



Predicting diseases through machine learning models

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PROSTATE CANCER
RESEARCH INITIATIVE
knowledge | action | impact

Disclosures

Founder and CEO of Nanostics Inc

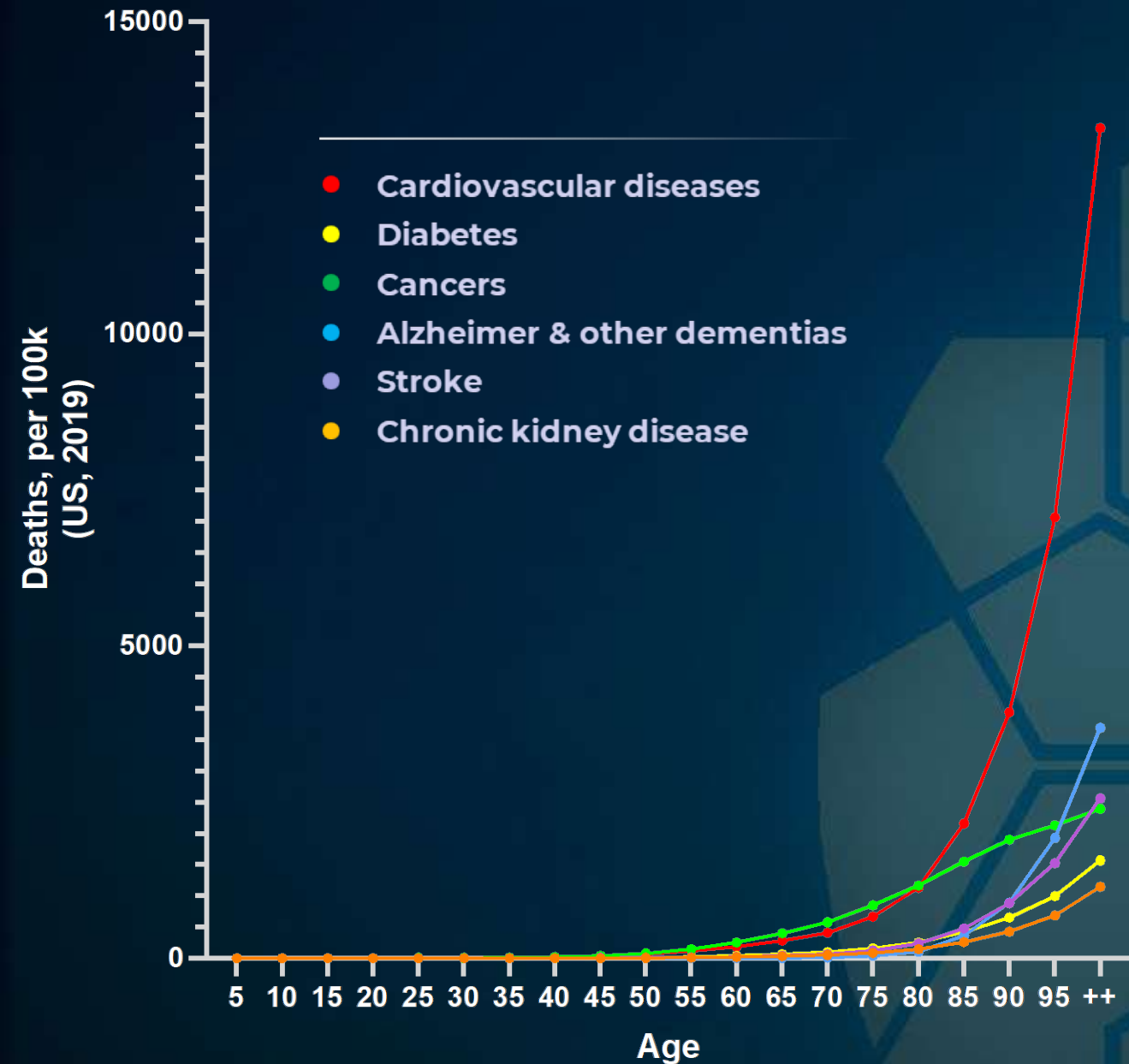
Founder and CEO of Entos Pharmaceuticals

Founder and CSO of OncoSenX

Founder and CEO of Aegis Life

CSO at Oisin Biotechnologies

Incidence of chronic disease increases as we age...



... and carries a significant burden.

- As much as 50% of the global burden of disease is chronic illness (WHO).
- Chronic disease is a significant concern for Canadians, with one-half (51.6%) of the population over the age of 20 having one or more chronic diseases.
- Chronic diseases are estimated to cost Canadians \$68 billion in direct healthcare costs and \$122 billion in productivity losses.



Thanks to CanPath, we have access to BIG HEALTH DATA

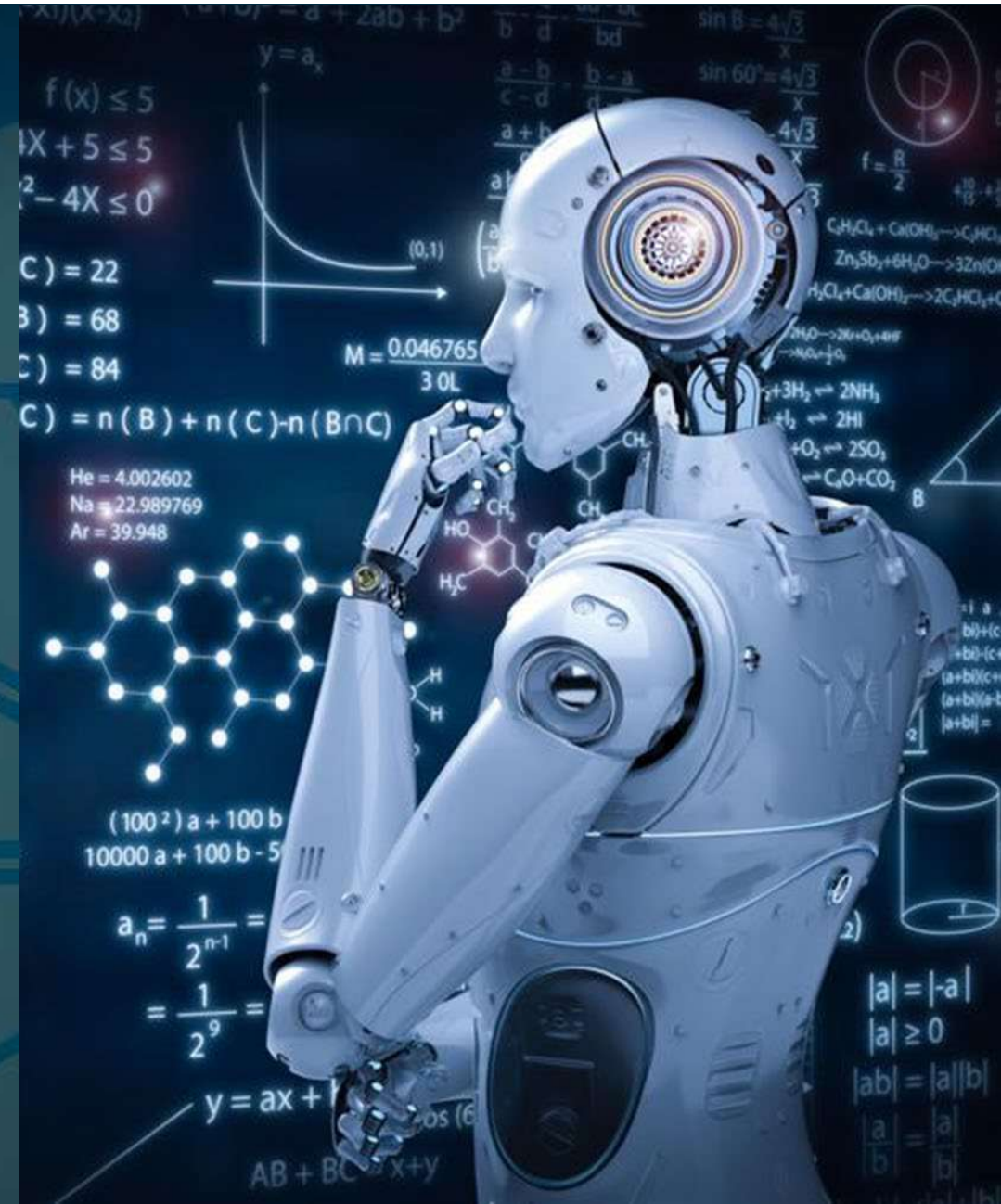
- Comprehensive genomic, clinical, behavioural and environmental data on 330,000 Canadians
- Baseline and follow-up data from five regional cohorts have been harmonized across the country, creating a pan-Canadian resource of more than 2,800 measures of participant health and lifestyle factors.
- Value is increasing over time as new data are added, technology advances, and incident health outcomes are captured.

