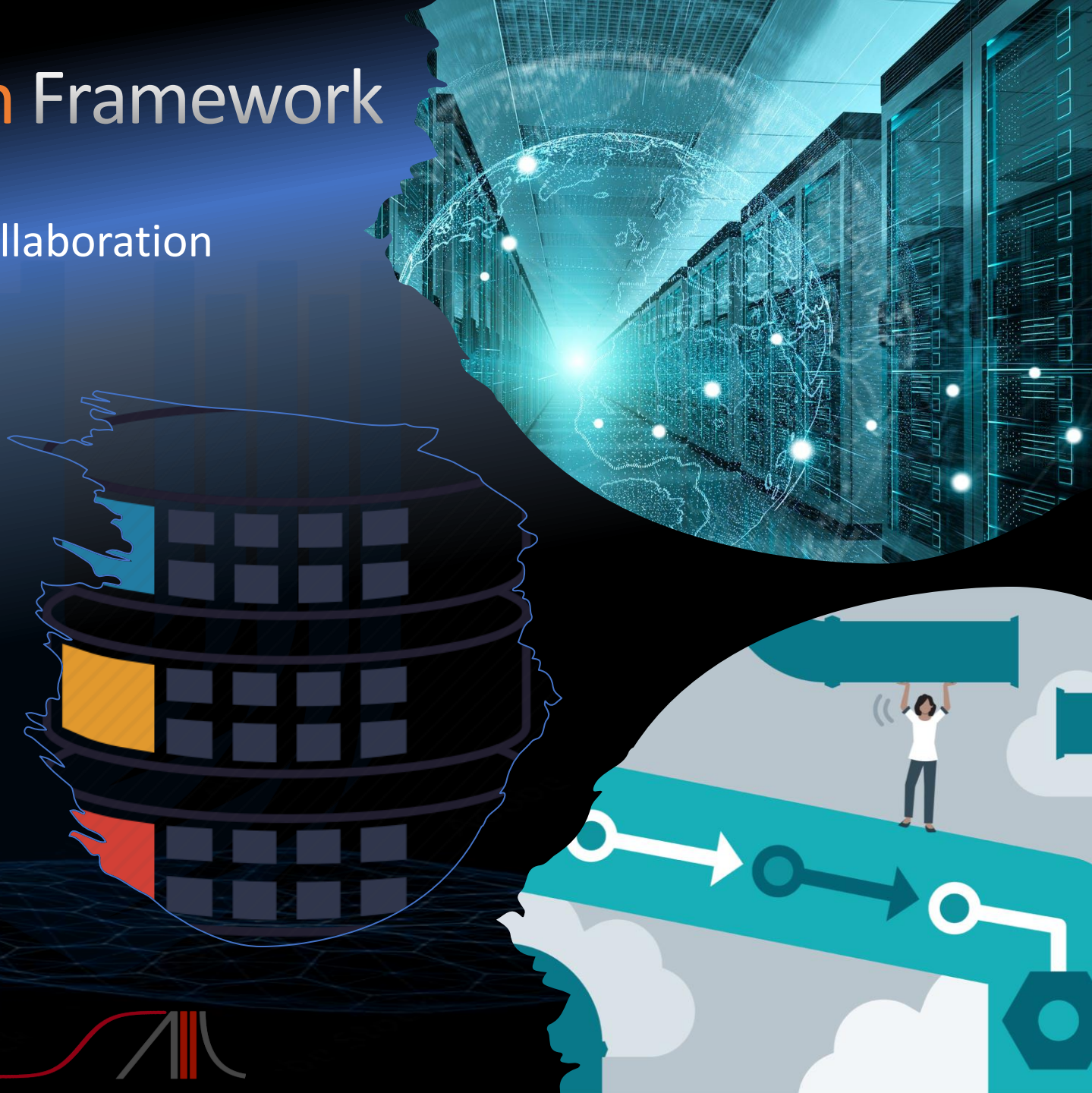
SAILL-Net Login

Welcome to SAILL Public



# Surgical Operating System Framework

- Open Access Model to Promote Collaboration
- Standardization
  - Annotation
  - Data Structure
- Clear Policies and Regulations
- Transparency and Oversight
- Address Ownership Issues

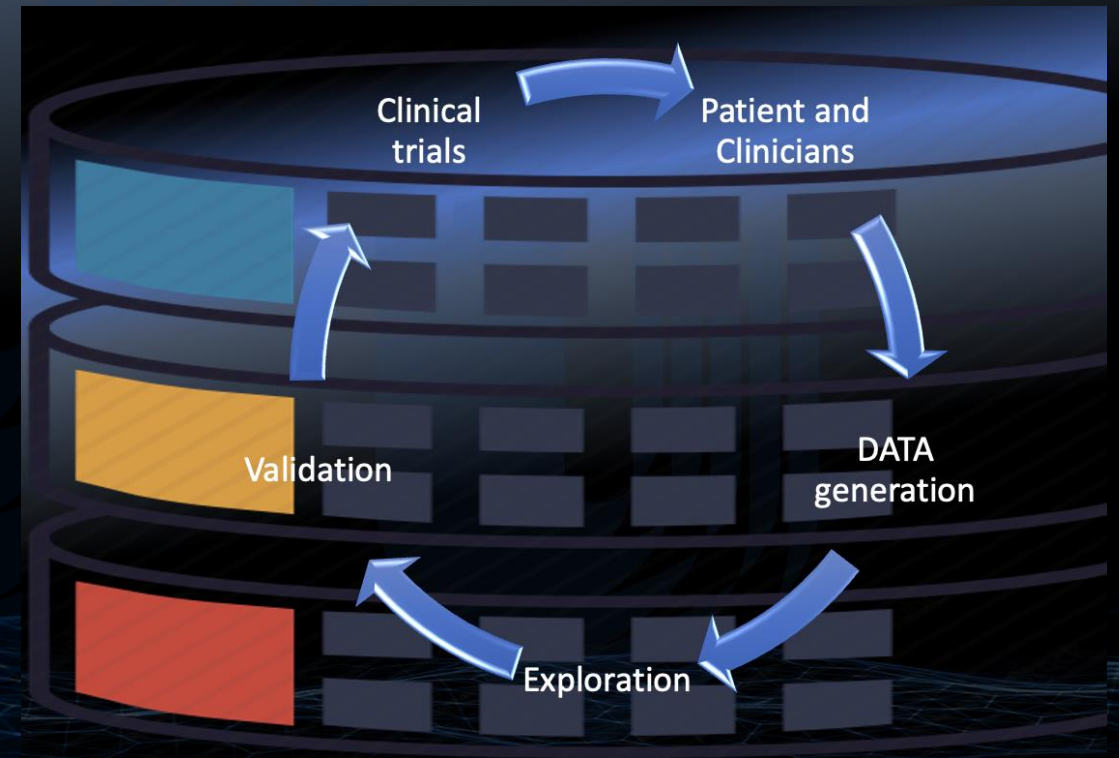
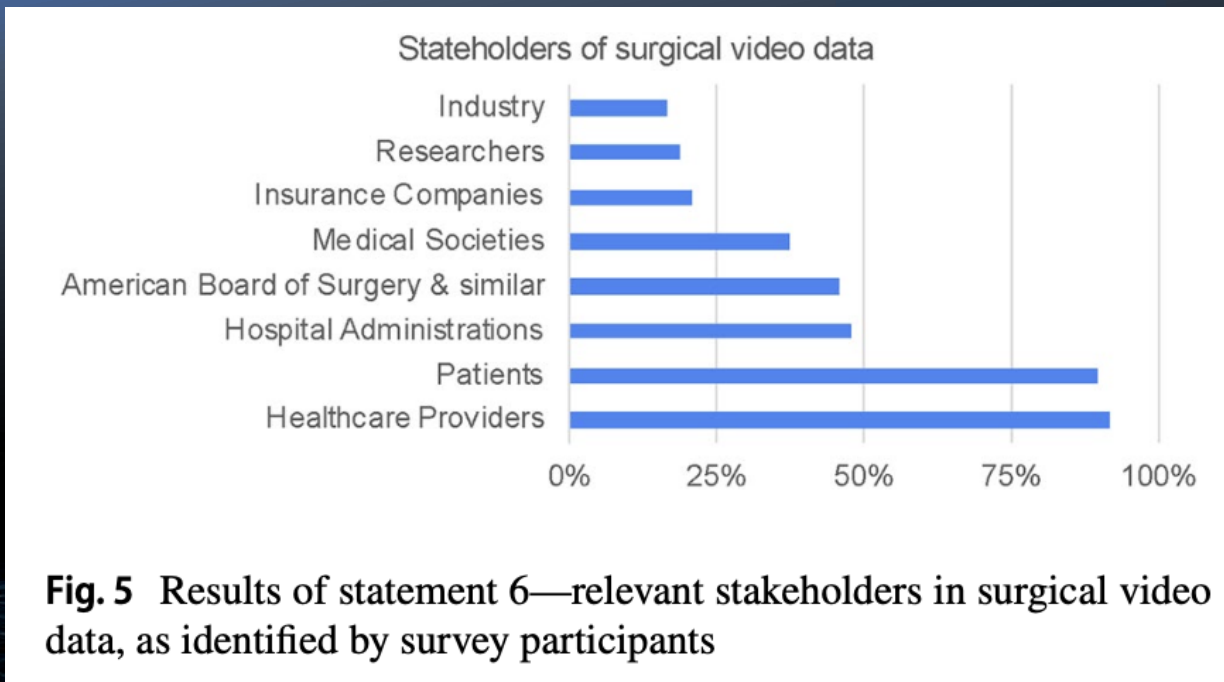




# Surgical AI Governance Stakeholders

## Regulations, Policies and Oversight

Data governance is a principled approach to managing data during its life cycle, from acquisition to use to disposal.



