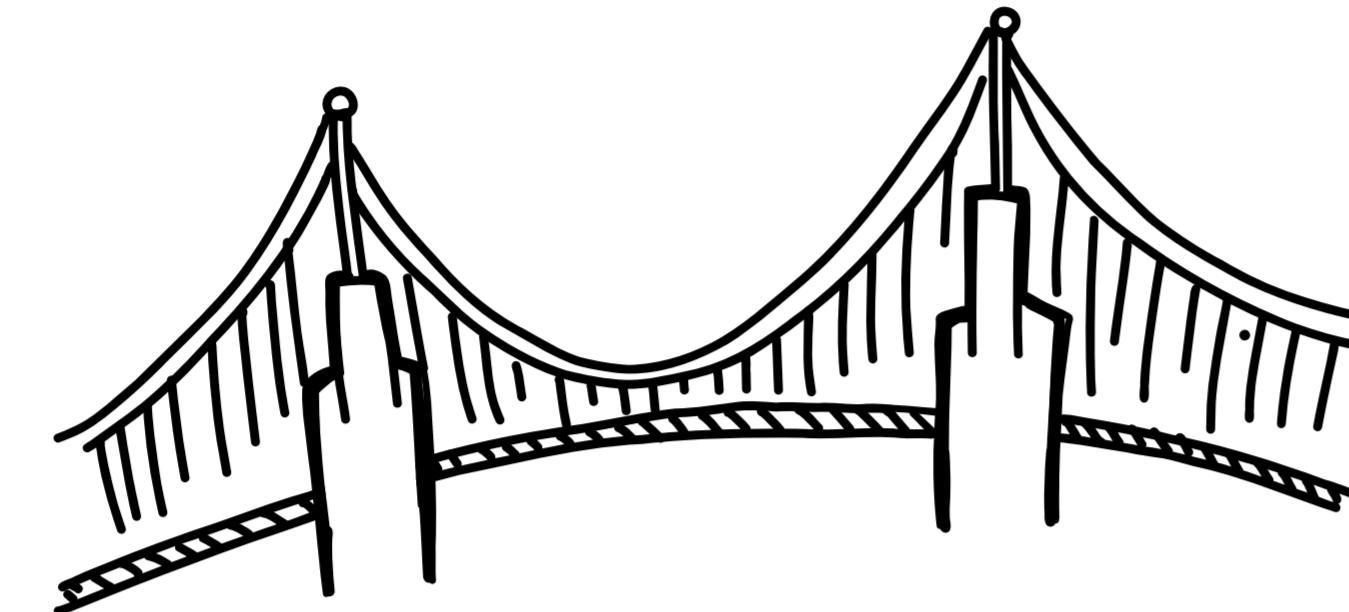
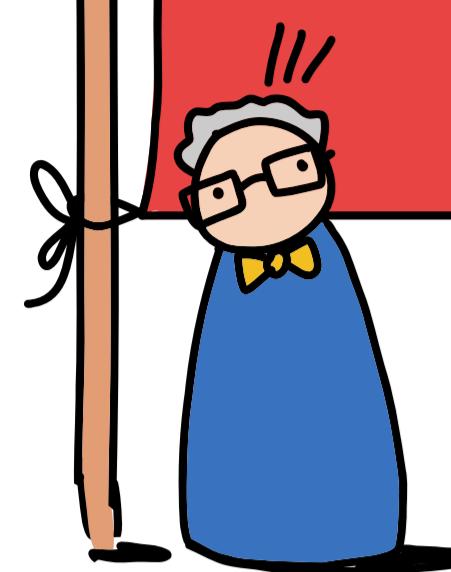
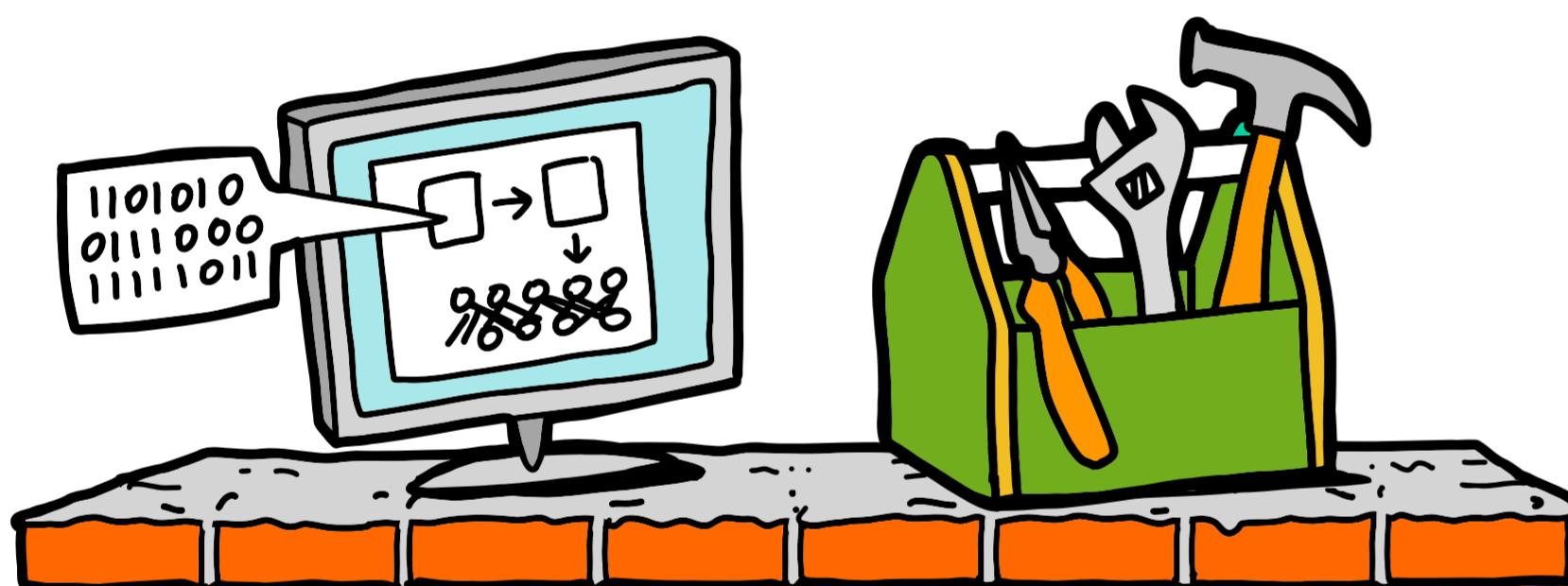


MACHINE LEARNING for SCIENCE



BRIDGING ML & SCIENCE:
DIRECTIONS in MATHEMATICS

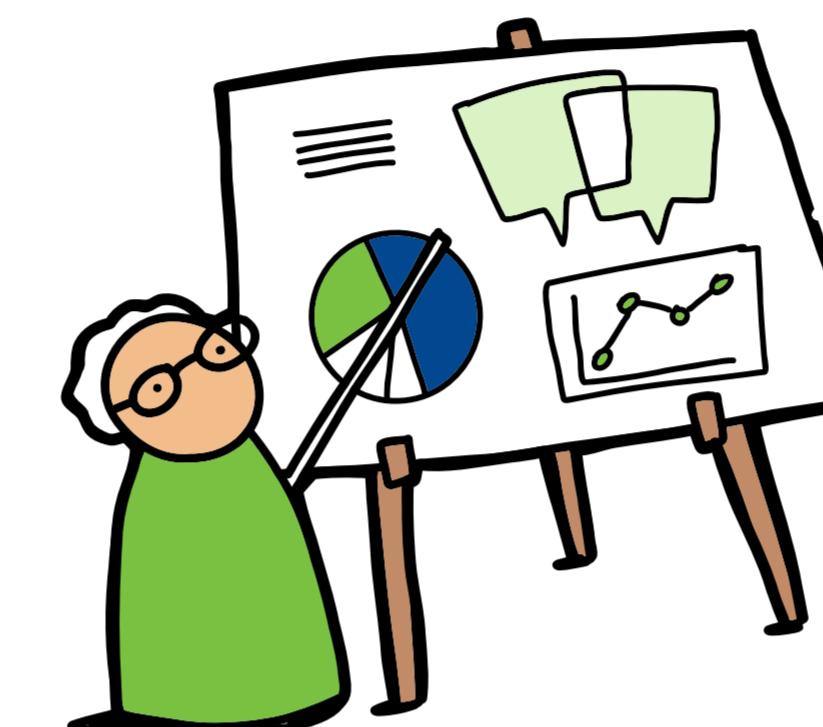


FOUNDATIONAL CONCEPTS
& EMERGING METHODS

- ROBERT C. WILLIAMSON
- PHILIPP HENNIG
- CARL-HENRIK EK
- RICHARD WILKINSON
- BUBACARR BAH

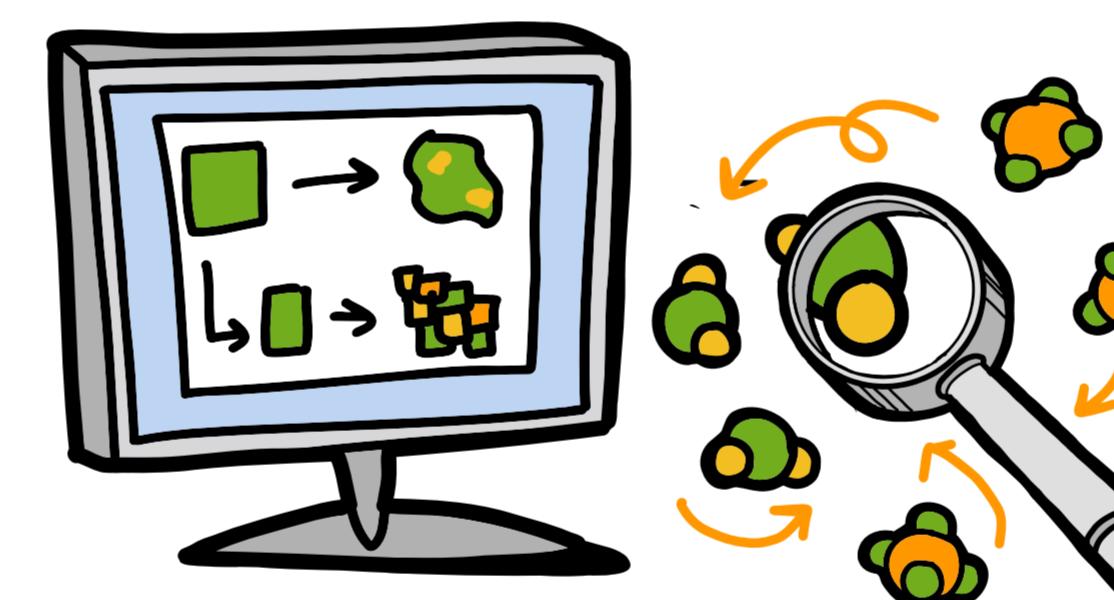
- BERT KAPPEN
- WILSON GREGORY
- SAMUEL KASKI
- JAKOB MACKE
- NIKI KILBERTUS

MATHEMATICS at the INTERFACE of DATA-DRIVEN & MECHANISTIC MODELLING



LESSONS from the APPLICATION
of MACHINE LEARNING in SCIENCE

- MAREN BÜTTNER
- CHRISTIAN IGEL
- DINA MACHUVE
- CHALLENGER MISHRA
- KLAUS-ROBERT MÜLLER



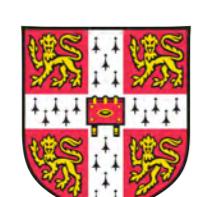
MACHINE LEARNING for
EARTH & CLIMATE SCIENCES