How do we find patterns in thousands of measures over hundreds of thousands of individuals?

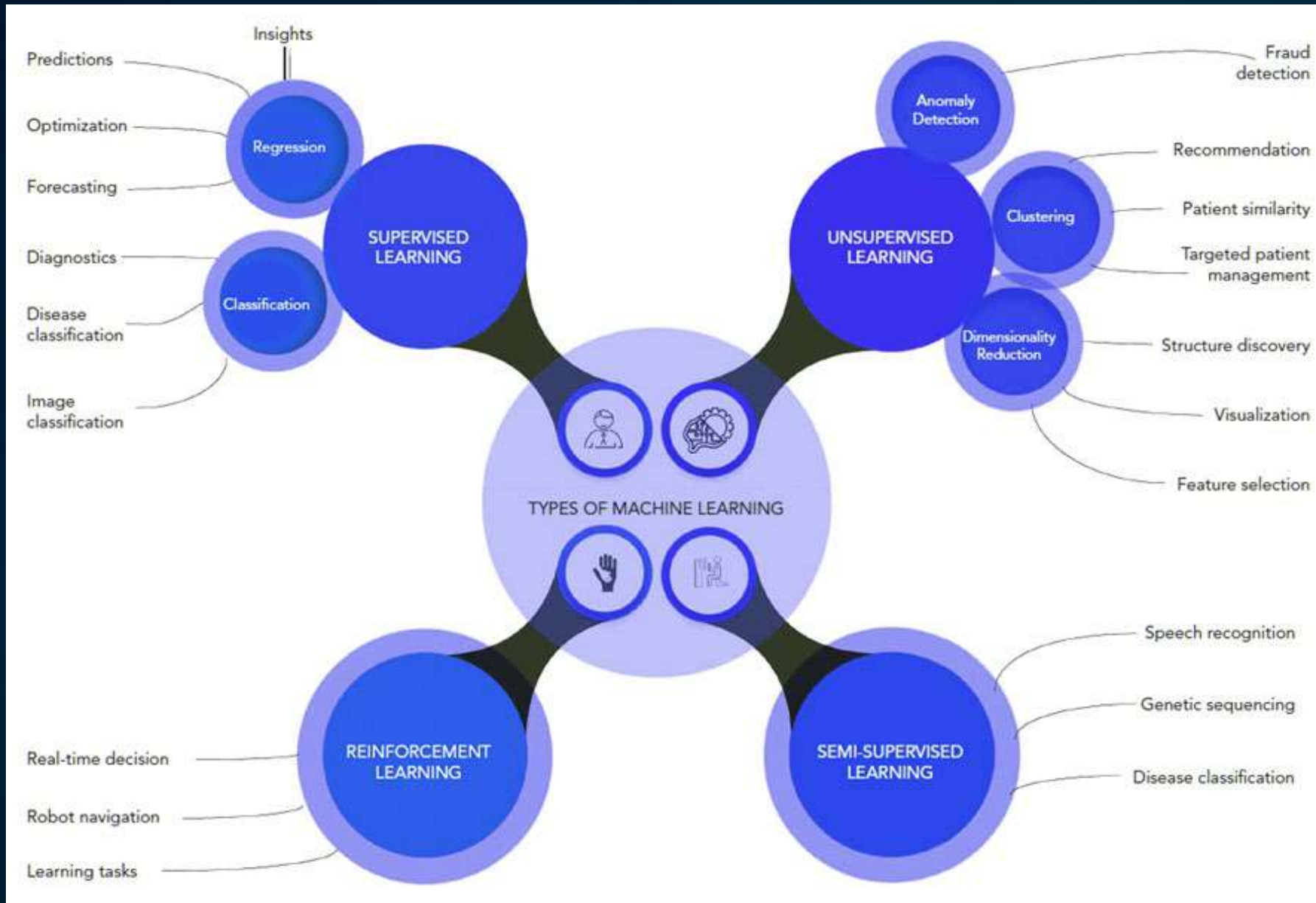
Machine learning (ML or AI)?

- The use and development of computer systems that are able to learn and adapt without following explicit instructions, by using algorithms and statistical models to analyze and draw inferences from patterns in data.

– Oxford Languages, September 29, 2021

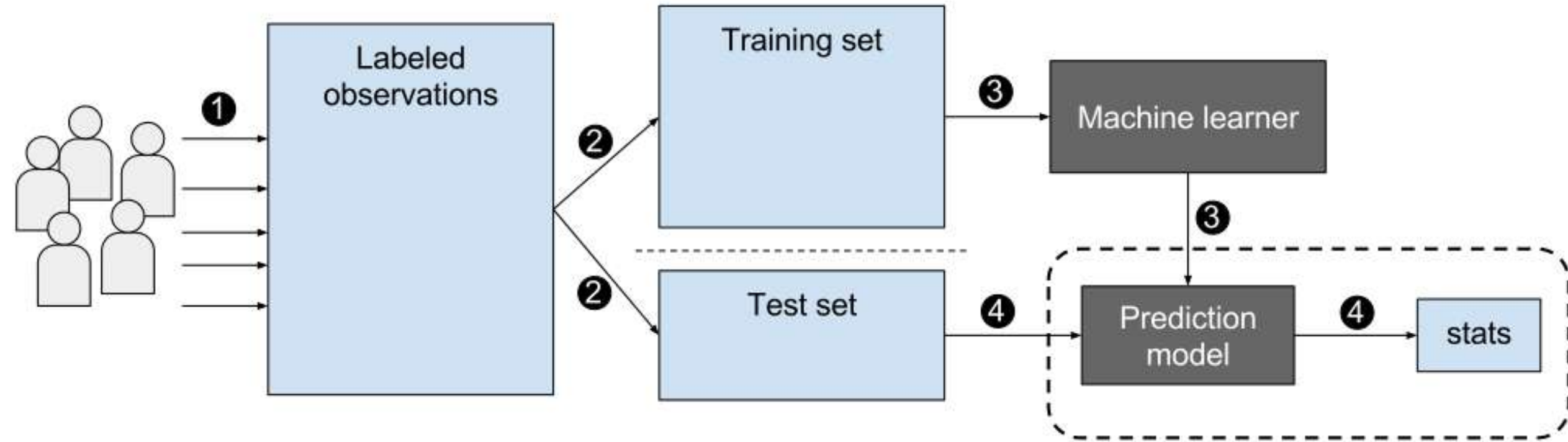


Types of machine learning



Machine learning (ML or AI)?

- Algorithms which use features (e.g., age or lab results) to predict labels (e.g., “no cancer” or “has cancer”) on observations (e.g., patients) which have known values for features and labels.



Machine learning algorithm can make predictions on new data

Many different machine learning algorithms

Linear models

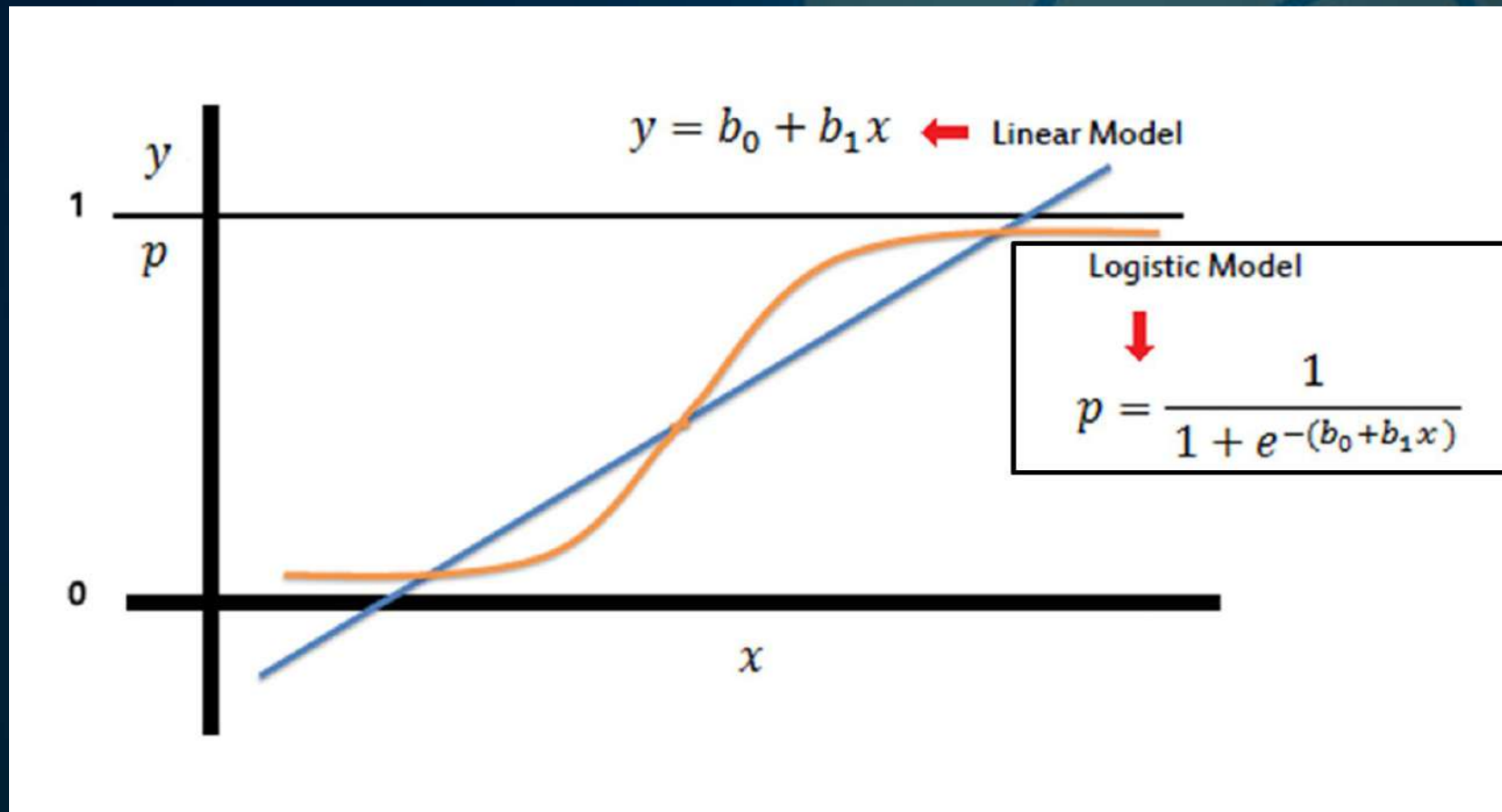
- Logistic regression
- Linear discriminant analysis
- Linear support vector machines