# Trustworthy and Responsible AI Network (TRAIN)

New consortium of healthcare leaders announces formation of Trustworthy & Responsible AI Network (TRAIN), making safe and fair AI accessible to every healthcare organization

March 11, 2024 | Microsoft Source

**AMSTERDAM — June 17, 2024** — Monday, at [HLTH Europe](#), the Trustworthy & Responsible AI Network (TRAIN), a consortium of healthcare leaders, announced its expansion to Europe with the objective to help organizations in the region operationalize responsible AI through technology-based guardrails. Organizations that have come together to form the European TRAIN include [Erasmus MC](#) (the Netherlands), [HUS Helsinki University Hospital](#) (Finland), [Sahlgrenska University Hospital](#) (Sweden), [Skåne University Hospital](#) (Sweden), [Universita Vita-Salute San Raffaele](#) (Italy), and [University Medical Center Utrecht](#) (the Netherlands), with Microsoft as the technology enabling partner. [Foundation 29](#), a nonprofit organization that aims to empower patients and transform healthcare through data-driven initiatives and innovative technologies, has also joined European TRAIN. The network is open to other healthcare organizations in Europe interested in joining.

# Designing a Surgical Operating System (S.OS)

A Blueprint for the Future of Surgery





- **Ethical and Trustworthy Data Generation, Model Development, and Validation**
- **Addressing the Critical Need for Benchmarking and Ethical Considerations**
  - **Data Privacy Generation**
  - **AI Model Development**
  - **Validation**
  - **Data Privacy and Governance**

S.OS is a conceptual framework that aims to seamlessly integrate surgical teams, operating rooms, patient data, and devices.

S

OPERATING SYSTEM

Standardization

Efficiency

Safety



# S. OS Features



Efficiency &  
Scheduling



Data & Safety  
Management



Technological  
Integration



User Interface  
& Experience



Security &  
Access Control



Communication  
& Collaboration



Analytics &  
Monitoring



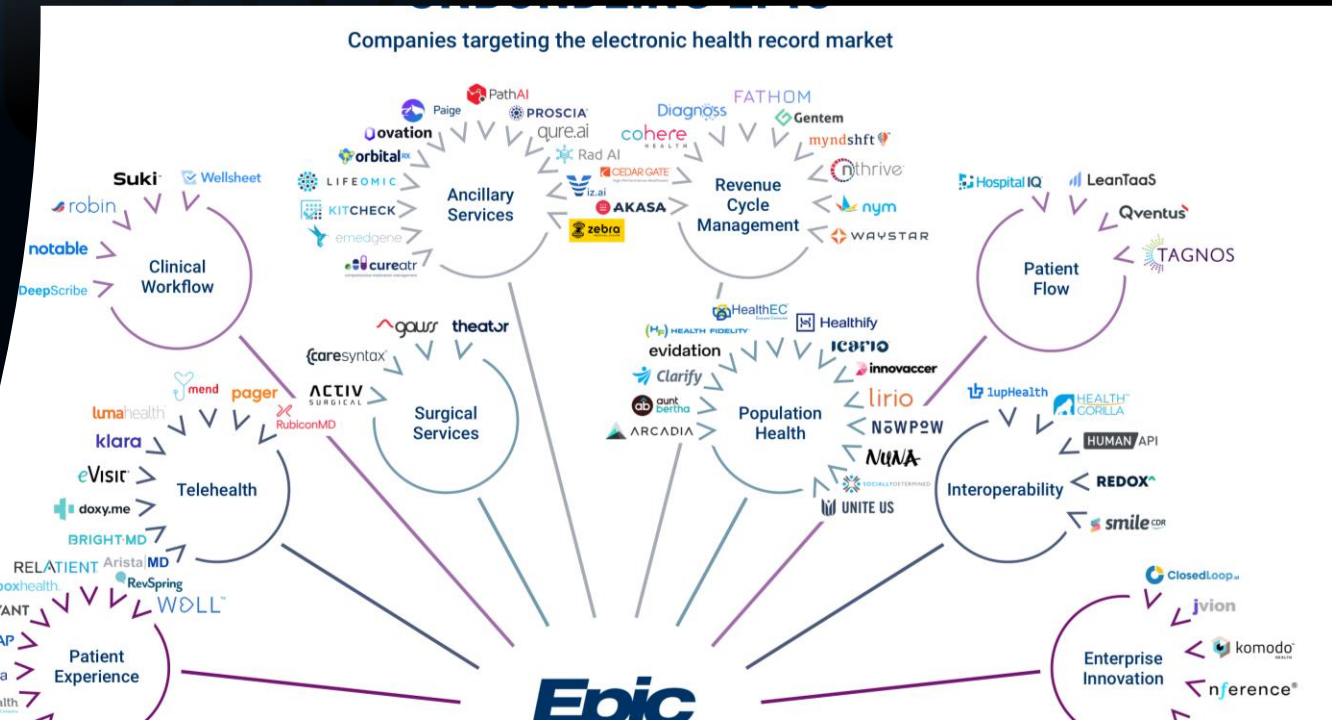
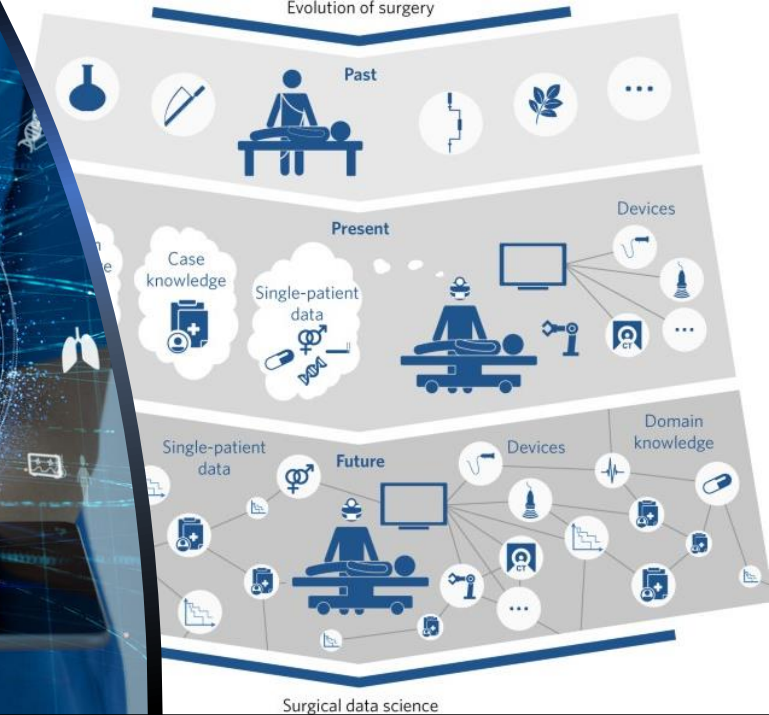
Utilities &  
Special Features



# Data Generation and Management

Organizing Critical Information

Maintenance of structure data



# Model Development

- Data training
- Algorithm optimization
- Model testing
- Continuous learning and improvement