- VERONIKA EYRING
- MARKUS REICHSTEIN
- GUSTAU CAMPS-VALLS
- CHRISTIAN REINERS,
ALEXANDER WINKLER

THIS CONFERENCE
was HELD at the
MATHEMATISCHES
FORSCHUNGSINSTITUT
OBERWOLFACH on
JULY 12 - 16, 2023

CLICK the
SECTION TITLES/
SCIENTISTS' NAMES
to Go to PAGE

ORGANISED by NEIL LAWRENCE,
JESSICA MONTGOMERY,
& BERNHARD SCHÖLKOPF



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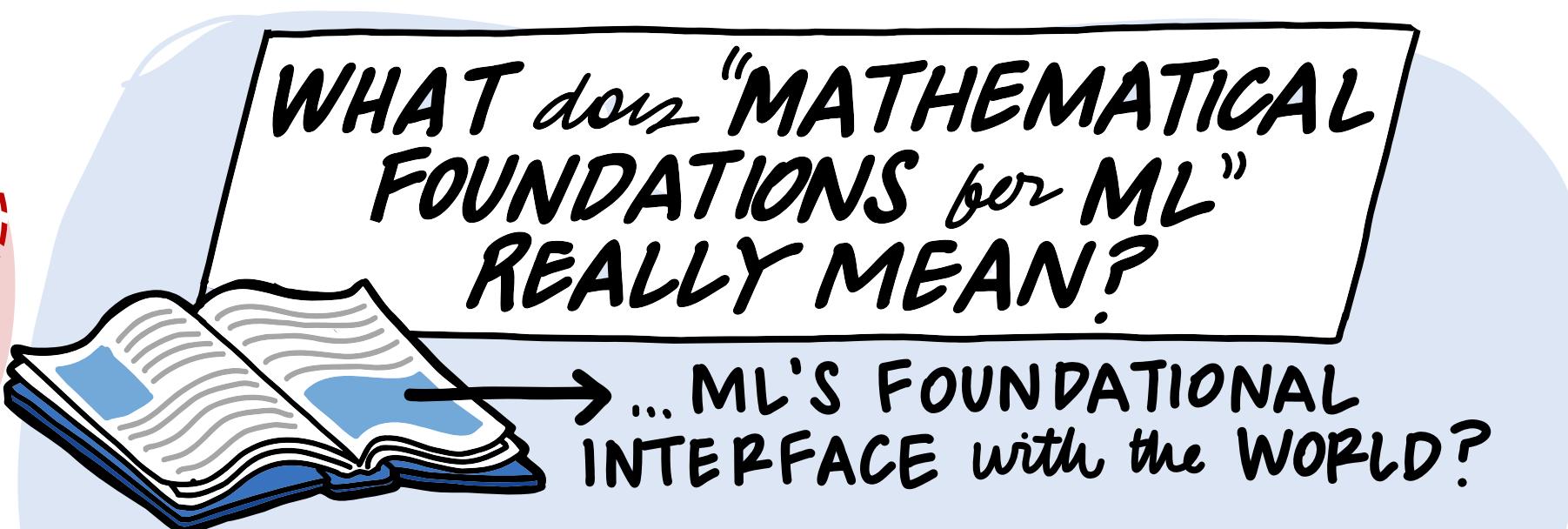
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BRIDGING ML & SCIENCE: DIRECTIONS in MATHEMATICS

TODAY'S ML is USEFUL for ...

- ✓ STITCHING TOGETHER DIFFERENT TYPES of DATA for a NUANCED VIEW.
- ✓ EXTRACTING INSIGHTS from DATA for MORE ACCURATE UNDERSTANDING of the PROPERTIES of a SYSTEM
- ✓ IDENTIFYING AREAS for EXPERIMENTATION & THEORIZING
- ✓ SPEEDING UP ANALYSIS
- ✓ SIMULATING COMPLEX SYSTEMS

WHAT does "MATHEMATICAL FOUNDATIONS for ML" REALLY MEAN?



...ML's FOUNDATIONAL INTERFACE with the WORLD?

MATHS HELPS US DESCRIBE RELATIONSHIPS

- ...to REPRESENT DOMAIN KNOWLEDGE to be EMBEDDED in ML SYSTEMS
- ...to EXPRESS CONCEPTS like FAIRNESS or INTERPRETABILITY
- ...to CREATE MODELS w/ UNCERTAINTY AWARENESS
- ...to HANDLE HIGH-DIMENSIONAL & SPARSE DATA
- ...to SUPPORT VALIDATION & BENCHMARKING
- ...to MODEL USERS
- ...to PRODUCE EFFICIENT METHODS

FOUNDATIONAL CONCEPTS & EMERGING MATHS can ACT as BRIDGES!



LIKE...

- ...STATISTICS & PROBABILITY
- ...RIEMANNIAN GEOMETRY
- ...LINEAR ALGEBRA
- ...OPTIMISM TECHNIQUES & CALCULUS

& ALSO...

- ...PROBABILISTIC NUMERICS
- ...IMPRECISE PROBABILITIES
- ...FINSLERIAN GEOMETRY
- ...ADJOINT LATENT FORCE MODELS
- ...EQUIVARIANCES

COULD we CREATE a SUITE of these AI TOOLS to DEPLOY ACROSS FIELDS?

RECENT CLIMATE CHANGE is UNPRECEDENTED

MODELS HELP us SIMULATE & ANALYSE the CLIMATE SYSTEM.

MODELS are VITAL, we NEED THEM to be TRUSTWORTHY

- ✓ STABLE
- ✓ PHYSICALLY-PLAUSIBLE
- ✓ BASED ON CAUSAL DRIVERS

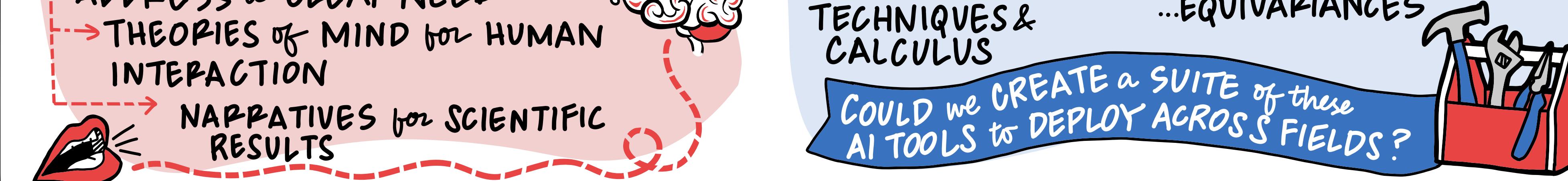
CURRENT ML APPLICATIONS in CLIMATE SCIENCE HELP US...

- ★ UNDERSTAND SYSTEMS for CO₂ EXCHANGE & CARBON DYNAMICS
- ★ PREDICT LANDSCAPE & ECOSYSTEM CHANGES UNDER CHANGING CLIMATE CONDITIONS
- ★ FORECAST EXTREME WEATHER EVENTS
- ★ I.D. CAUSAL RELATIONSHIP THAT DRIVE THESE CHANGES

OPEN CHALLENGES...

- ...PARAMETER ESTIMATION & BAYESIAN INFERENCE
- ...EXPLAINABLE AI
- ...IMPROVED GENERALISATION ACROSS CHANGING DISTRIBUTIONS
- ...UNCERTAINTY QUANTIFICATION
- ...MANAGING HYBRID DATA-DRIVEN & MECHANISTIC MODELS

we NEED a COMMUNITY DEDICATED to DEVELOPING this WORK!



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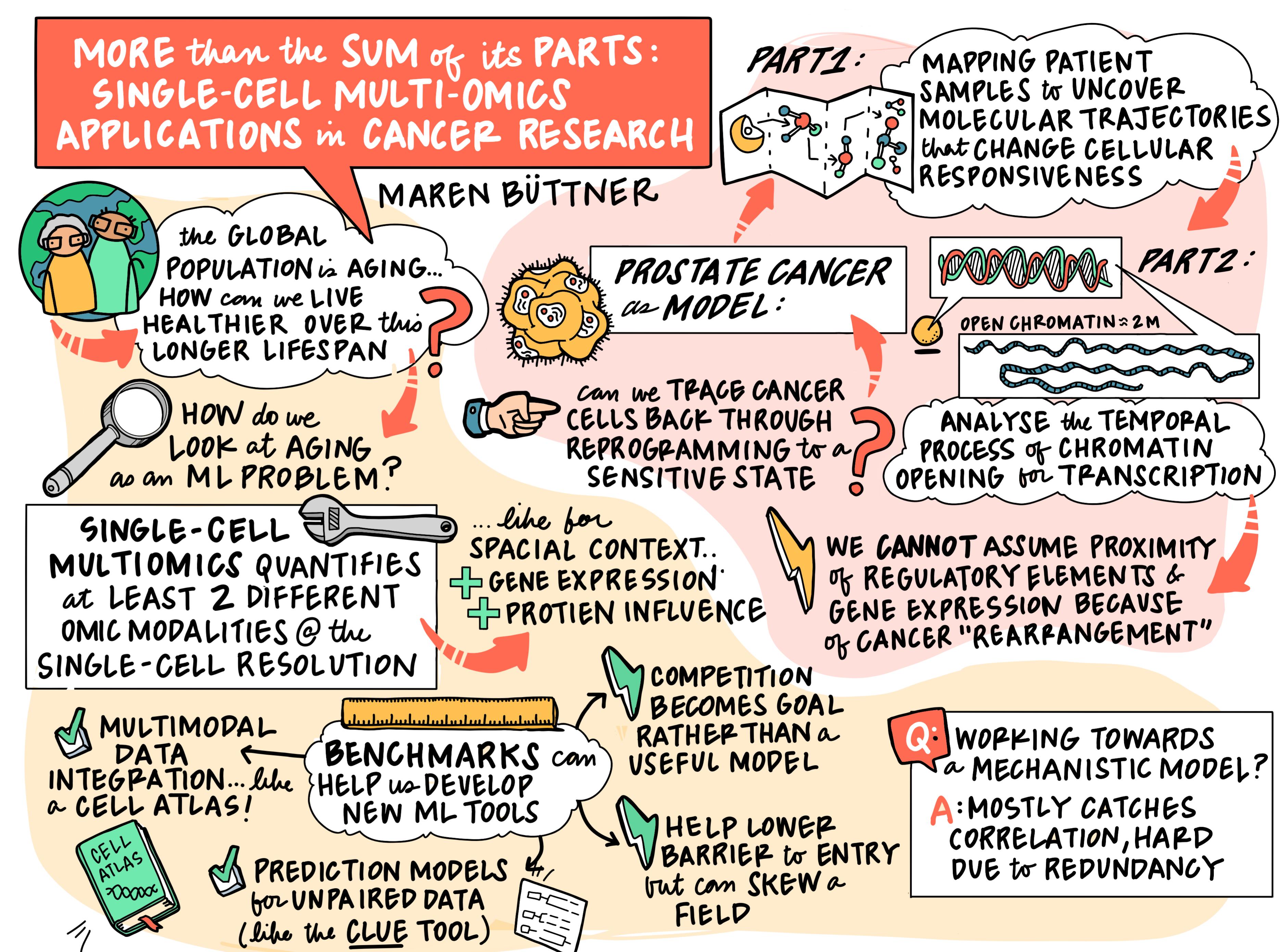


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More than the Sum of its Parts: Single-Cell Multi-omics Applications in Cancer Research