Non-linear models

- K-nearest neighbors
- Decision trees
- Neural networks

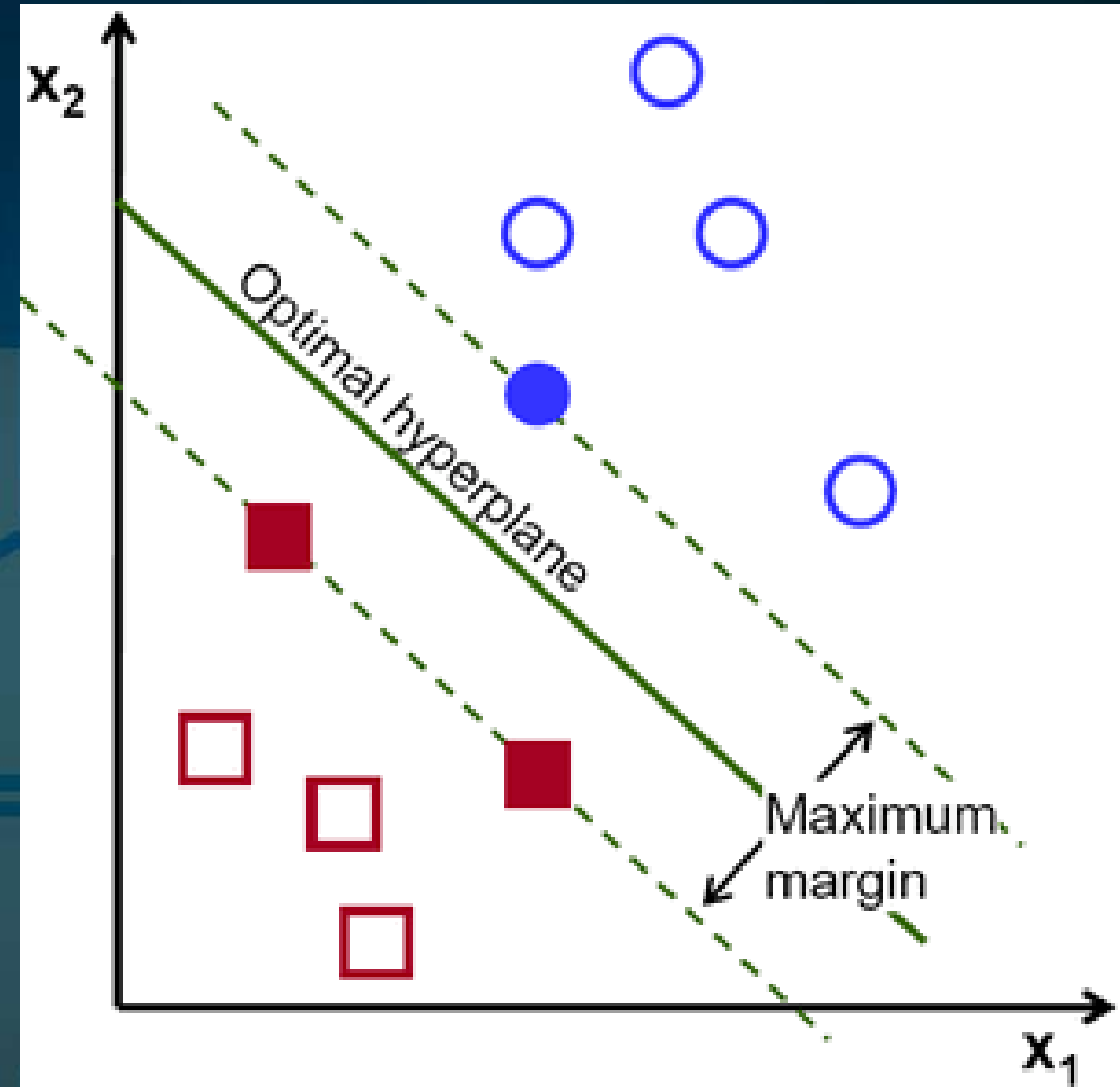
Logistic regression

- Logistic regression is a statistical analysis method to predict a binary outcome, such as yes or no, based on prior observations of a data set.



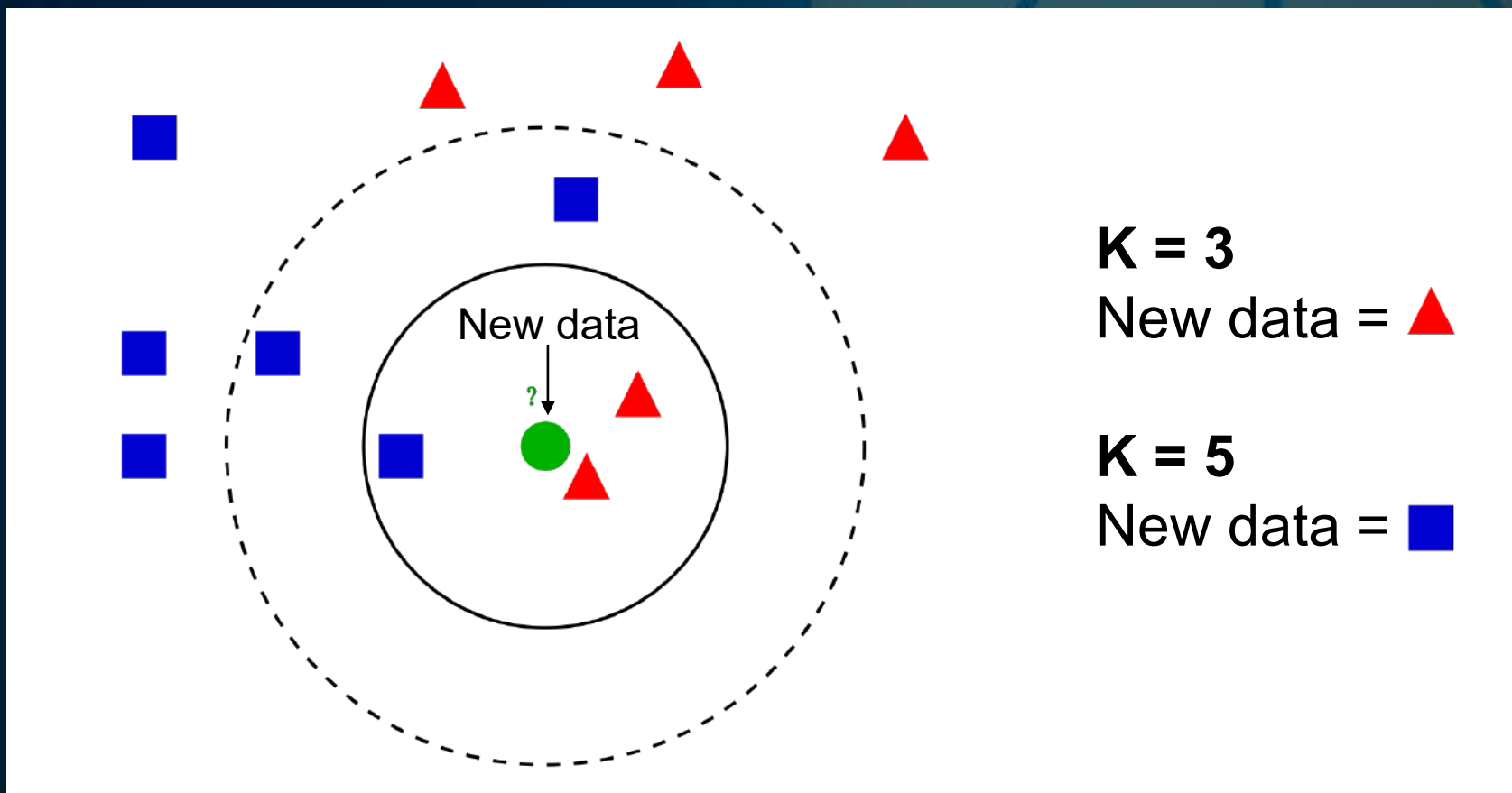
Support Vector Machines (SVM)

- Identifies the hyperplane with maximal distance from points in each group. Points on each side of hyperplane belong to separate groups.