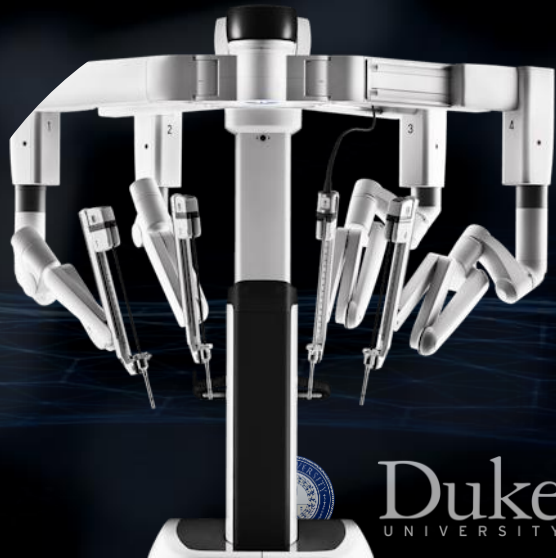


# Trustworthiness of AI Systems Assurance:

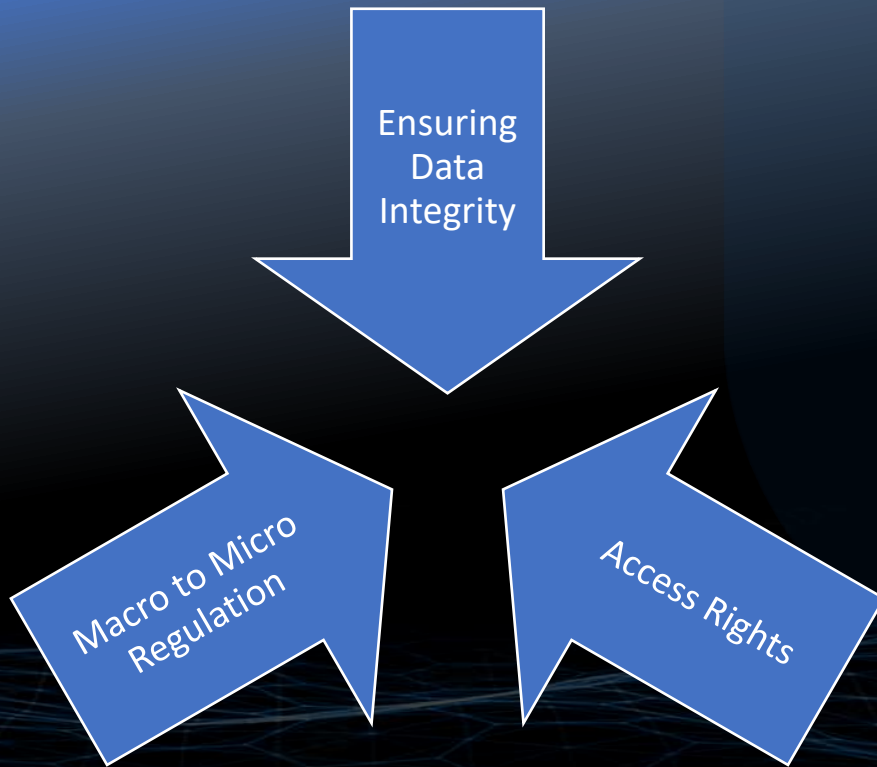
- Thorough validation
- Benchmarking
- Ensure AI systems are accurate, safe, and ethical.



# Solution Management



# Security and Access Control





# Surgical Video Foundation Models

These models serve as a fundamental base, trained on large datasets, and can be adapted to a variety of surgical tasks such as:

- Video analysis
- Complication prediction
- Real-time guidance
- Automation

# Annotation

## Temporal Hierarchy

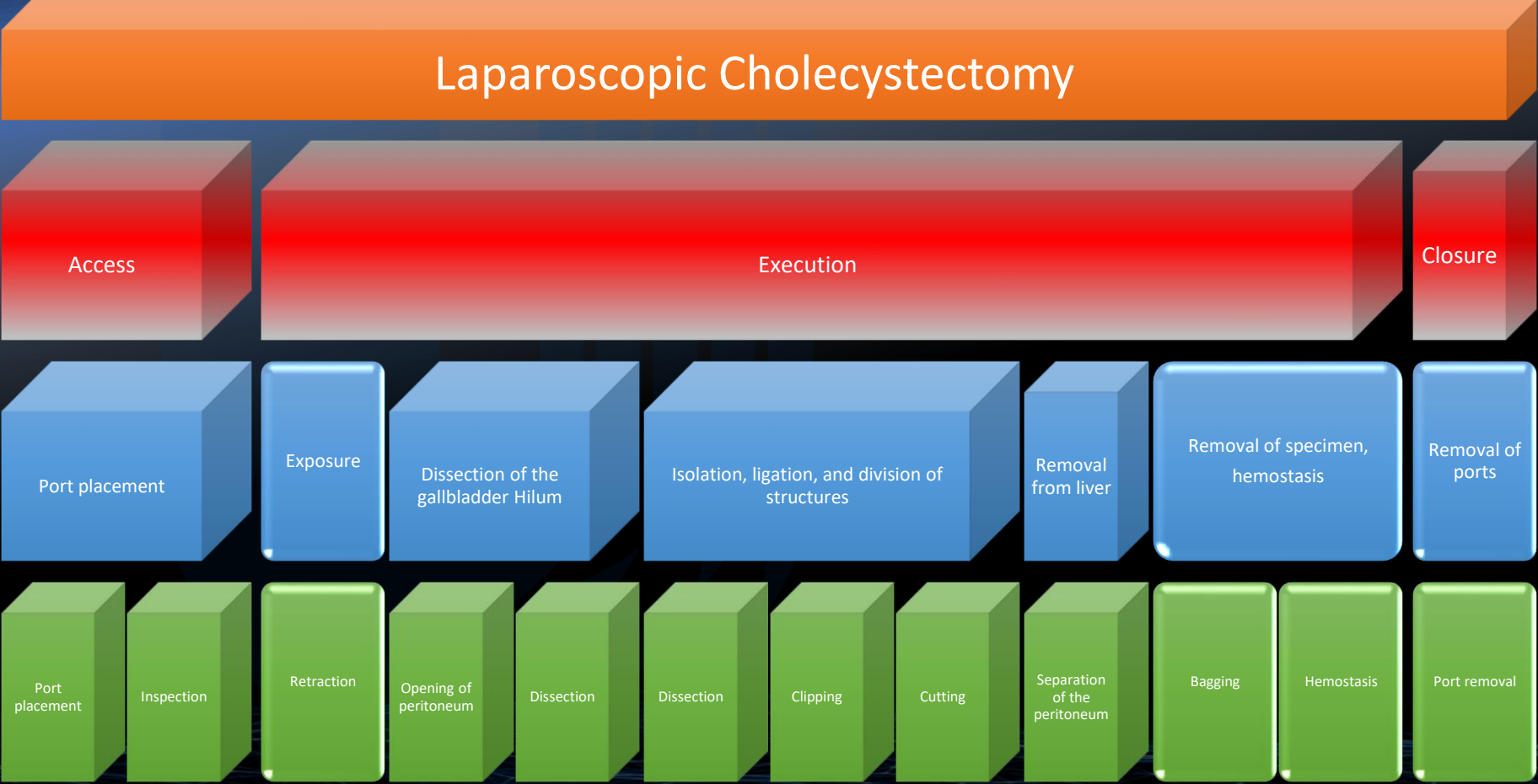


Operation

PHASES

Steps

Tasks and Actions



# Annotation

## Spatial Hierarchy

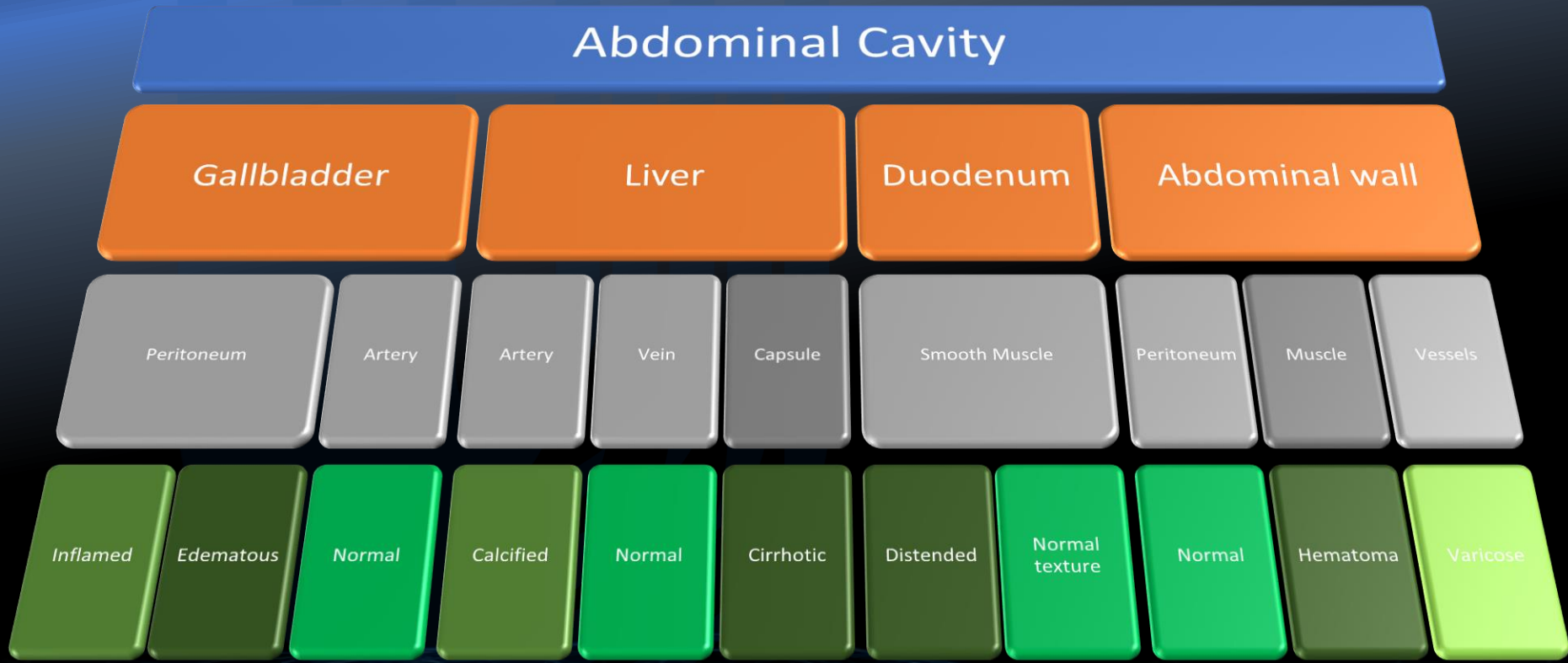


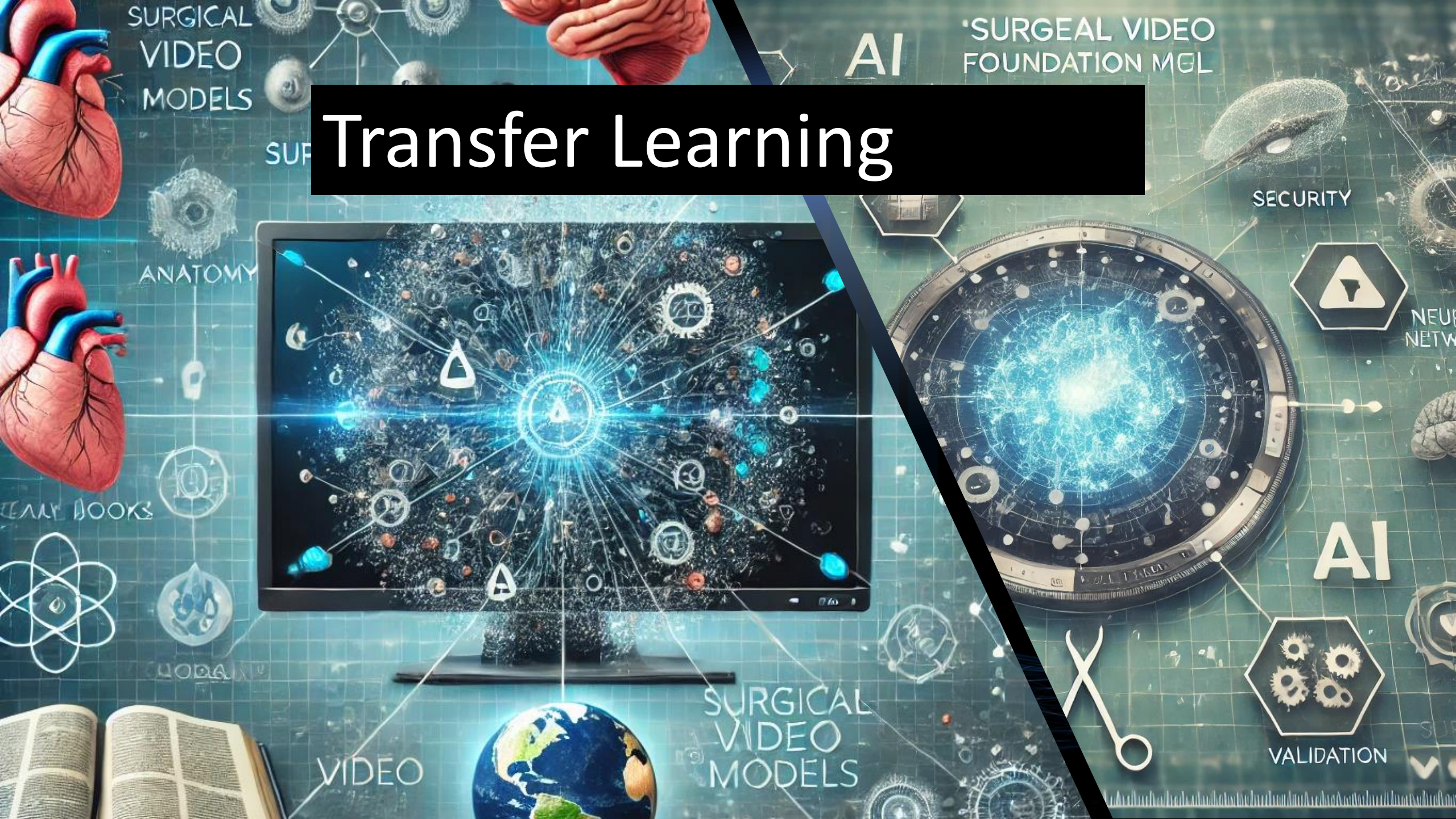
Reginal Anatomy

Specific anatomy

General anatomy

Tissue characteristics





SURGICAL  
VIDEO  
MODELS

AI

SURGICAL VIDEO  
FOUNDATION MGL

# Transfer Learning



SECURITY



NEURAL  
NETWORK

AI



VALIDATION



SURGICAL  
VIDEO  
MODELS

VIDEO



TECHNICAL BOOKS

ANATOMY



MODERN

SUR



S. OS



## Cognitive Augmentation

- Information
- Guidance
- Safety
- Operational Efficiency