Maren Büttner
Calico Life Sciences
San Francisco, CA



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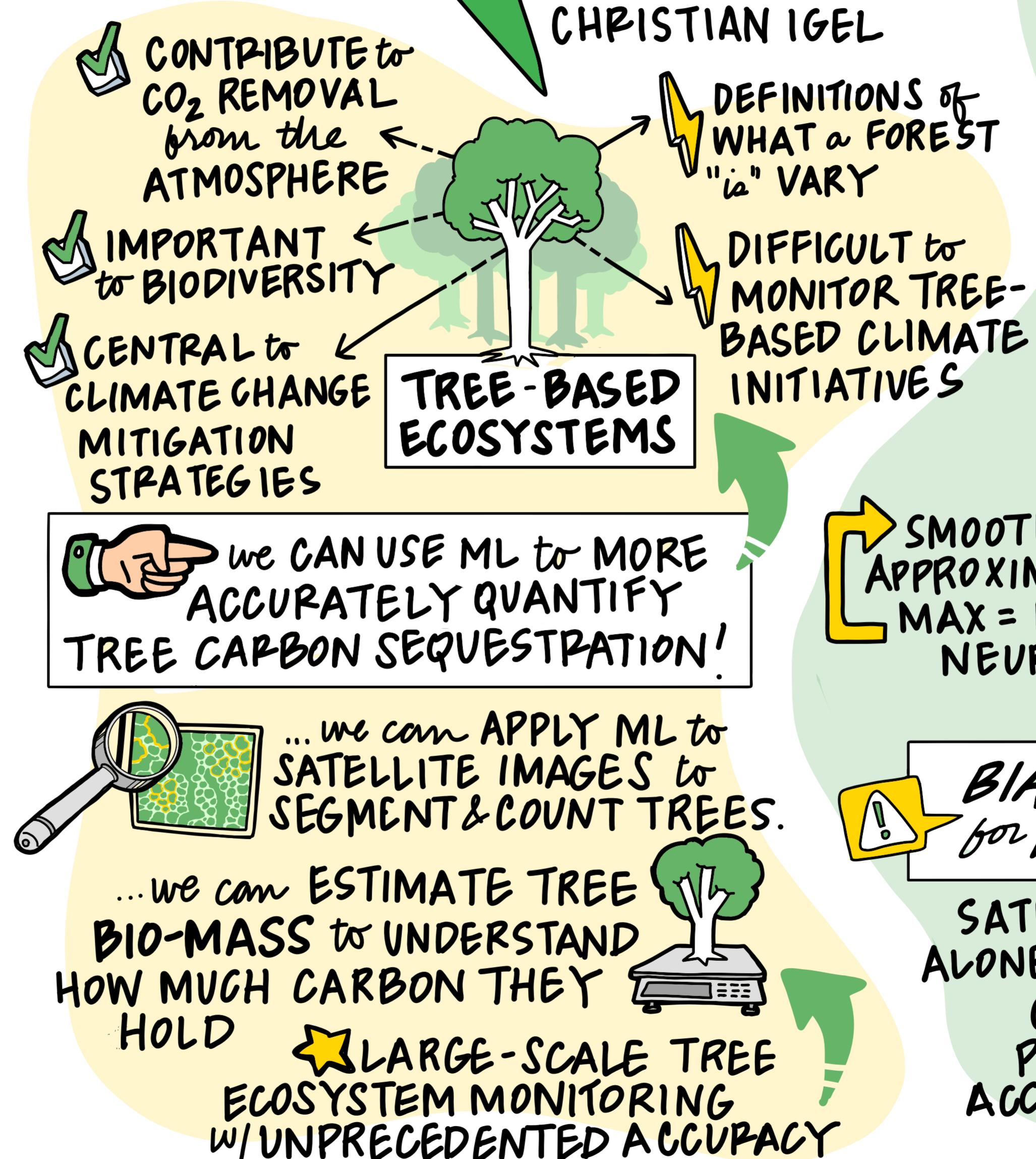
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Regression of Ecosystem Properties: Bias, Monotonicity, & Uncertainty

Christian Igel

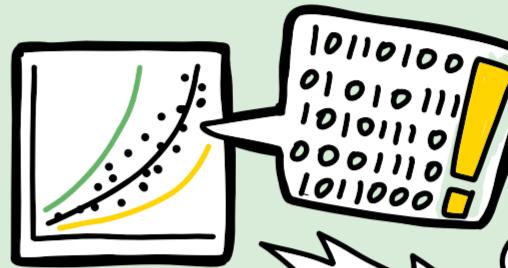
University of Copenhagen
Copenhagen, DN

REGRESSION of ECOSYSTEM PROPERTIES: BIAS, MONOTONICITY, & UNCERTAINTY



1 FITTING ALLOMETRIC EQUATIONS

we NEED a GOOD FIT to DATA w/ LOW BIAS, & WANT TIGHT UNCERTAINTY ESTIMATES



CHALLENGE: NON-SYMMETRIC, HETROSKEPSASTIC ERROR DISTRIBUTION

2 MONOTONIC NEURAL NETWORKS for BIOMASS PREDICTION

SMOOTH MONOTONIC APPROXIMATIONS of MIN/ MAX = NO SILENT NEURONS

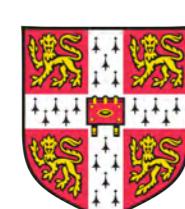
SMOOTH MODULE can be USED in a LARGER DL SYS. & TRAINED E-to-E.

BIAS in DEEP LEARNING for REGRESSIONS

SATELLITE IMAGES ALONE AREN'T ENOUGH... COMBINE w/ LiDAR POINT CLOUDS for MORE ACCURATE BIOMASS EST.



I.D. SPECIES!



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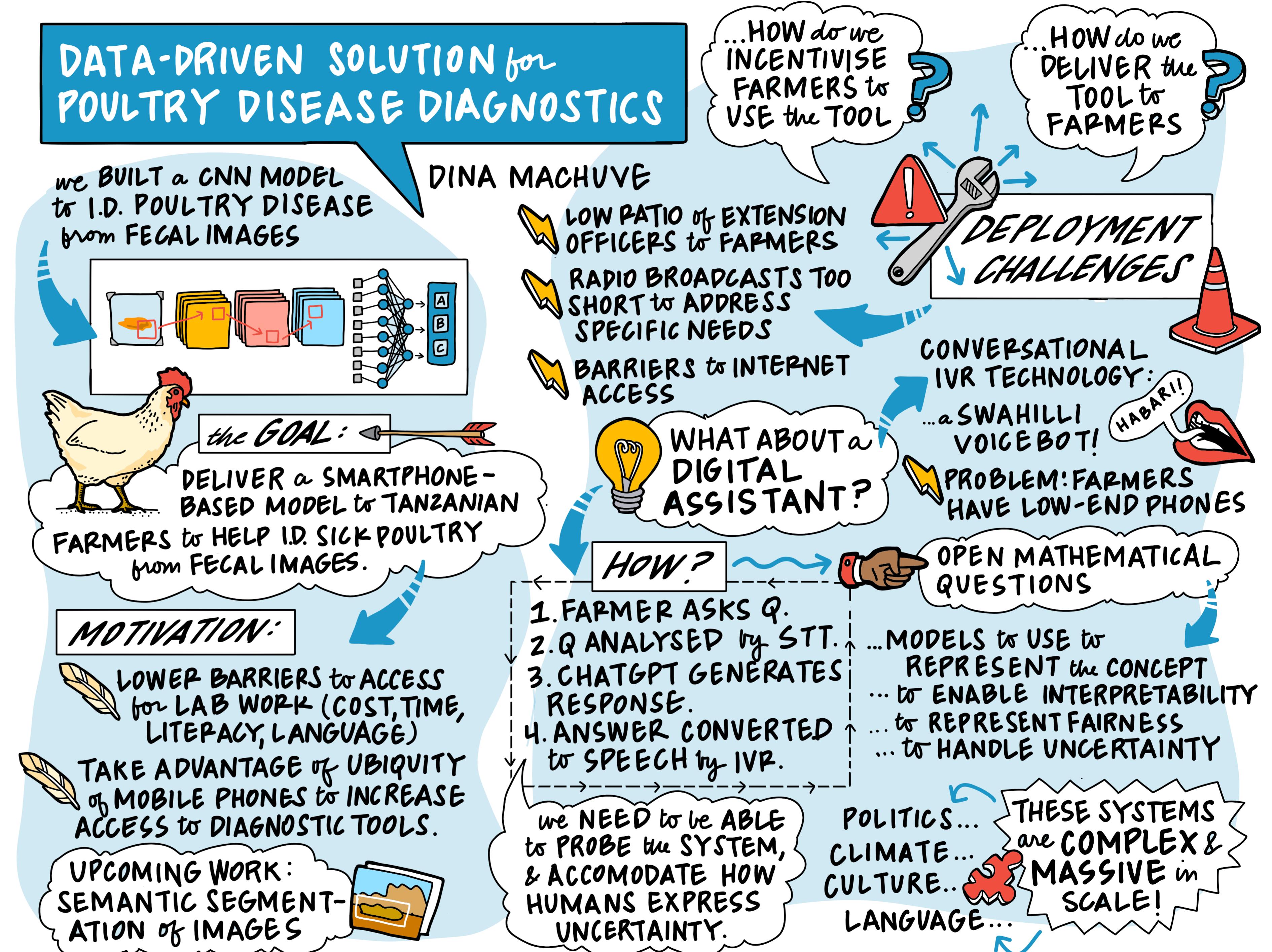


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Data-Driven Solution for Poultry Disease Diagnostics

Dina Machuve
DevData Analytics
Arusha, TZ



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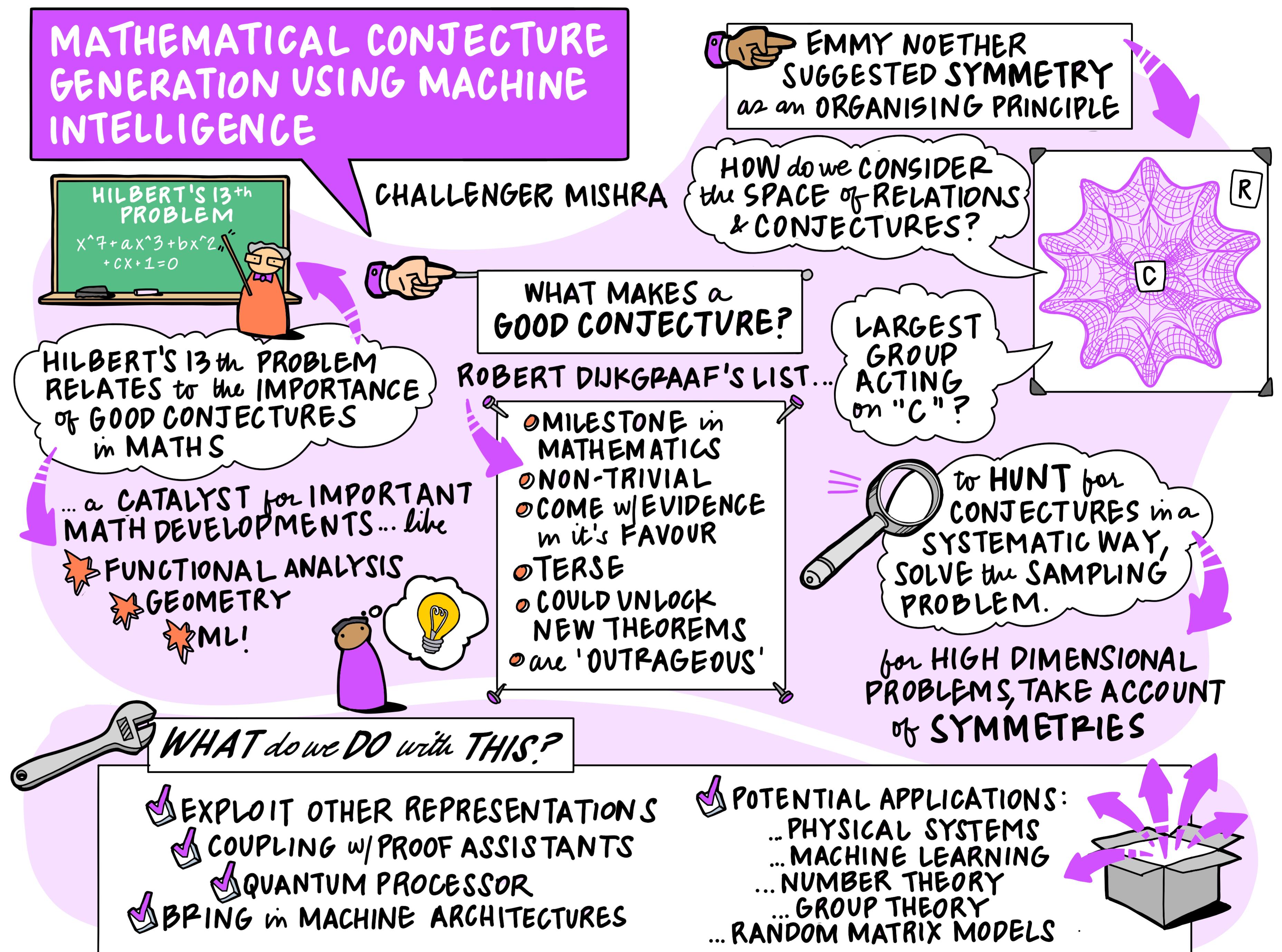
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Mathematical Conjecture Generation using Machine Intelligence