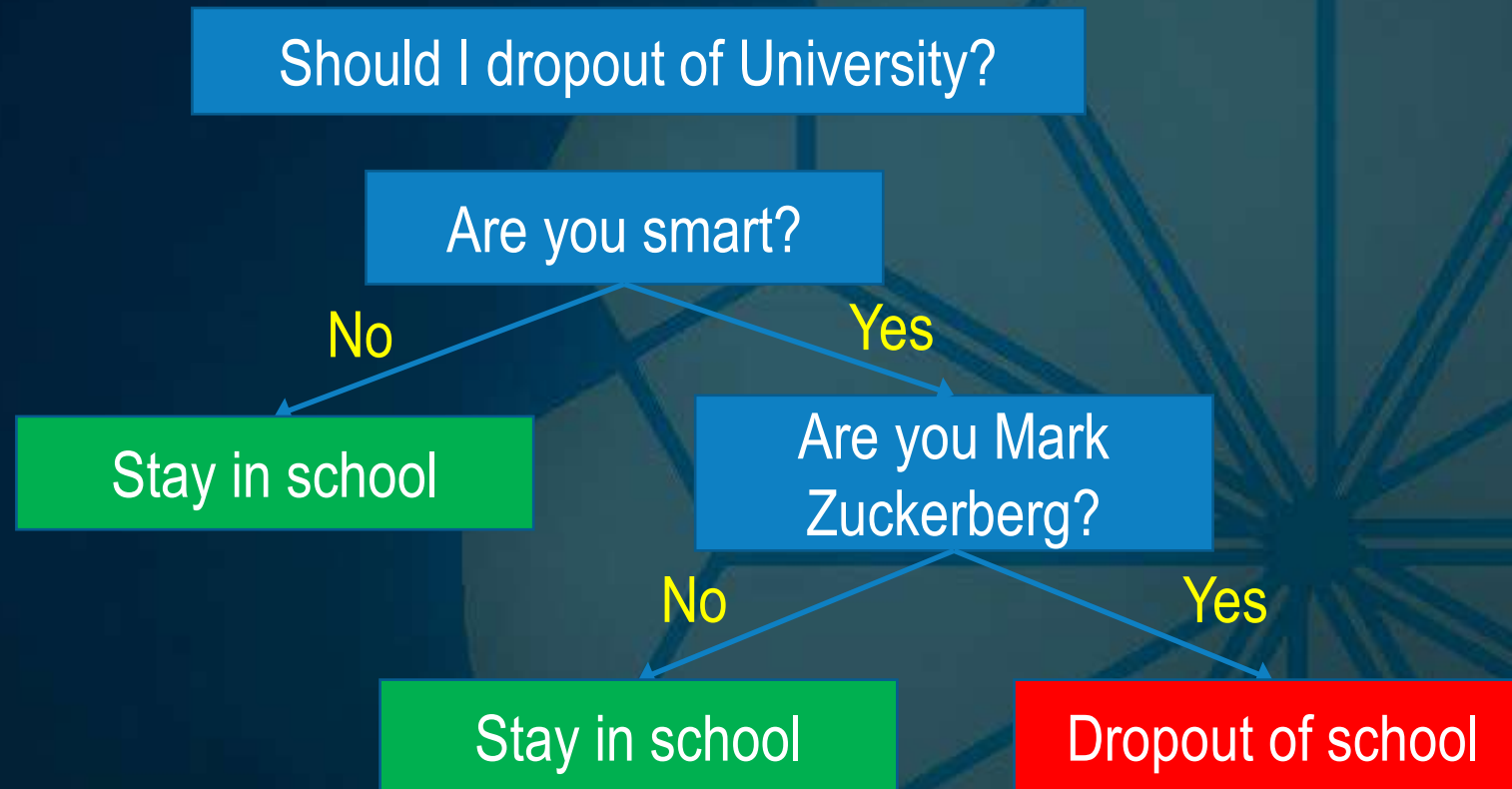
K-nearest neighbors

- Classify new data based on the k-number of training observations the new data was closest to



Decision trees

- Classify new data based on a series of questions with the classification label at the end of the path.



EXtreme GRadient BOosting (XGBoost)

- XGBoost is an ensemble, boosted, decision tree-based model.
- **Ensemble:** Results of multiple decisions trees averaged into 1 result
- **Boosted:** Each additional decision tree is designed to correct misclassified observations

Pros

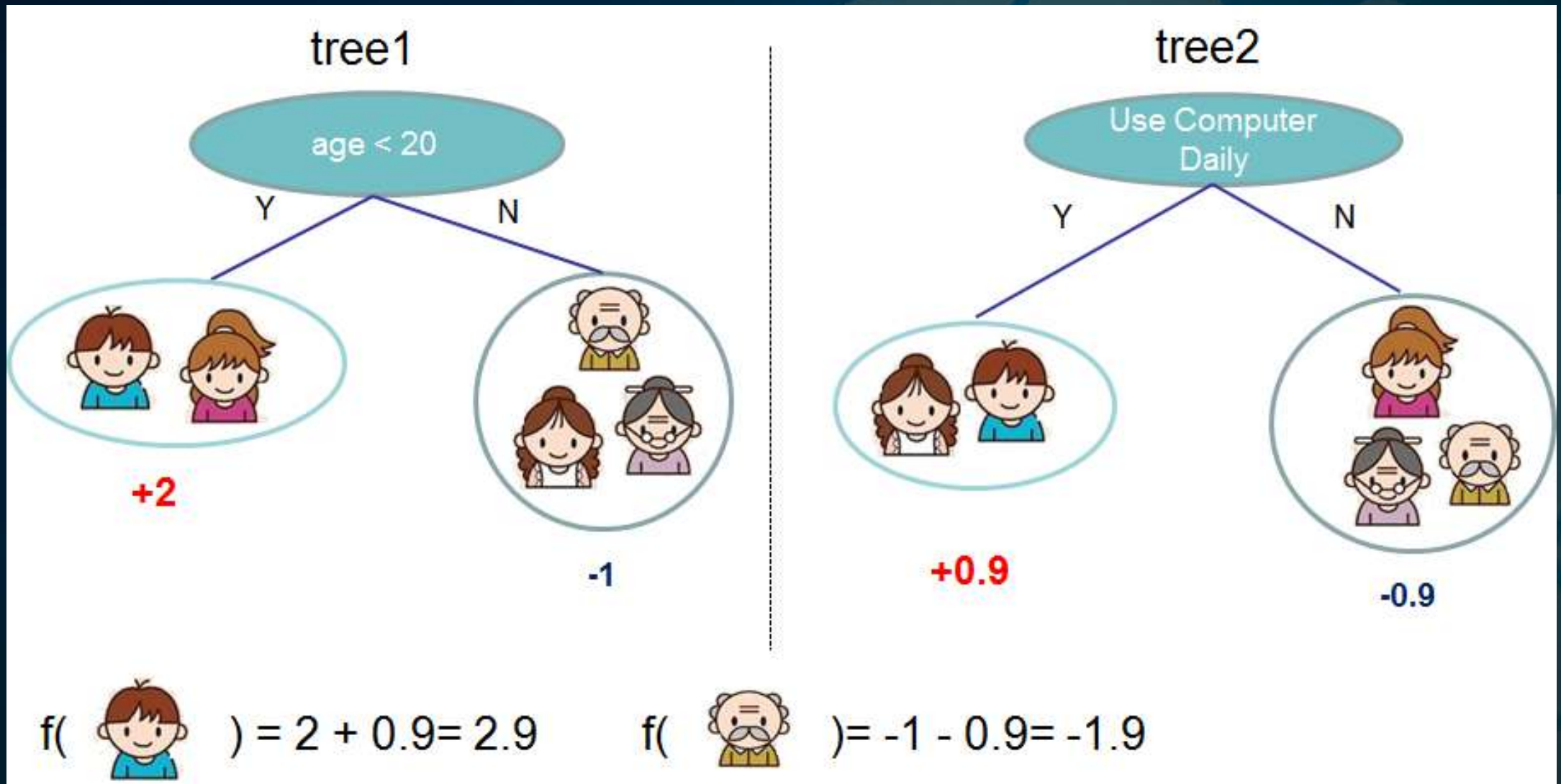
- Very high classification accuracy on tabular data
- Easily handles:
 - Correlative features
 - Features with different scales
 - Missing data
- Regularization helps minimize overfitting
- Very fast training times