# Analysis of intraoperative video



Experimental Phase

Normal case

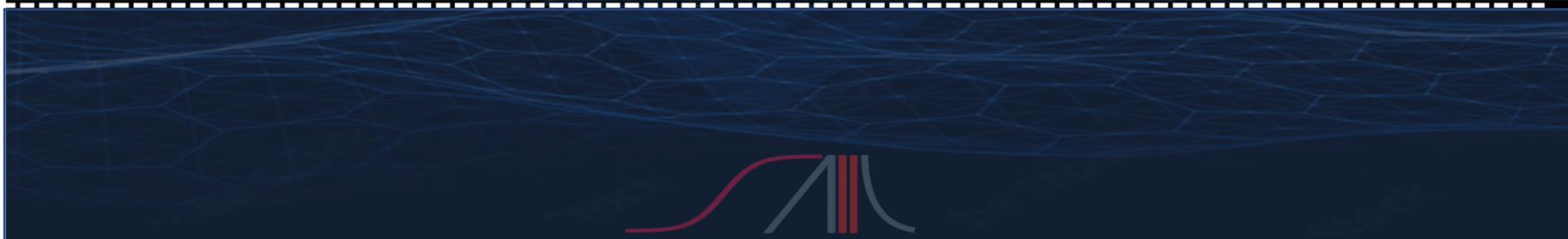


Normal  
Surgical Fingerprint

Normal Range

Access	Exposure	Dissection	Stapling	Bagging	Leak Test	Closure
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Deviation



# Real-time Analysis and Deviation Detection



Experimental Phase

Abnormal case



Abnormal  
Surgical Fingerprint

© 2024 Ozanan R. Meireles. All rights reserved. For more information, visit: [www.saill.org](http://www.saill.org)

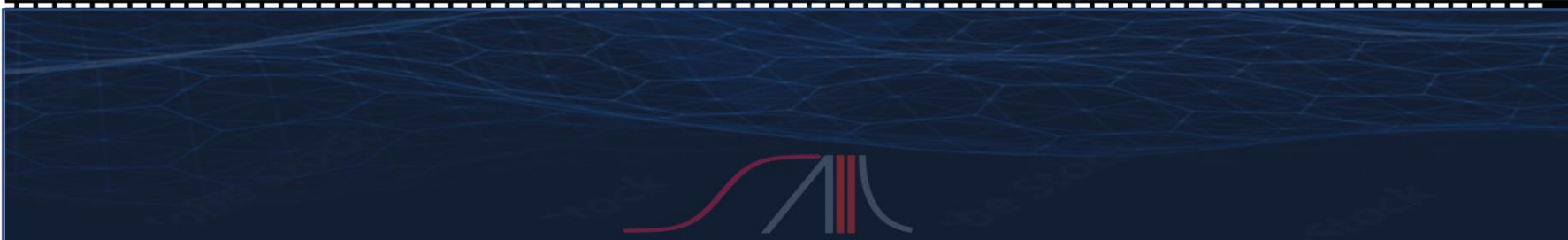


## Detecting Deviations

Normal Range

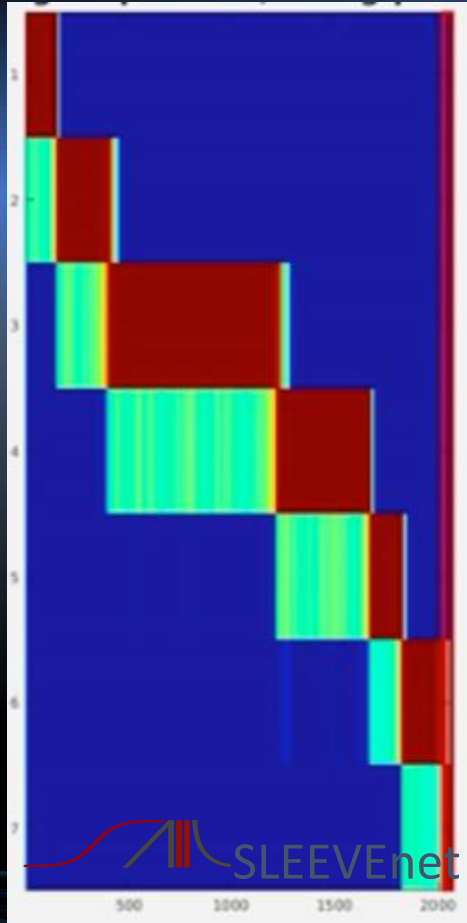
Access	Exposure	Dissection	Stapling	Bagging	Leak Test	Closure
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Deviation