Challenger Mishra

University of Cambridge
Cambridge, UK



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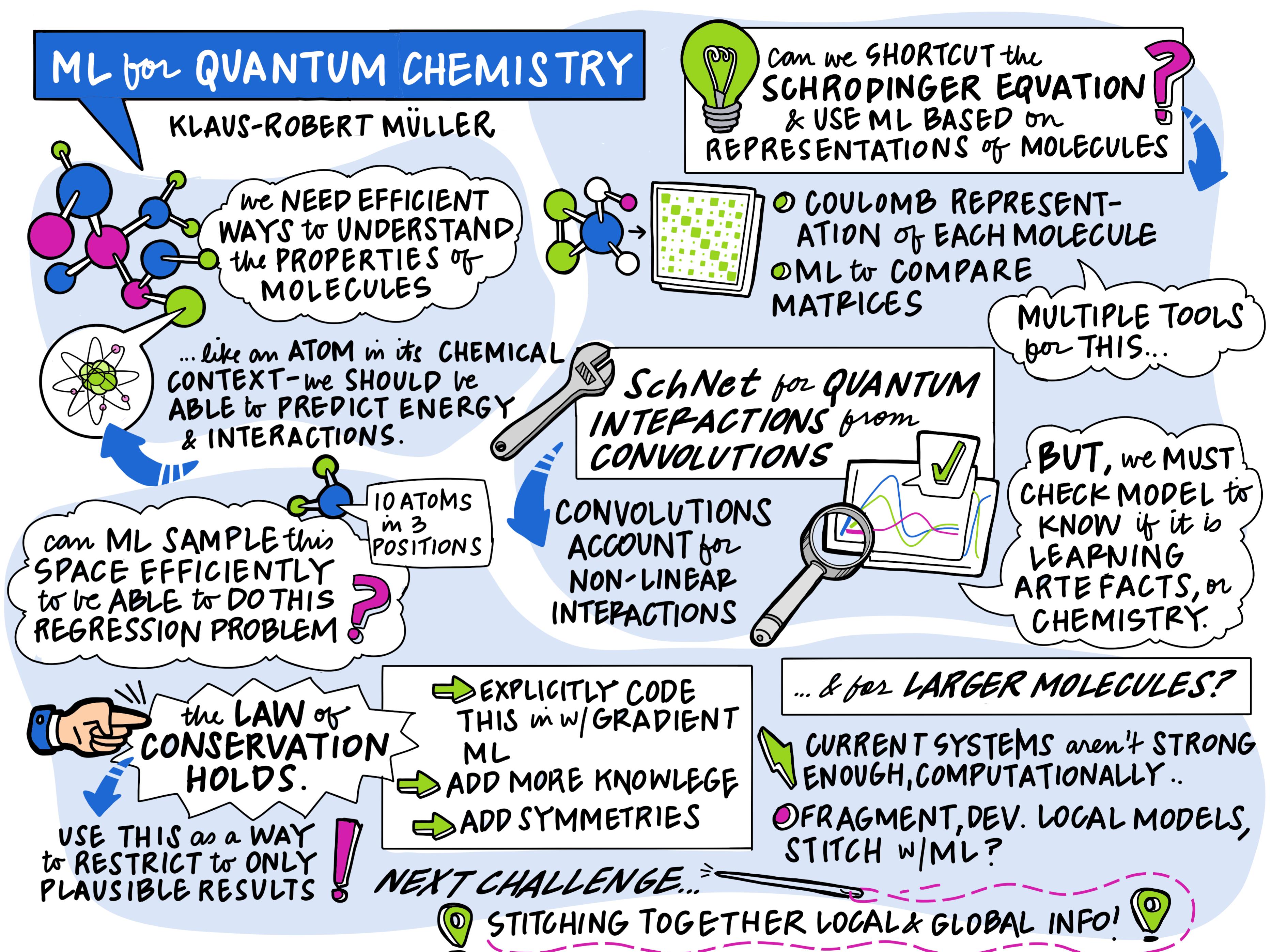


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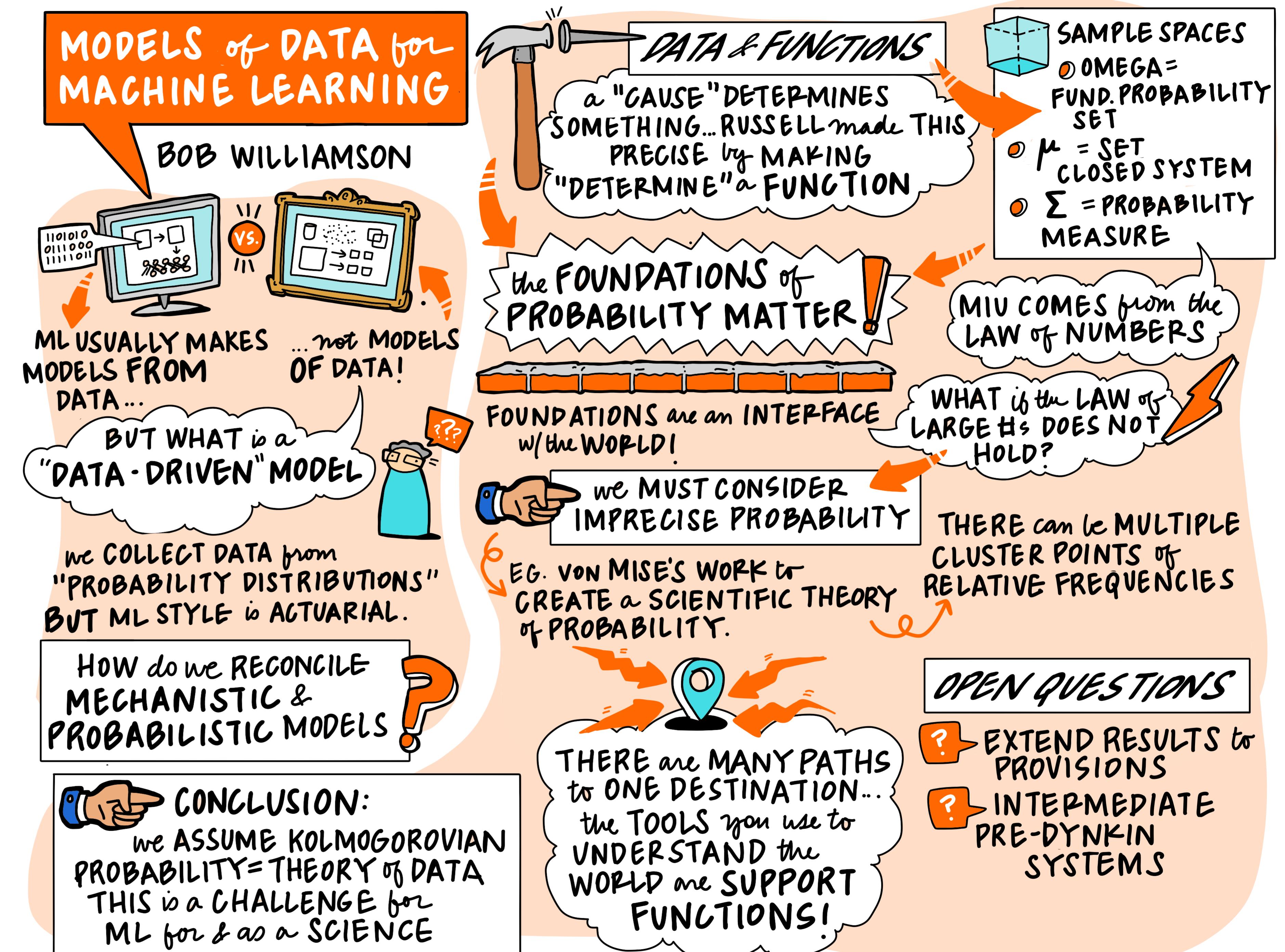
ML for Quantum Chemistry

Klaus-Robert Müller
Technical University Berlin
Berlin, DE



Models of Data for Machine Learning

Robert C. Williamson
University of Tübingen
Tübingen, DE



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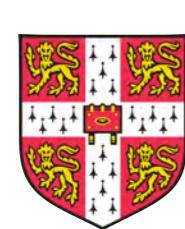
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Deep Learning only works if its Bayesian, & Bayesian Deep Learning is Easy

Philipp Hennig

University of Tübingen,
Max Planck Institute for Intelligent
Systems
Tübingen, DE