Cons

- Models can still overfit training data
- More difficult to explain model

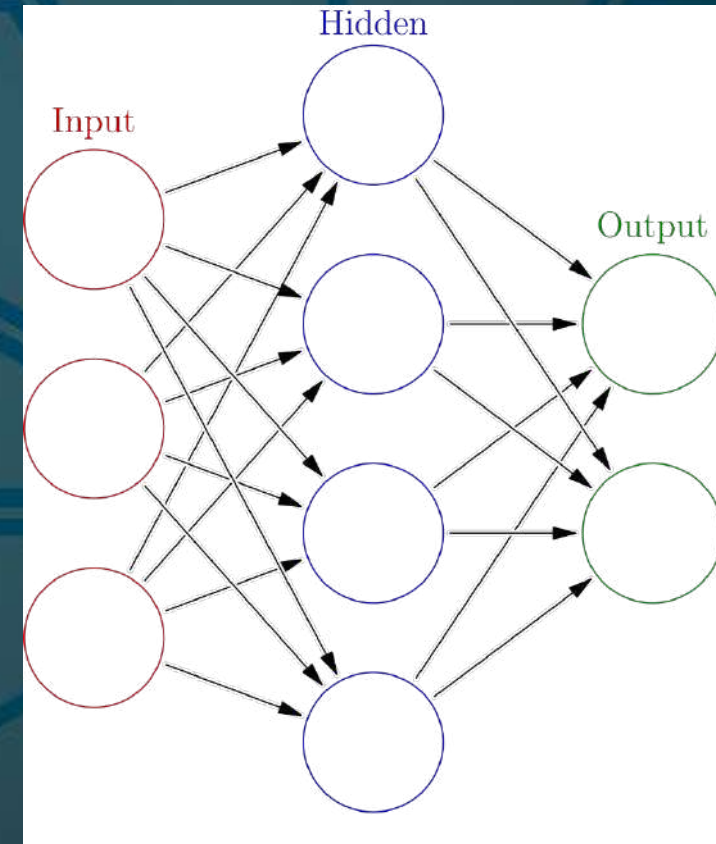
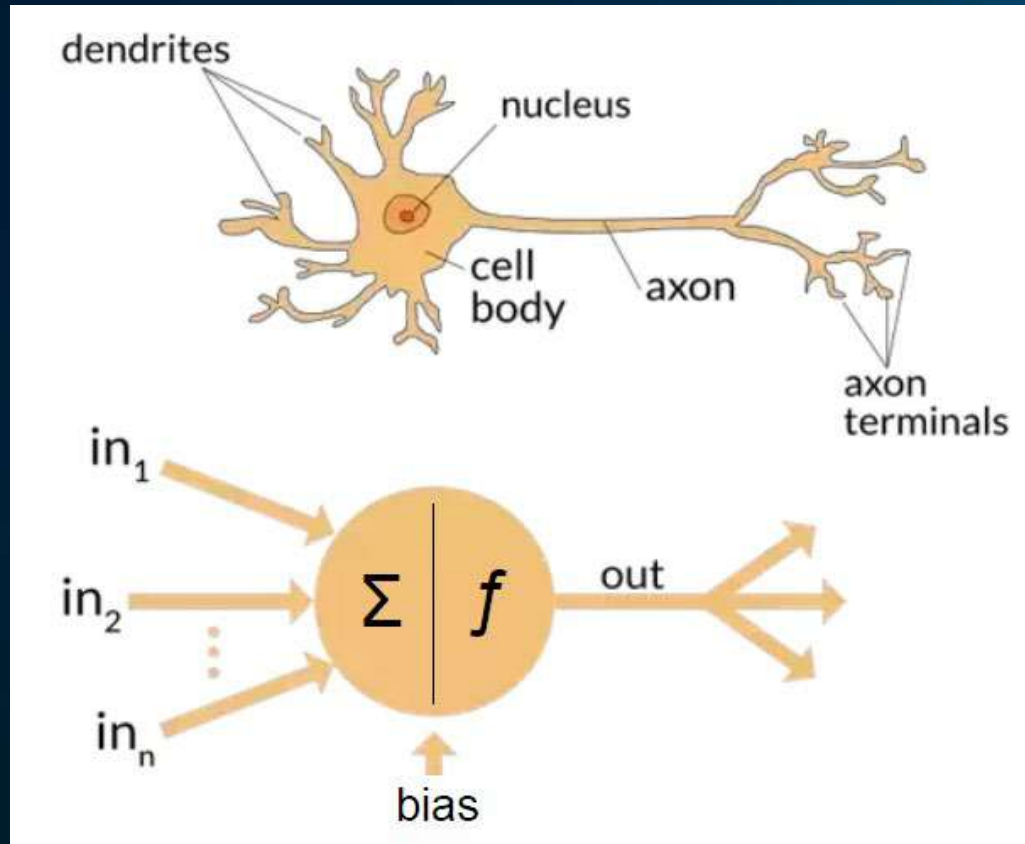
Boosting combines multiple trees to improve prediction

Does this person like video games?



Artificial Neural Network (ANN)

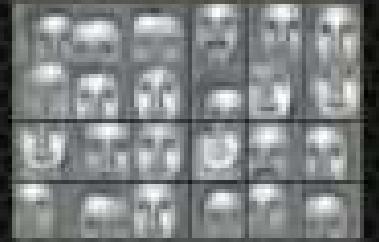
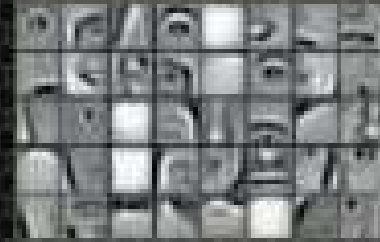
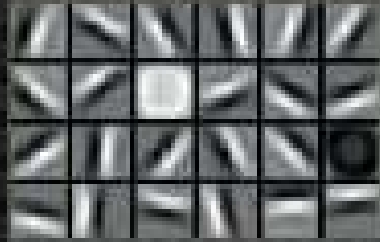
- Inspired by biology, ANN have multiple connected “neurons”. Each neuron has input values multiplied by weights which are summed and modified by a transfer function. This output value is the input value for downstream neurons.



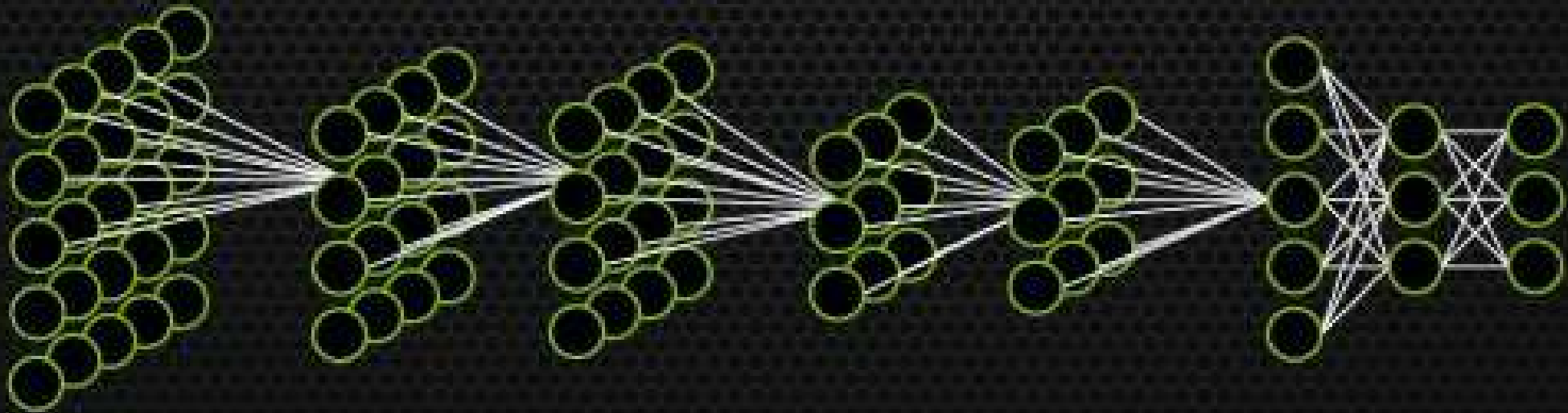
Convolutional Neural Networks (CNN)