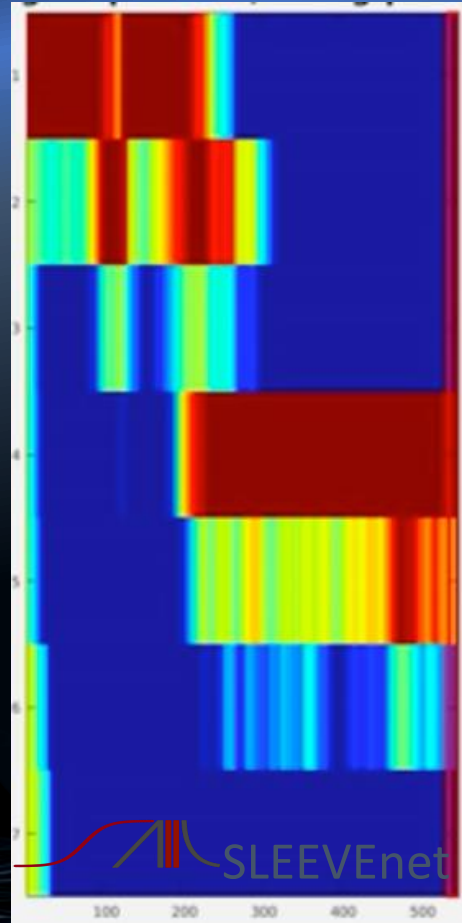


# Surgical Fingerprint – Sleeve Gastrectomy

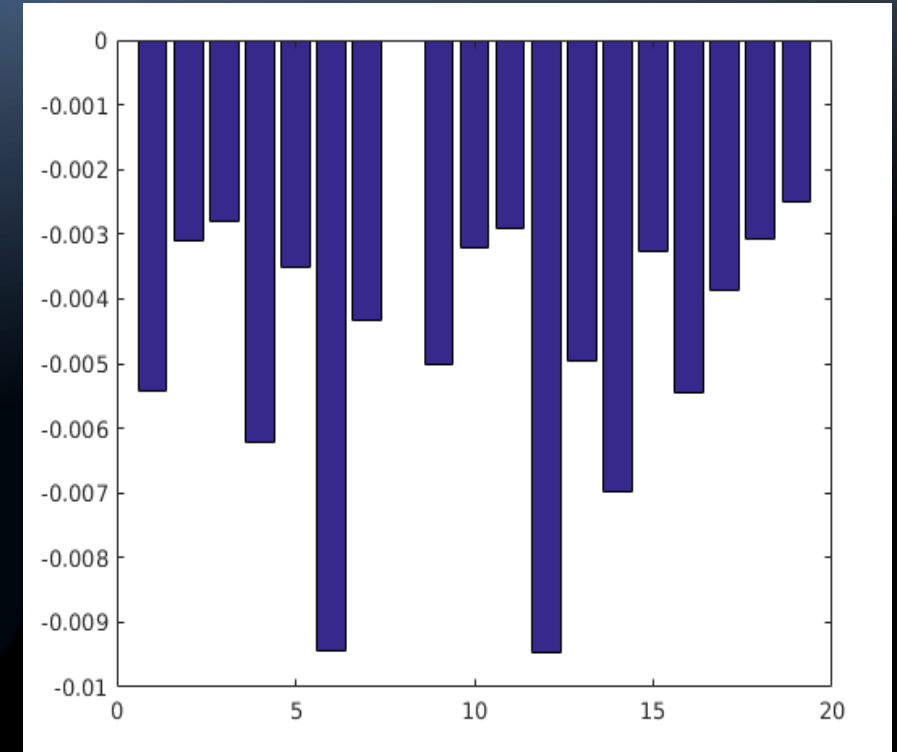
Case A



Case B



Normalized Cumulative log Probability

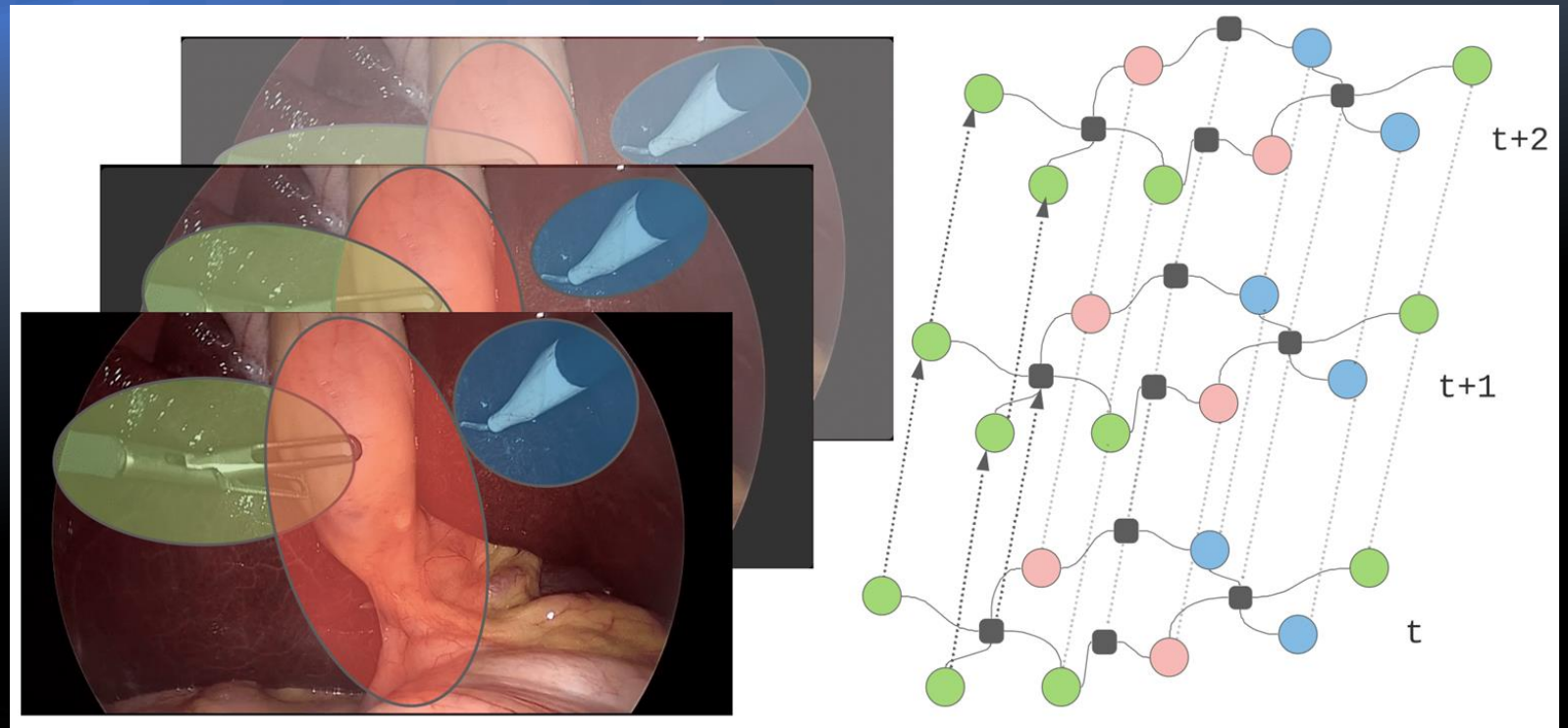




# Knowledge Graph in Surgery

which can be used for

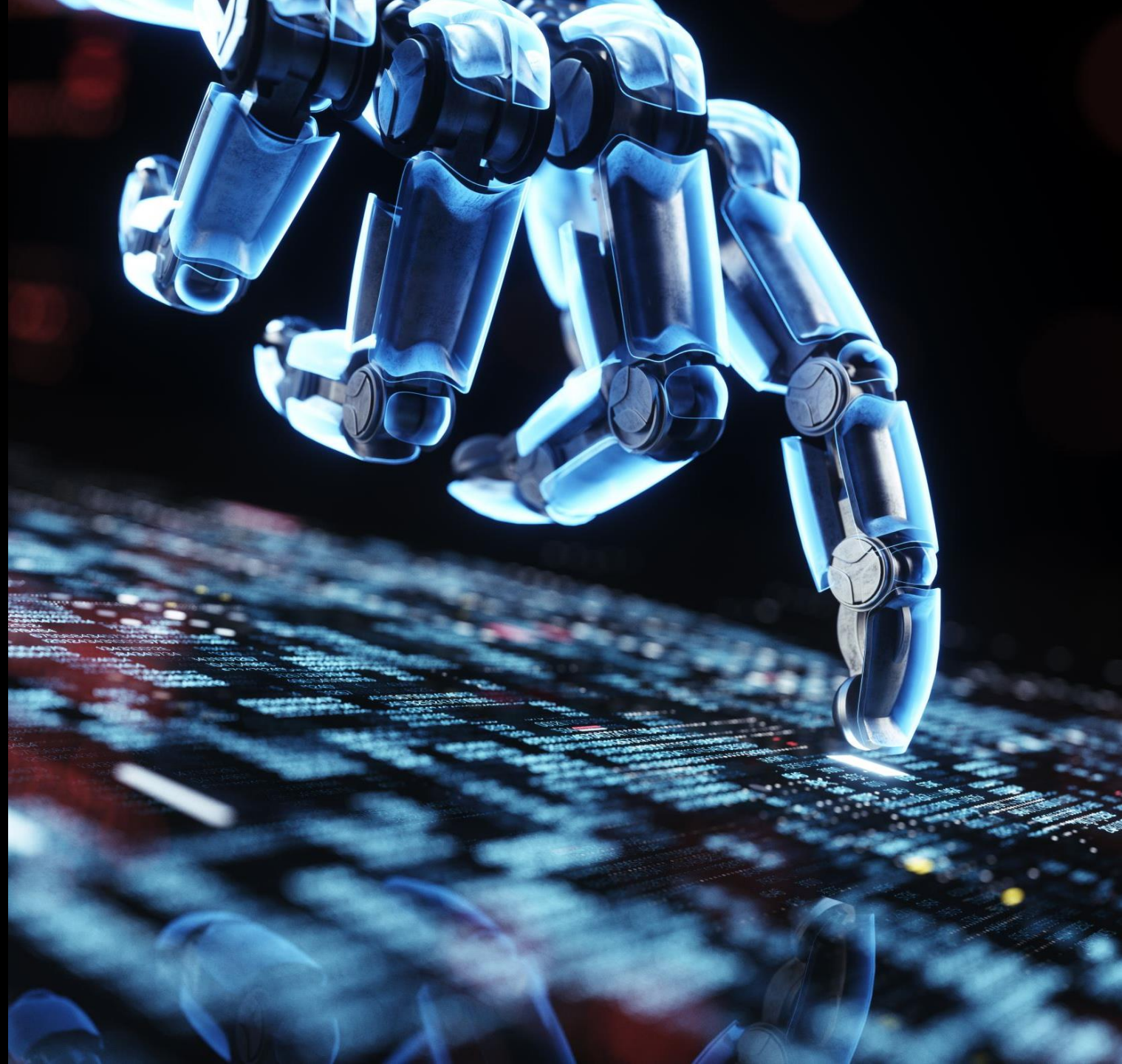
- Interactions of objects in surgery
  - Tool-action-tissue interactions.
- Relations of abstract concepts
  - Parkland scale and its components



Concept Graph Neural Networks for Surgical Video Understanding  
**Y. Ban**, J. Eckhoff, T. M. Ward, D. A. Hashimoto, O.R. Meireles, D. Rus, G. Rosman  
Under revision to IEEE Transactions on Medical Imaging 2022

Utilities and Apps

Automation



# Potential Current Applications - S.OS Apps

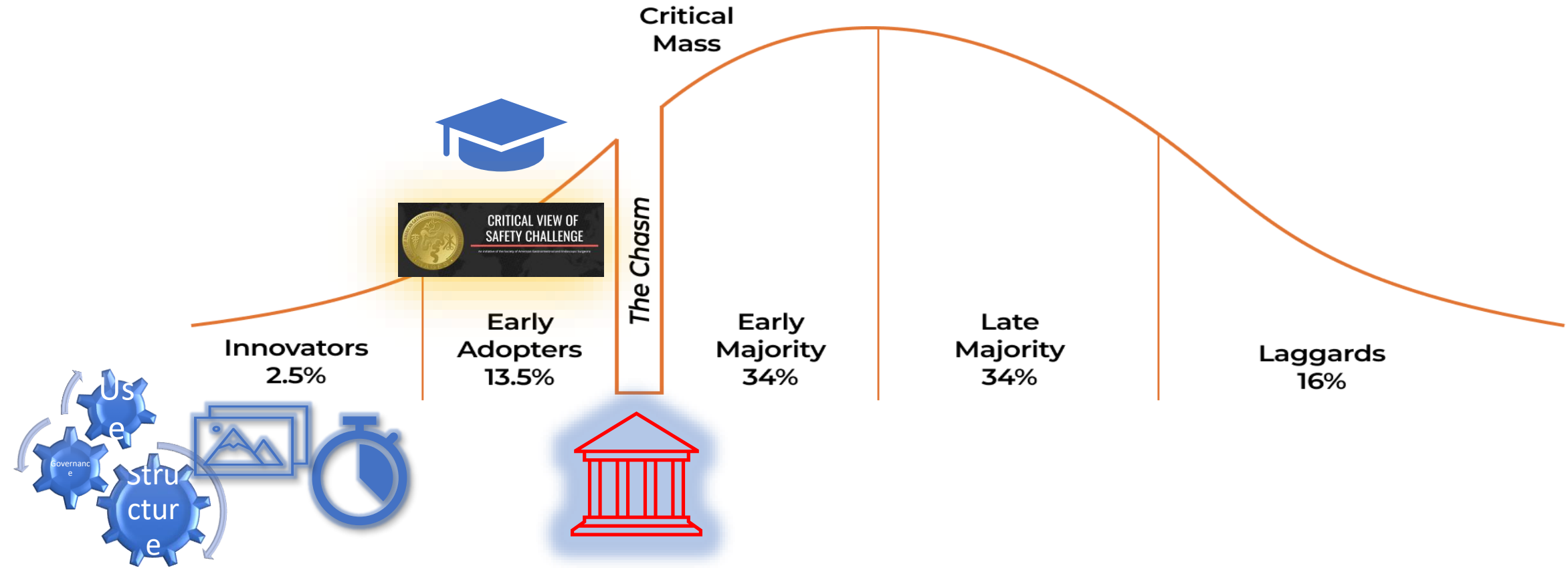
- Attending Notification System
- Operative Report Generation
- Billing
- Compliance Monitoring
- Resource Prediction, and Allocation
- **Tele-mentoring**
- **Operative Case Length Prediction**
- **Attention Awareness**