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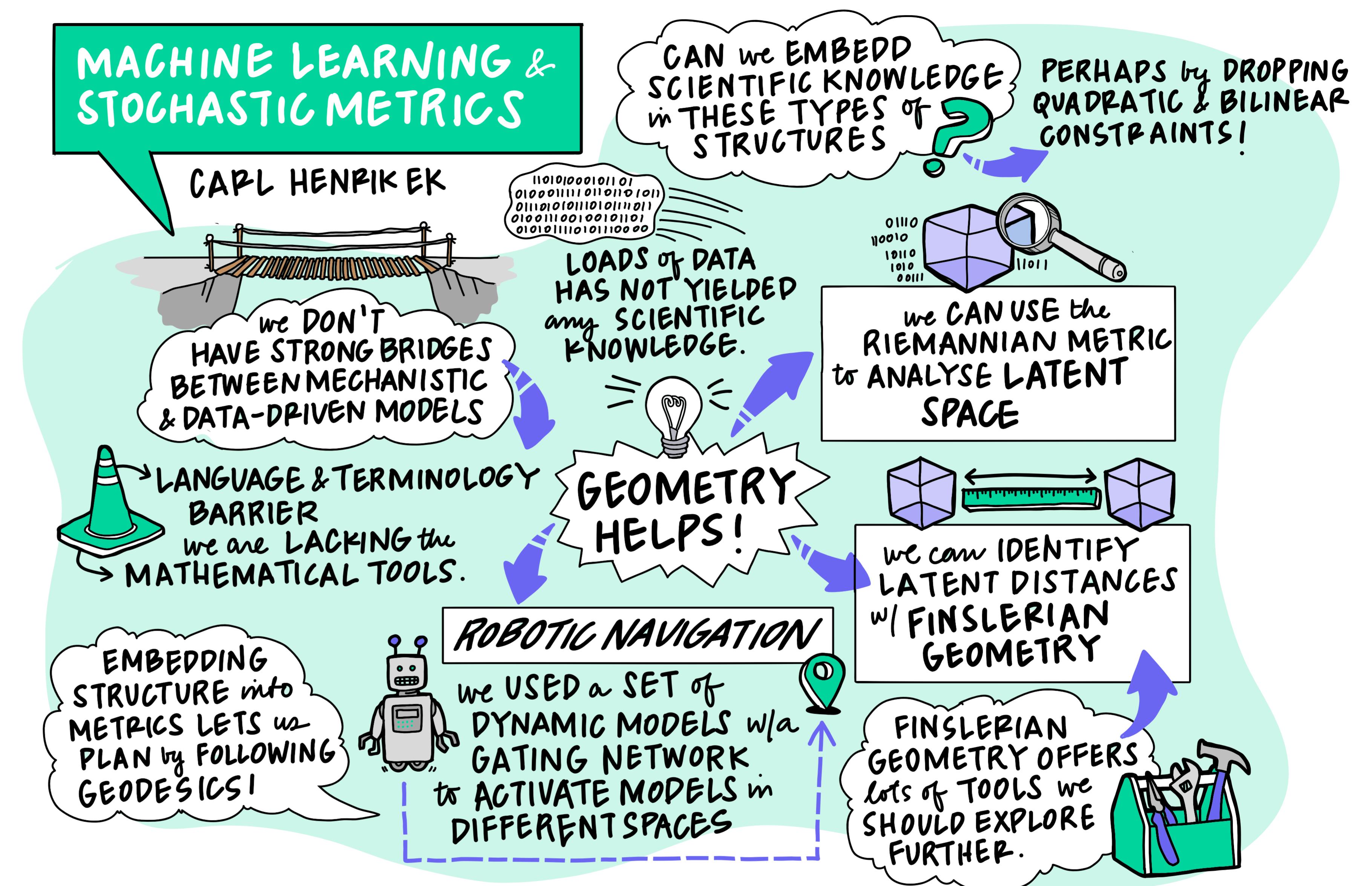


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Machine Learning and Stochastic Models

Carl Henrik Ek

University of Cambridge,
Cambridge, UK



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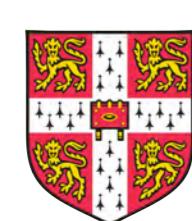
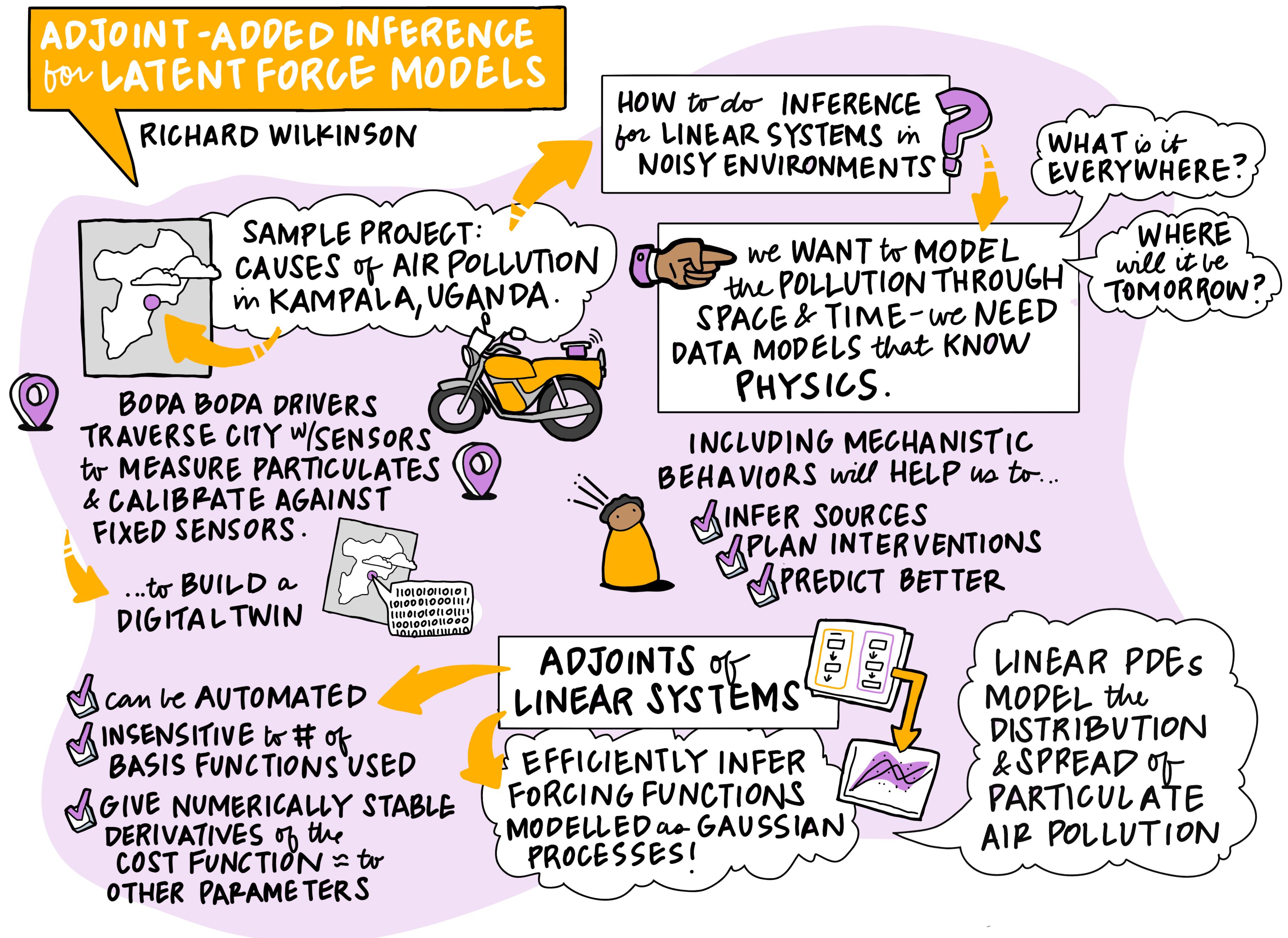


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Adjoint-added Inference for Latent Force Models

Richard Wilkinson
University of Nottingham
Nottingham, UK



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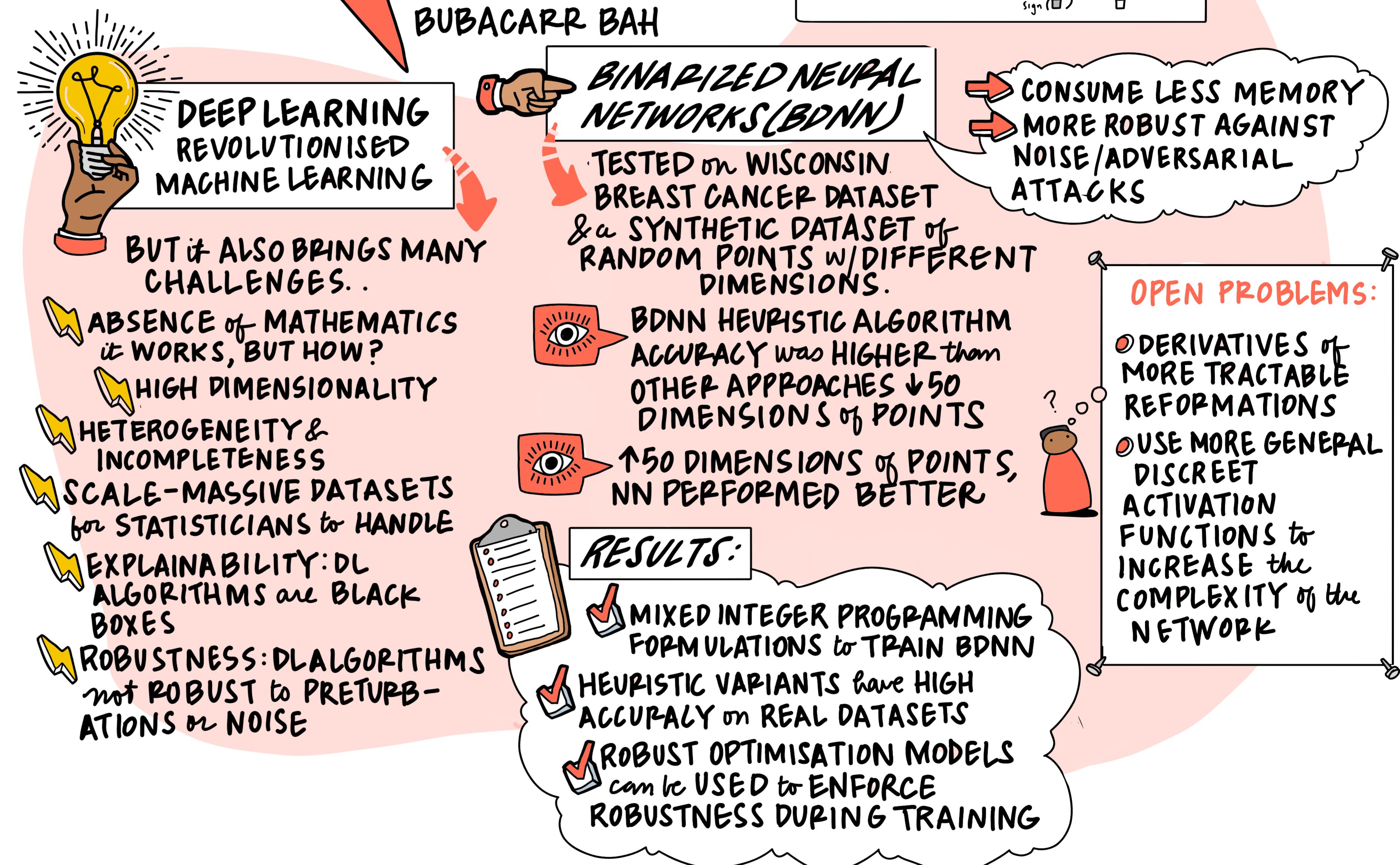
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Efficient & Robust Optimization Models for Trained Binarized Deep Neural Networks

Bubacarr Bah

Medical Council Unit The Gambia,
London School of Hygiene & Tropical
Medicine
Serrekunda, GM