Early layers learn edges/colors

Subsequent layers combine earlier layers to identify useful shapes for classification

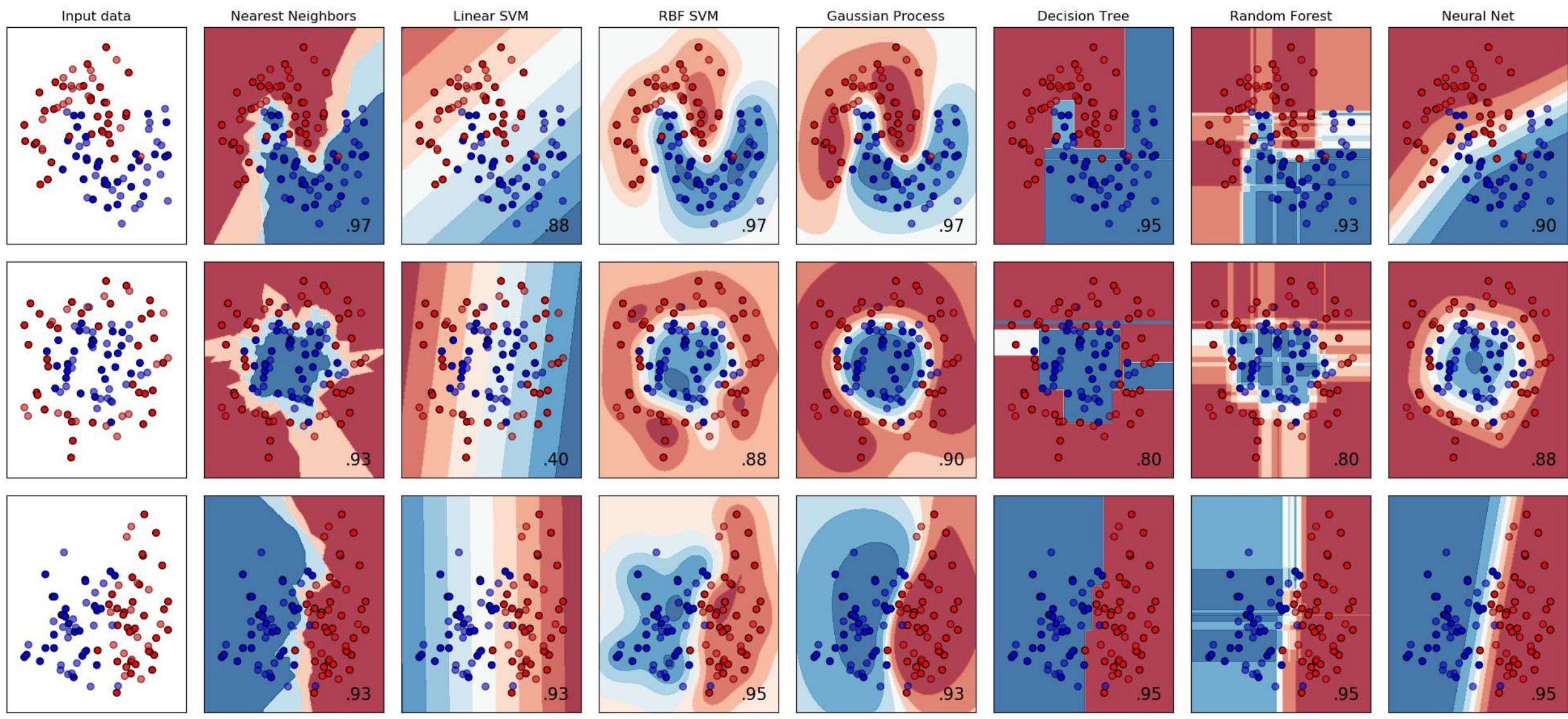


Image

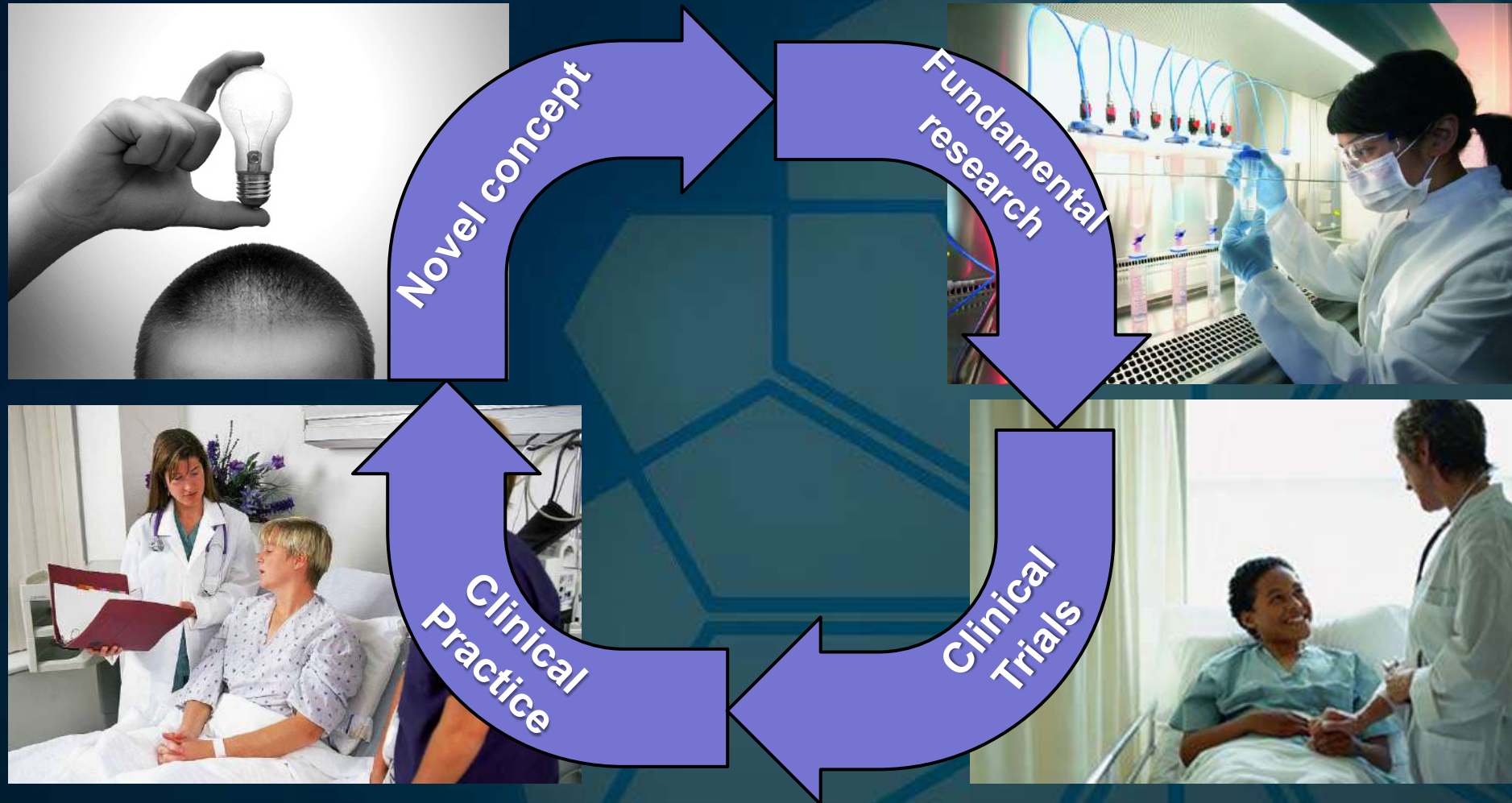


"Sara"

Performance of ML algorithms depends on the data



How can we ultimately make an impact on patients?