Future Steps



# Surgical AI Development



# Cultural Transformation



SHARING DATA



SHARING KNOWLEDGE

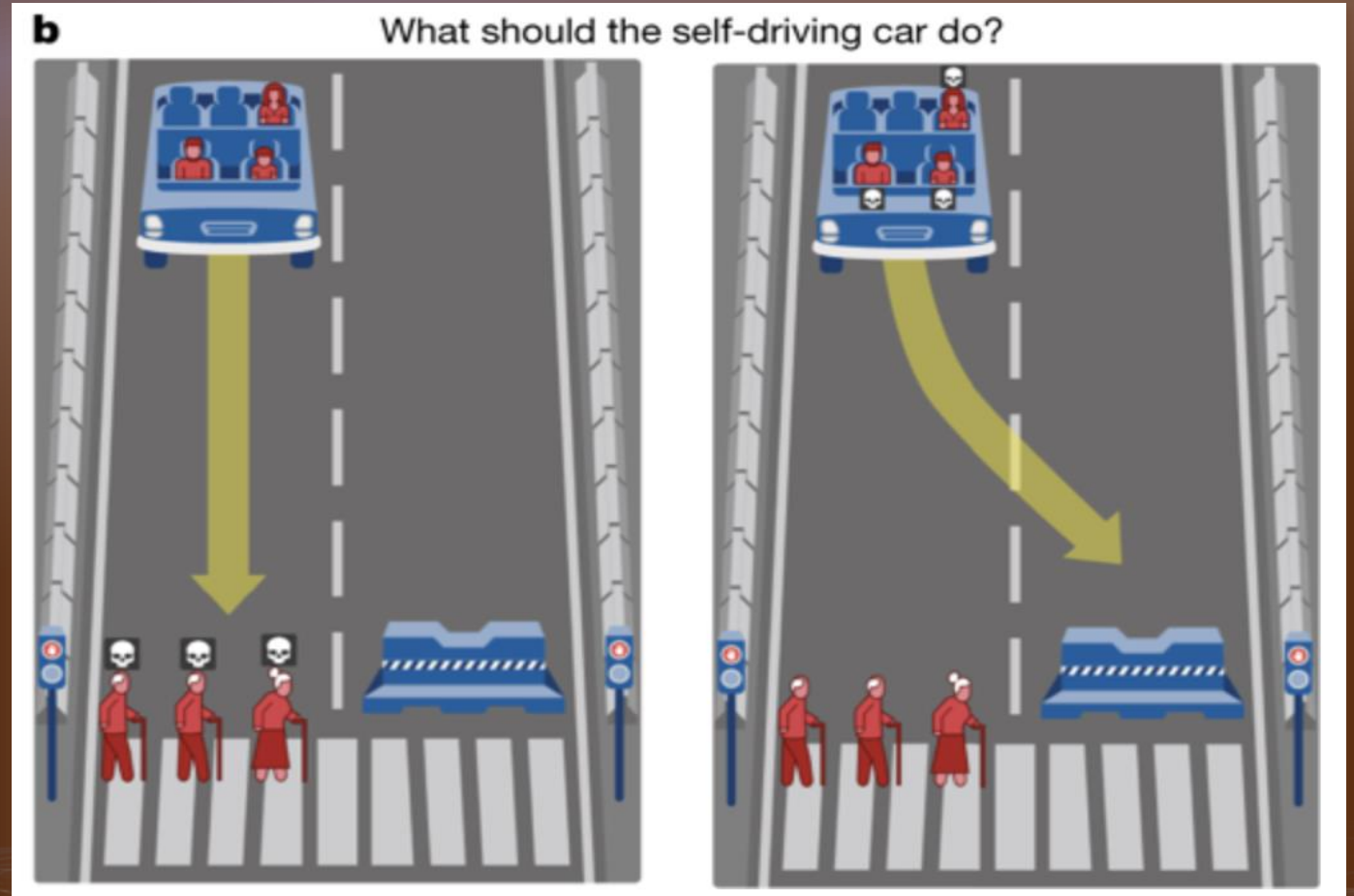


CULTURAL DIFFERENCES

# Ethical considerations

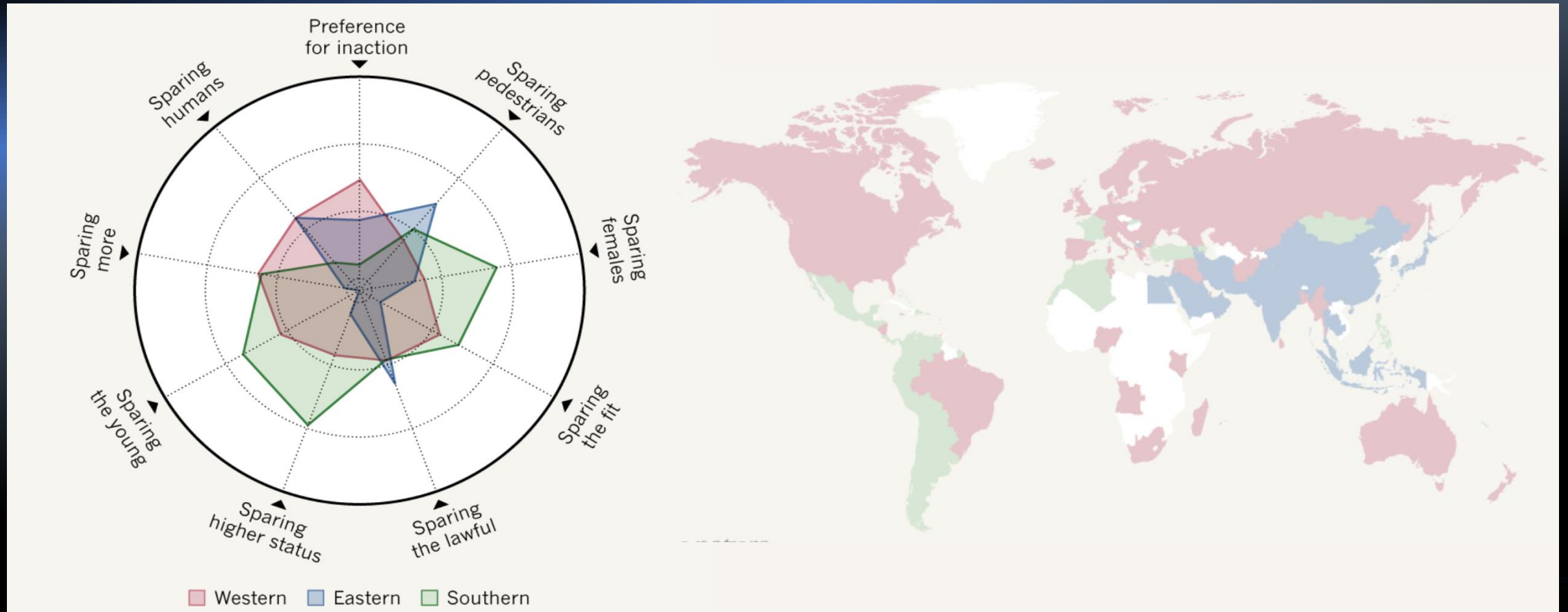
## The Moral Machine

<http://moralmachine.mit.edu>



Awad et al. 2018. *Nature*

# Ethical according to whom?



<http://moralmachine.mit.edu>



# Surgical AI

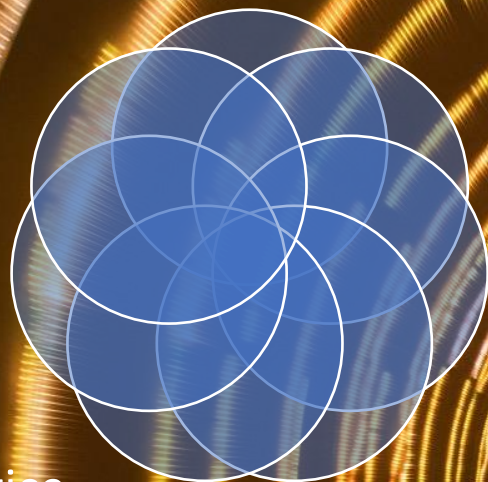


Surgeons

Patients

Industry

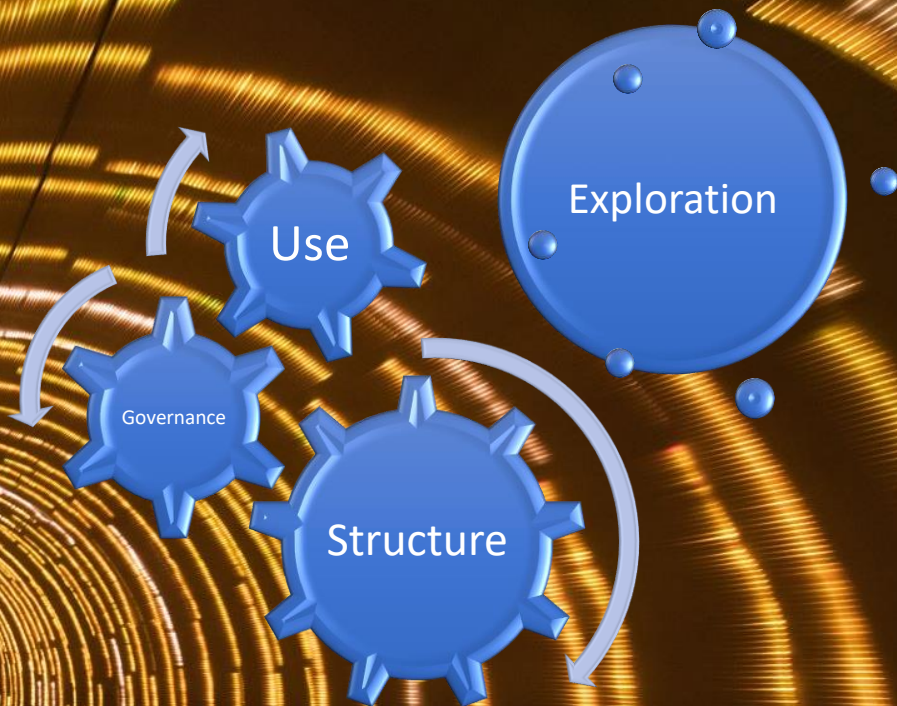
Malpractice  
Insurance  
companies



Government  
officials

Hospitals

Payers



A few years from now .....