EFFICIENT & ROBUST OPTIMIZATION MODELS for TRAINED BINARIZED DEEP NEURAL NETWORKS



Why Adiabatic Quantum Annealing is Unlikely to Yield Quantum Speed Up

Bert Kappen

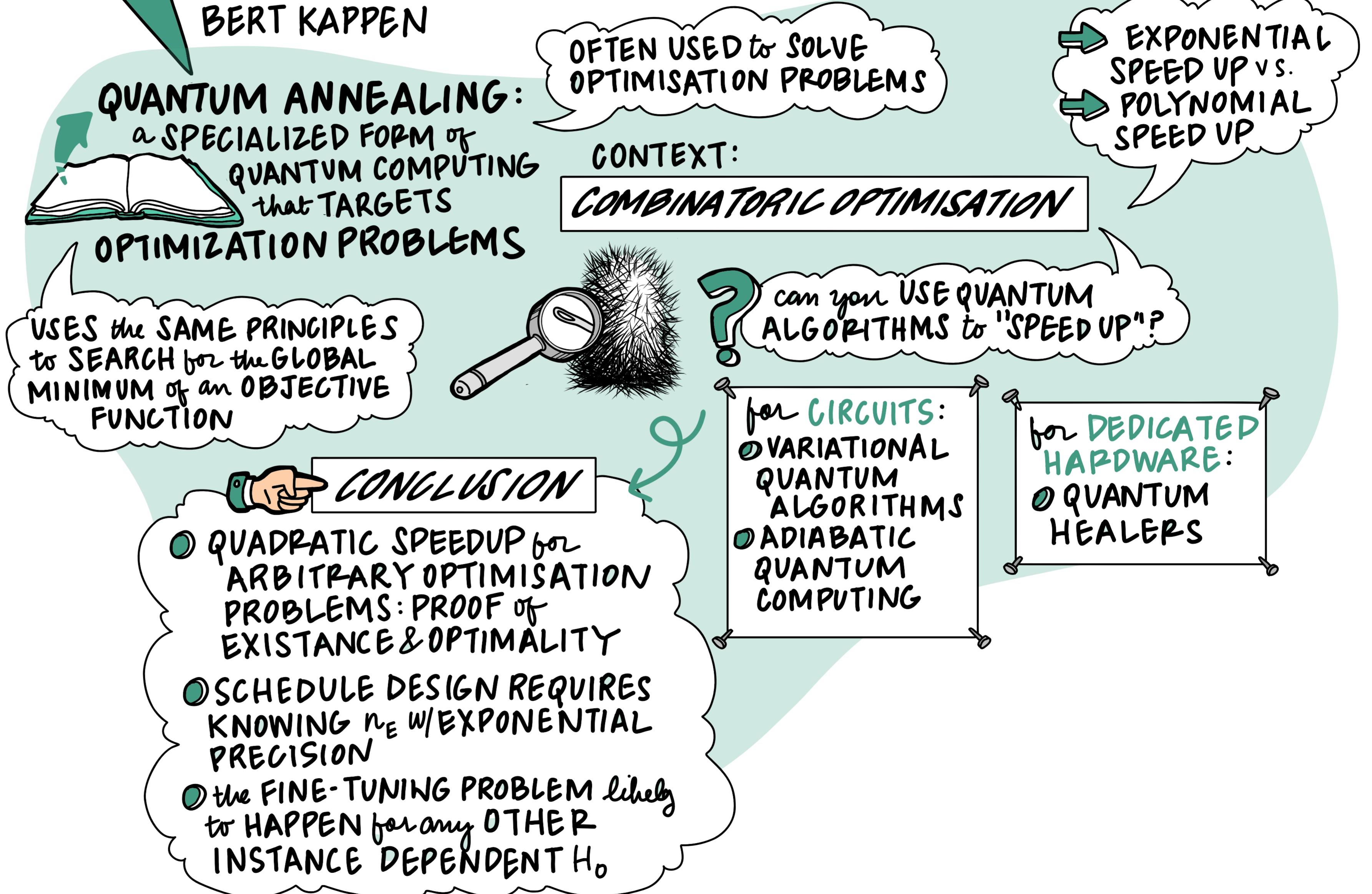
Radboud University
Nijmegen, NL

WHY ADIABATIC QUANTUM ANNEALING is UNLIKELY to YIELD QUANTUM SPEED UP

BERT KAPPEN

QUANTUM ANNEALING:
a SPECIALIZED FORM of QUANTUM COMPUTING
that TARGETS
OPTIMIZATION PROBLEMS

USES the SAME PRINCIPLES to SEARCH for the GLOBAL MINIMUM of an OBJECTIVE FUNCTION



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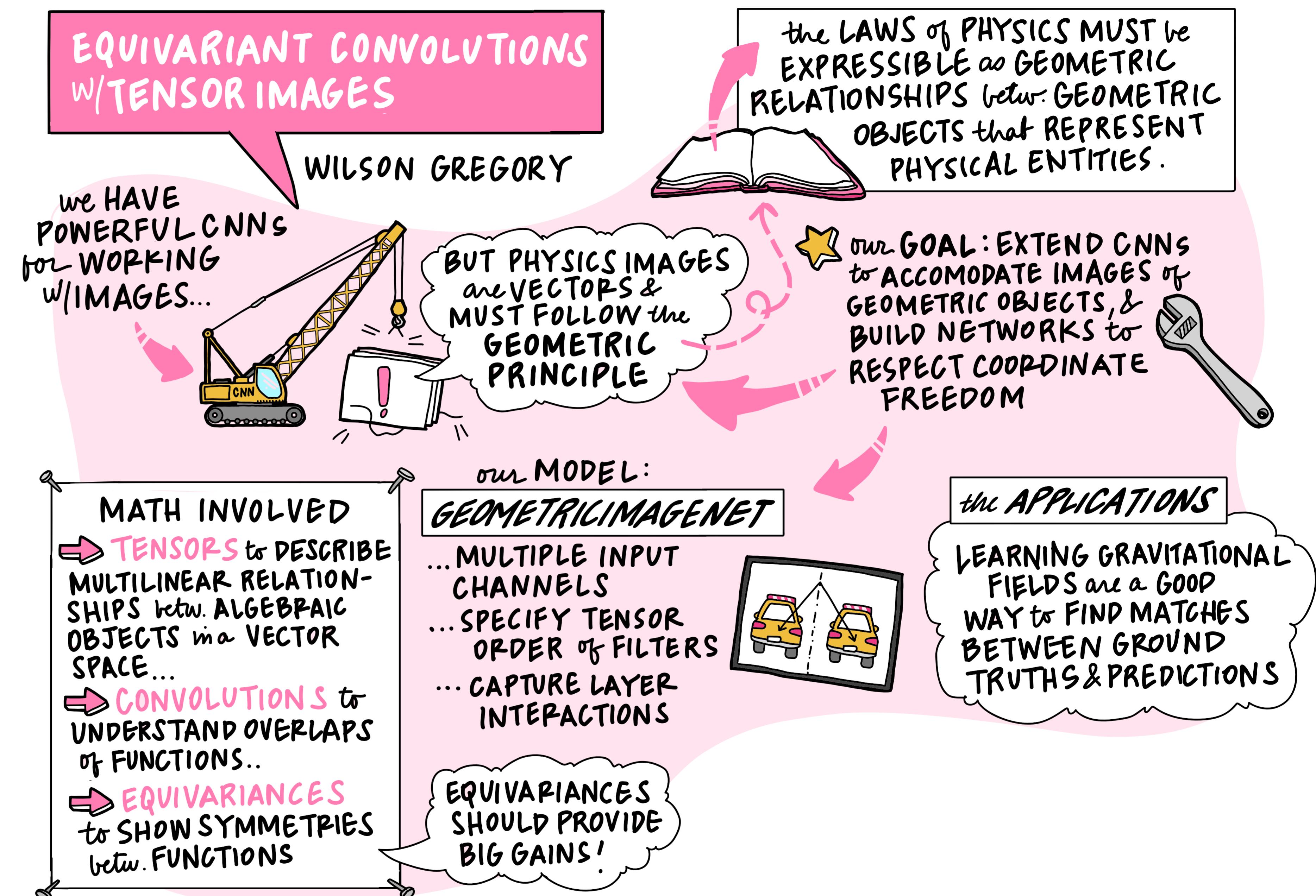


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Equivariant Convolutions with Tensor Images

Wilson Gregory
Johns Hopkins University
Baltimore, US



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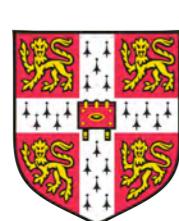
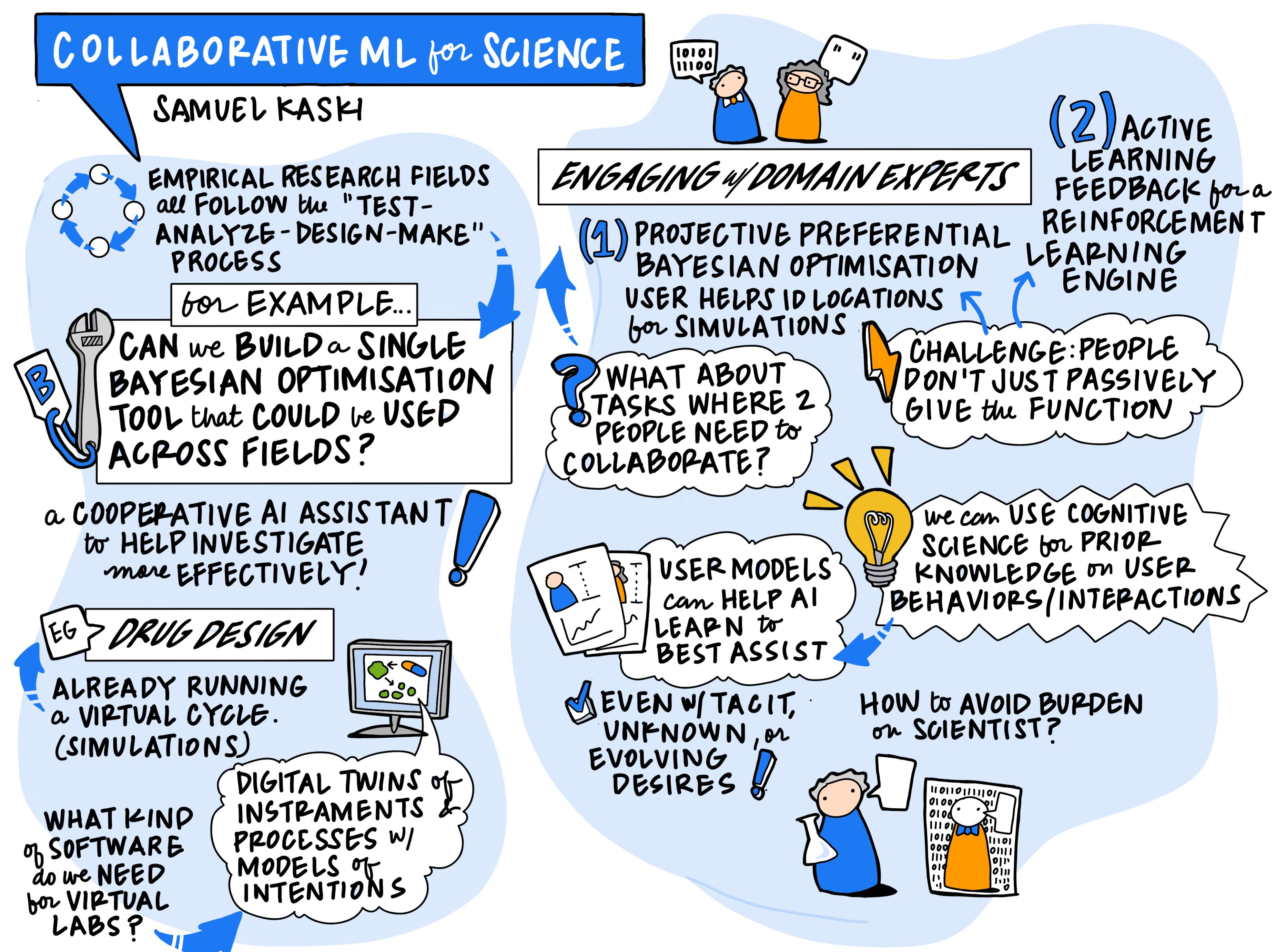
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Collaborative ML for Science