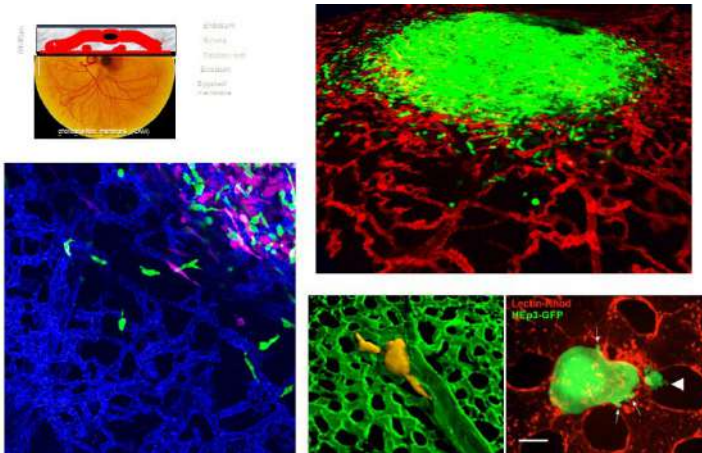


Our clinical translation ecosystem

Discovery Research

The Lewis Laboratory

- Mechanisms of metastasis
- Novel diagnostics
- Nanotechnology for imaging and therapy



Translational Research



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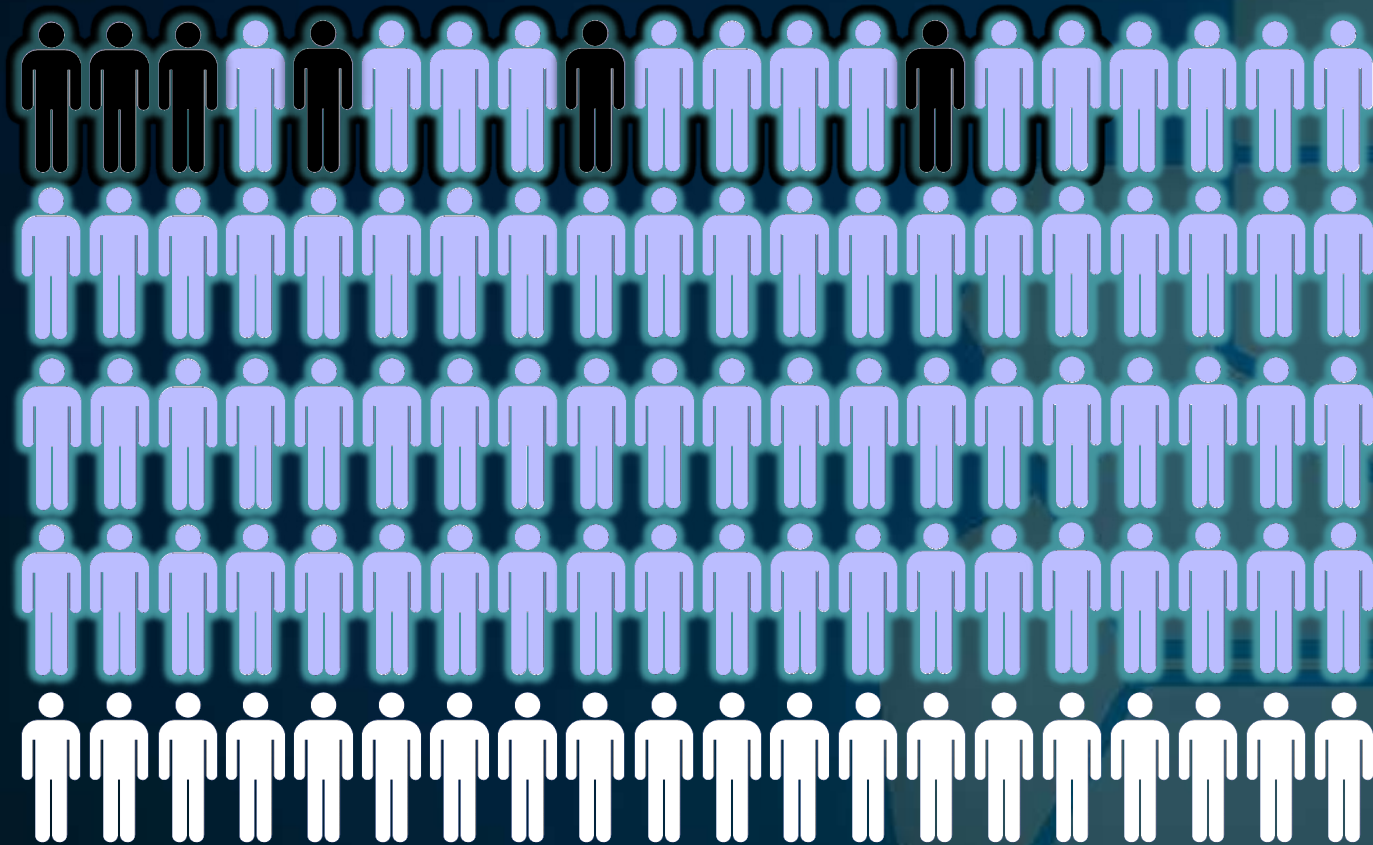
- Founded in 2013 by a provincial multi-disciplinary team of prostate cancer scientists, physicians, patients, healthcare employees and an international collaborative network
- Facilitate translational prostate cancer research
- Accelerate the translation of research from the laboratory to the clinic

Commercialization



Prostate cancer is the most commonly diagnosed cancer in men

36% of newly diagnosed cancers, and 10% of all cancer deaths in men



Out of every 100 men...

16 will be diagnosed with prostate cancer in their lifetime

In reality, up to 80 will have prostate cancer by age 70

And 3 will die from it.

But which 3 ?

The deadliest aspect of cancer

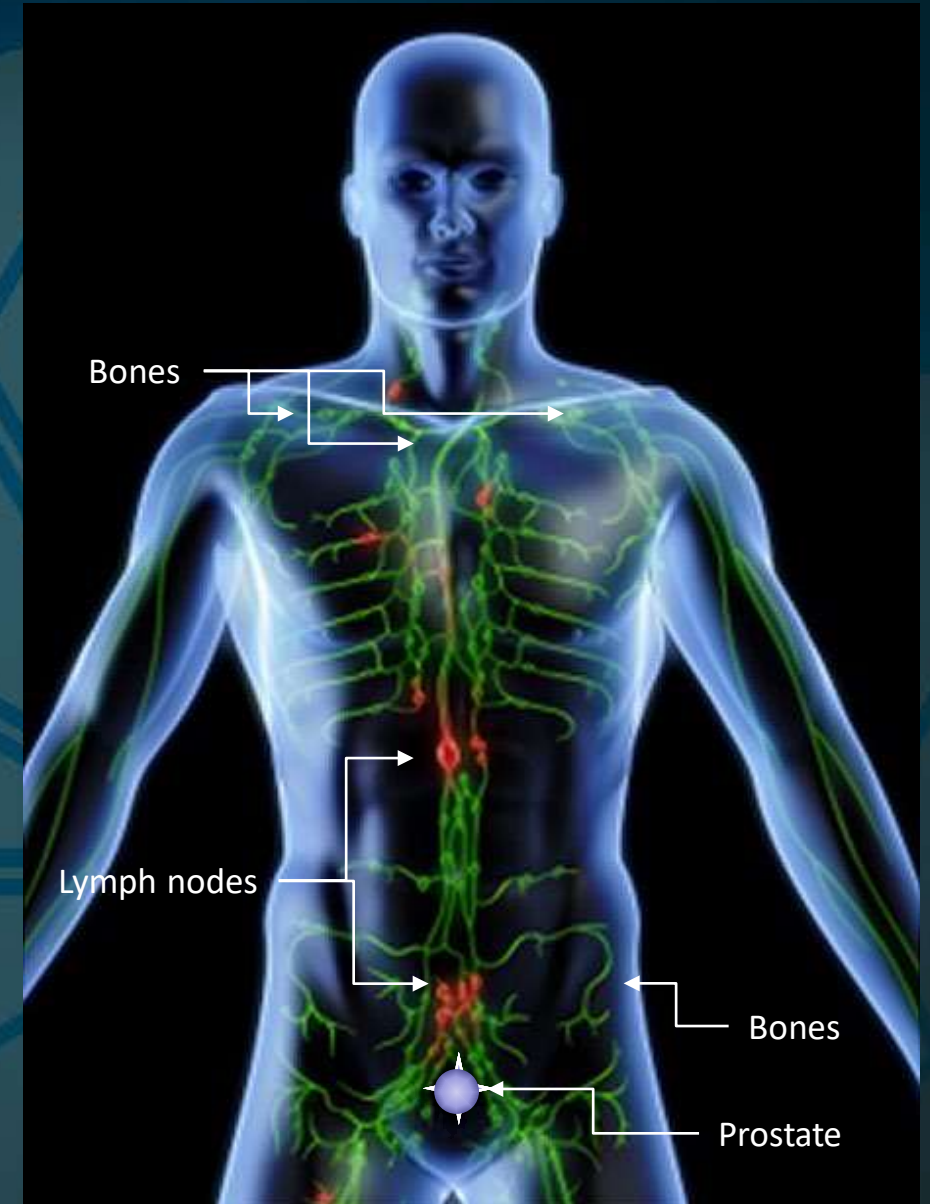
The deadliest aspect of prostate cancer is its spread, or metastasis

In North America, the average 5 year survival rate for localized prostate cancer is 100%