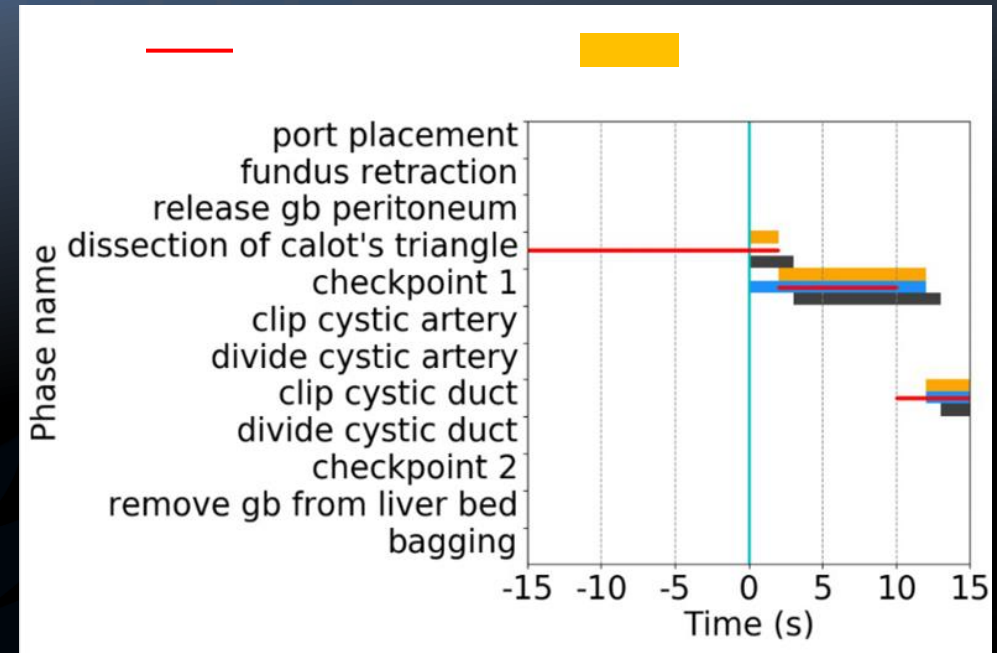
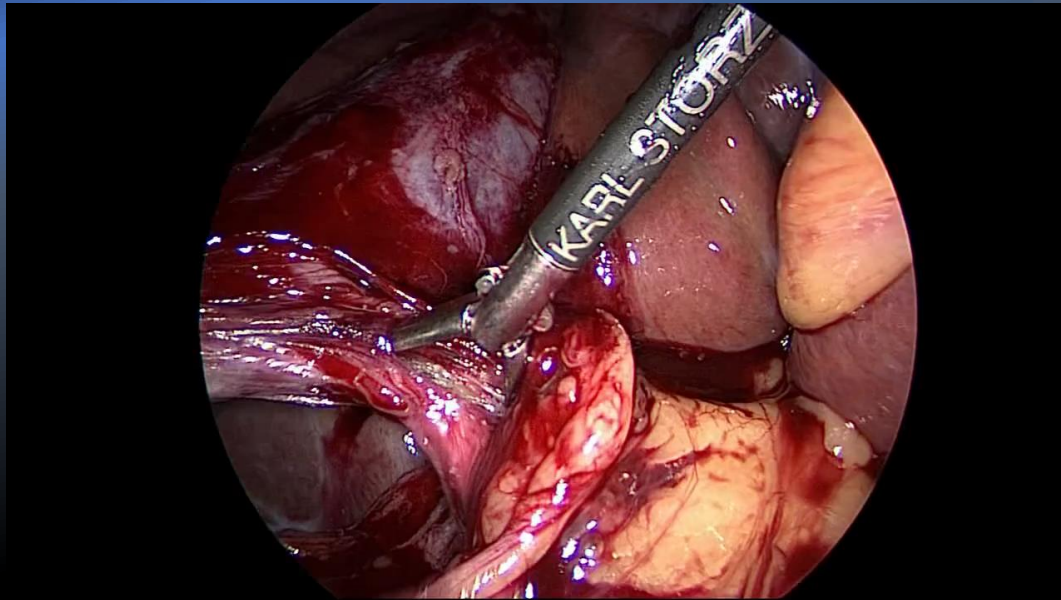
At your local hospital.

Initializing..... S.OS ..... /

A Blueprint for the Future of Medicine



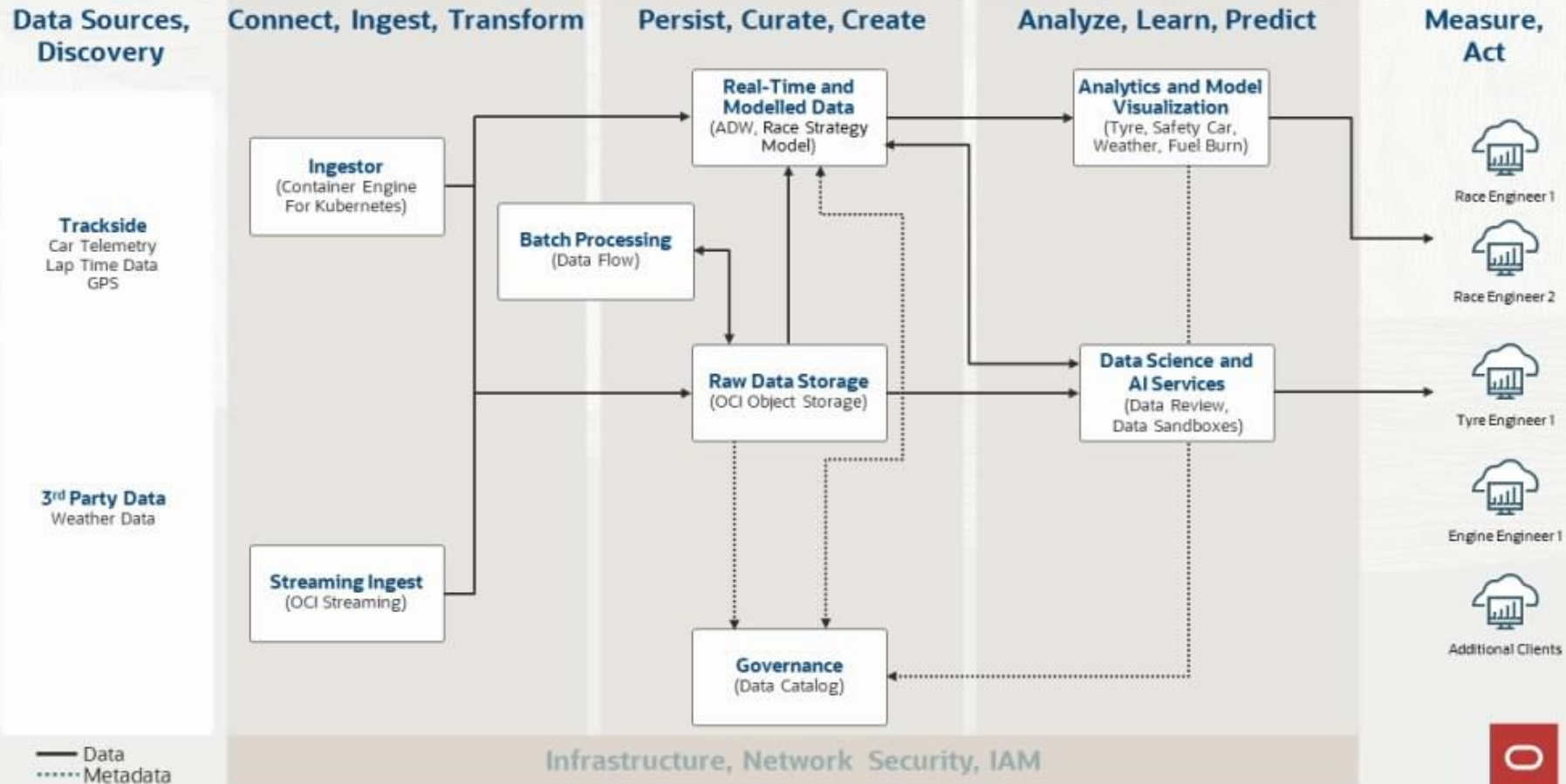
# Surgical Event Real Time Prediction



SUPR-GAN: SURgical PRediction GAN for Event Anticipation in Laparoscopic and Robotic Surgery

# Error Handling and Recovery

## Oracle Data Platform – Race Strategy



# Analysis of intraoperative video *with Decision Support* .

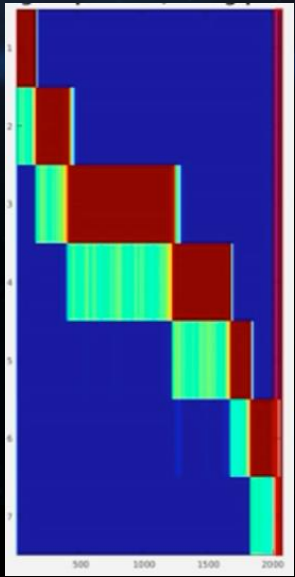
Conceptual Phase

## Case 3 – Preventing Complication



Normal range

Deviation





EARLY ADOPTERS

EARLY MAJORITY

CROSSING THE CHASM

EARLY ADOPTERS

EARLY MAJORITY



JP Morgan Chase  
eGOTIV  
Solutions

## Faculty and Fellows



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## SAAIL Team

 @MGHSAAIL

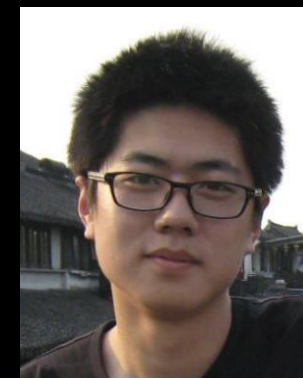
## Alumni



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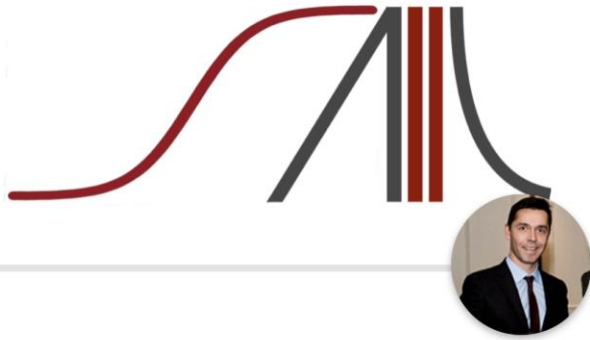


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# Thank you!



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