Samuel Kaski

University of Manchester,
Aalto University
Manchester, UK,
Espoo, FI



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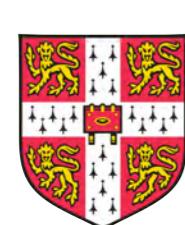
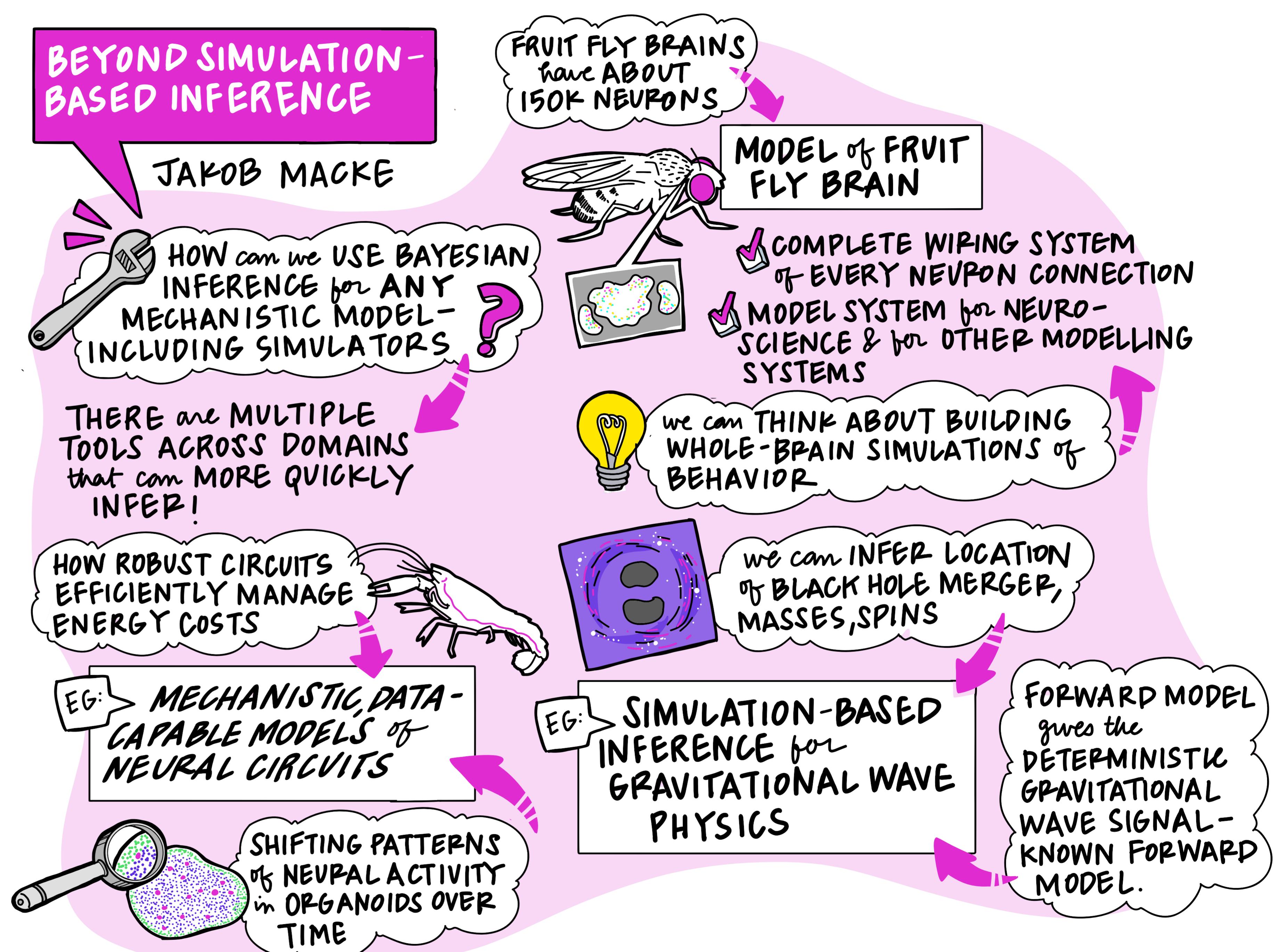
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Beyond Simulation- Based Inference

Jakob Macke

University of Tübingen
Tübingen, DE



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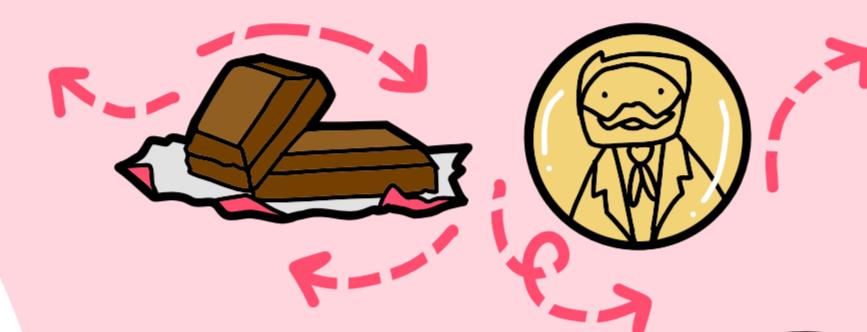
Experiment Design as Sequential Instrument Selection

Niki Kilbertus
Technical University of Munich
Munich, DE

EXPERIMENT DESIGN as SEQUENTIAL INSTRUMENT SELECTION

NIKI KILBERTUS

RANDOMISATION
is a POWERFUL TOOL!



HOW do we CHOOSE
the BEST TREATMENT
to FIND RELATIONSHIPS
in DATA COLLECTED
from EXPERIMENTS

- ➡ EFFECT of TREATMENT
on OUTCOME
- ➡ INSTRUMENT SHOULD
have an EFFECT on the
TREATMENT

! INSTRUMENT
SHOULD ONLY
EFFECT OUTCOME
via the
TREATMENT

HYPOTHESIS:

SCIENTIFIC EXPERIMENTS
OFTEN TARGET INSTRUMENTS,
NOT TREATMENTS

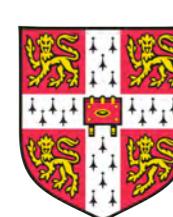
in NATURAL SCIENCES, we are ONLY
ABLE to RUN a FEW EXPERIMENTS,
w/ a LIMITED # of INSTRUMENTS.

⚡ FULL TREATMENT EFFECT
is NOT IDENTIFIABLE in a
SINGLE EXPERIMENT EVEN
in a LINEAR CASE

EG: DRUGS...
ANTIBIOTICS...

DISCUSSION

- ❓ HOW can we EXTEND these TECHNIQUES to
a NON-LINEAR SETTING
- ❓ FINDING VALUABLE METRICS for HOW
our CURRENT STATE is DOING
- ❓ NEED for ACTUAL OPTIMISATION/
SELECTION STRATEGIES
- ❓ WHAT can CAUSAL INFERENCE
METHODLOGY do for SCIENCE



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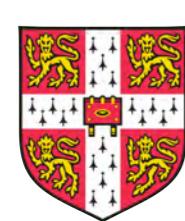
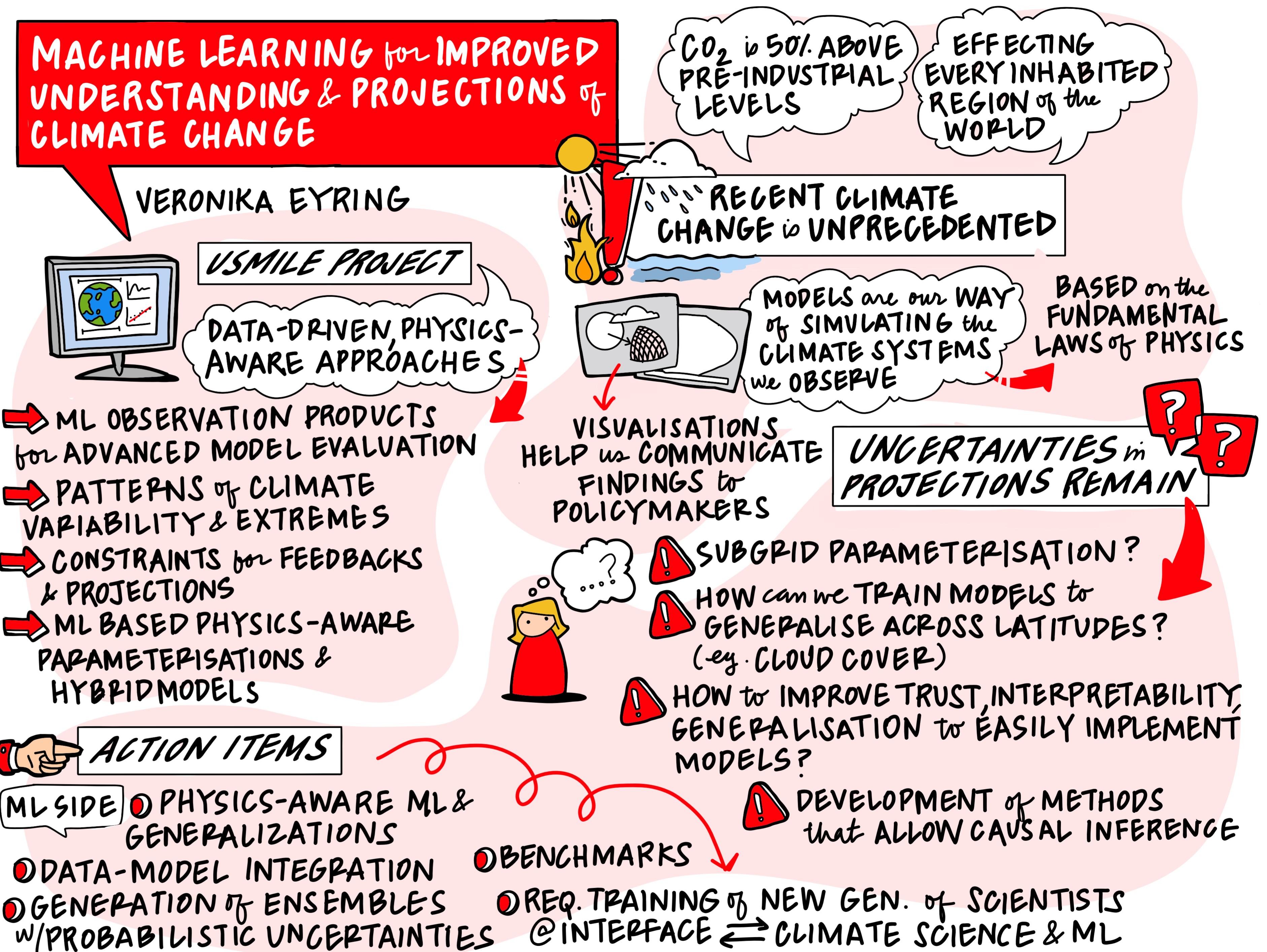


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Machine Learning for Improved Understanding & Projections of Climate Change

Veronika Eyring

German Aerospace Center (DLR) Institute of Atmospheric Physics,
University of Bremen
Bremen, DE



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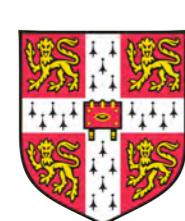
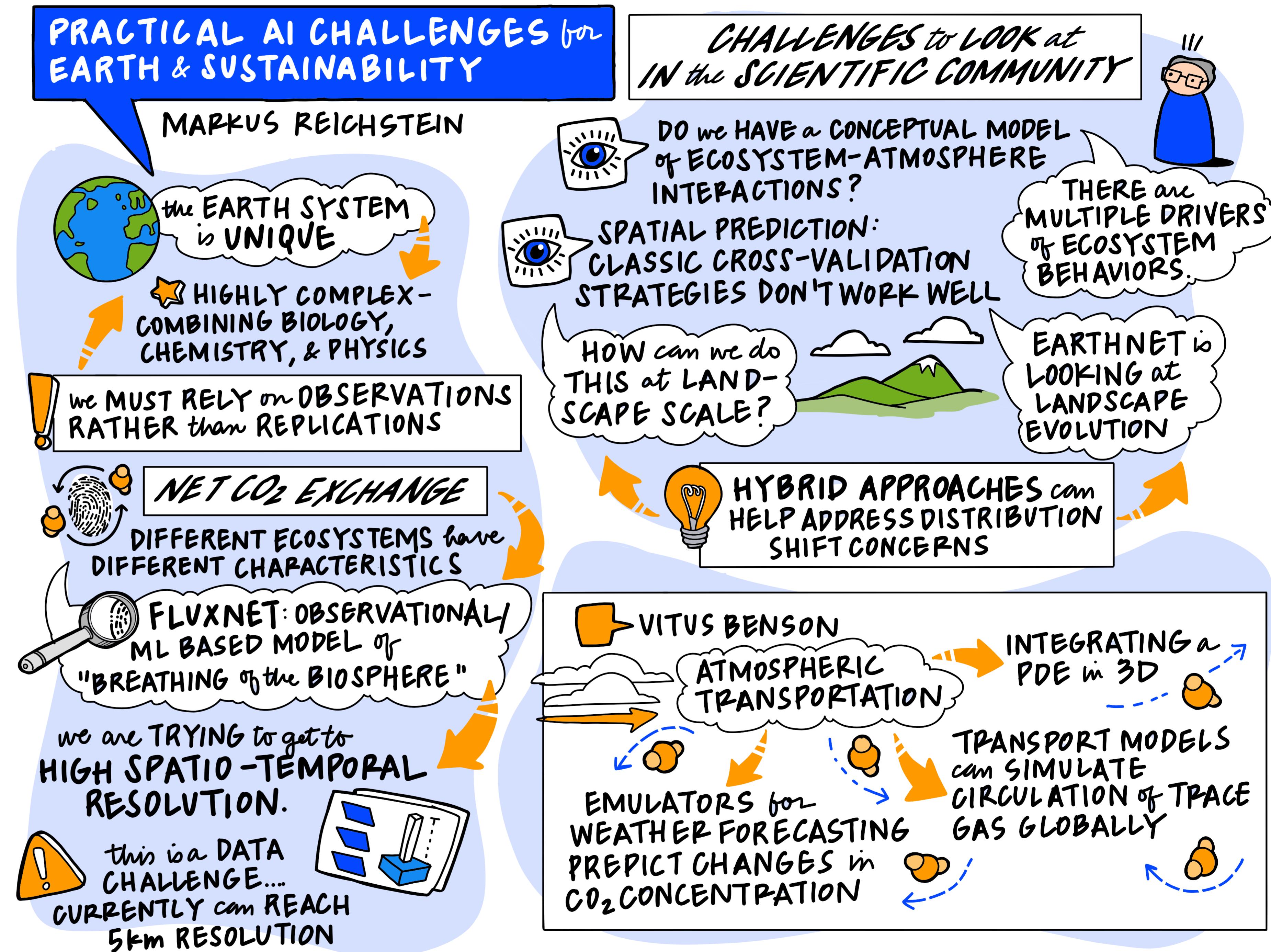


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Practical AI Challenges for Earth and Sustainability

Markus Reichstein

Max Planck Institute for
Biogeochemistry
Jena, DE



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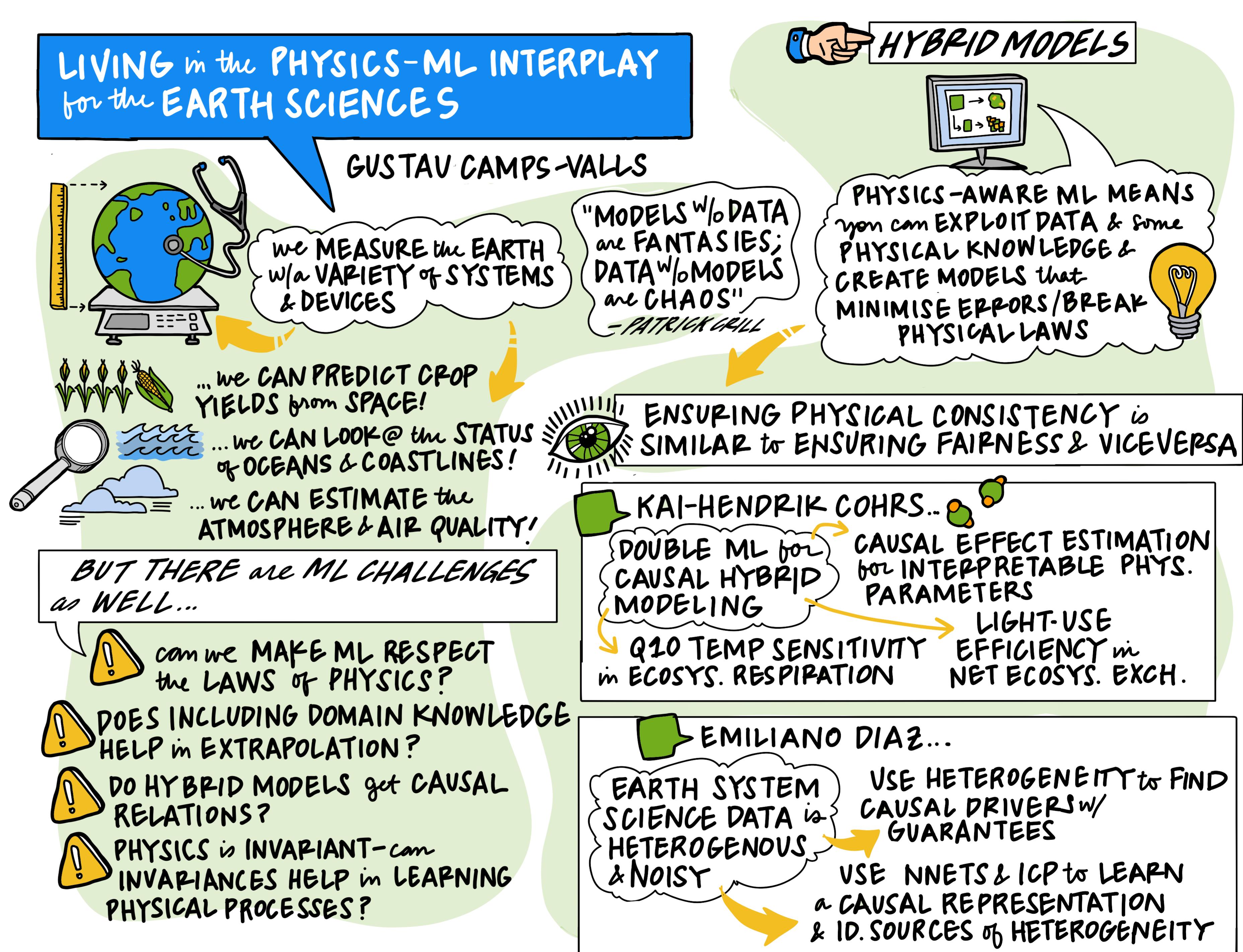
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Living in the Physics-ML Interplay for the Earth Sciences

Gustau Camps-Valls
Universitat de València
València, ES



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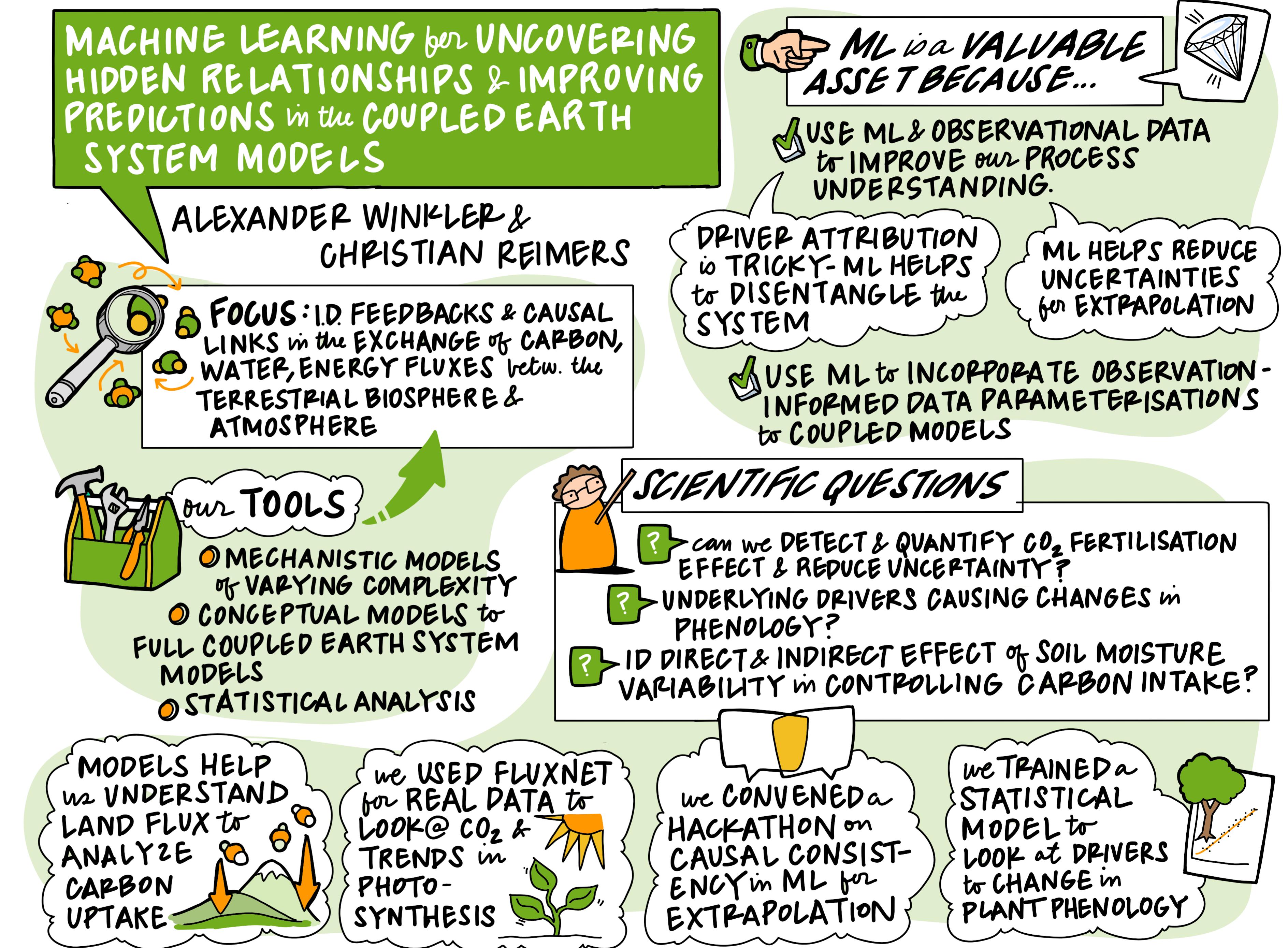


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Machine Learning for Uncovering Hidden Relationships & Improving Predictions in the Coupled Earth System Models

Christian Reimers &
Alexander Winkler

Max Planck Institute for Biogeochemistry
Aachen University
Jena, DE
Aachen, DE



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