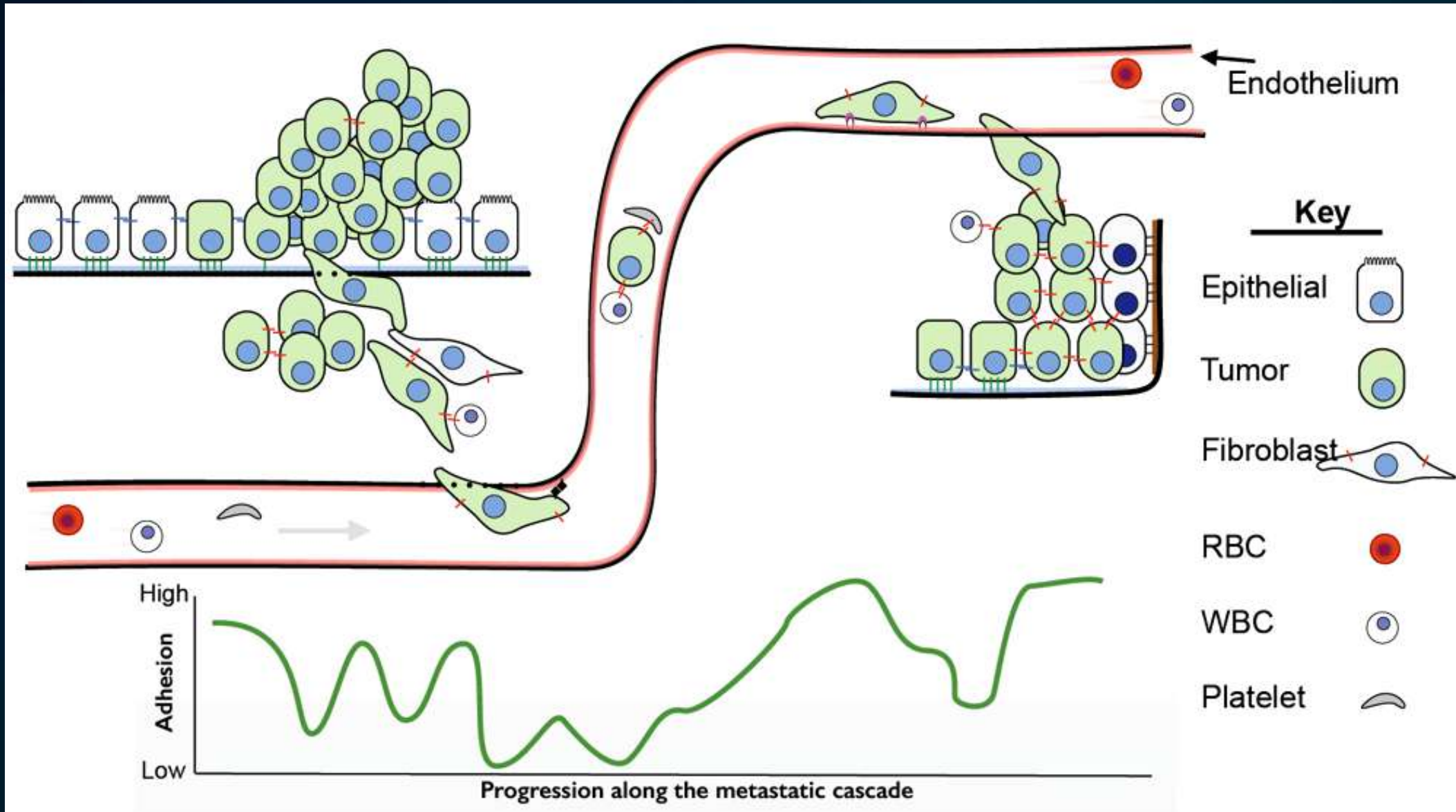
For metastatic cancer, it is less than 30%

Current diagnosis tools do not predict whether metastasis will occur

Current treatments do not prevent or cure metastasis

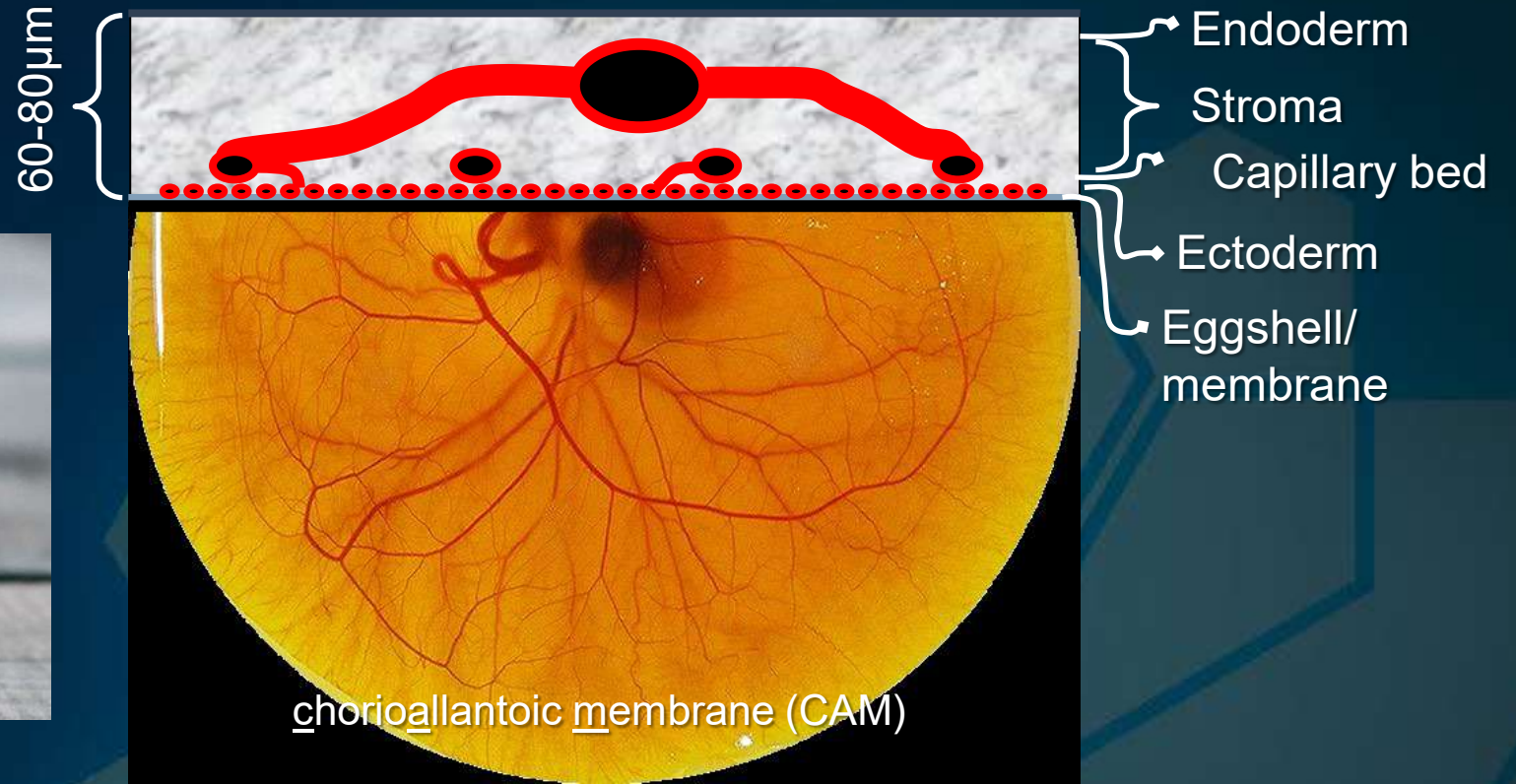


Metastasis is a complex, multi-step process



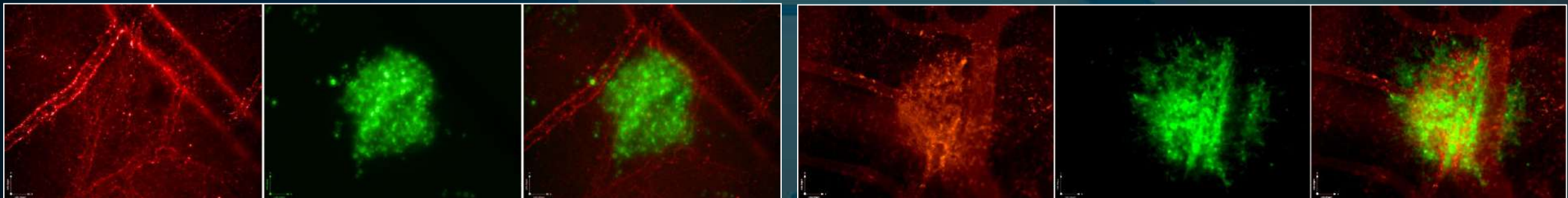
Modeling cancer dynamics in chicken embryos

Leong et al., *Nature Protocols* 2010

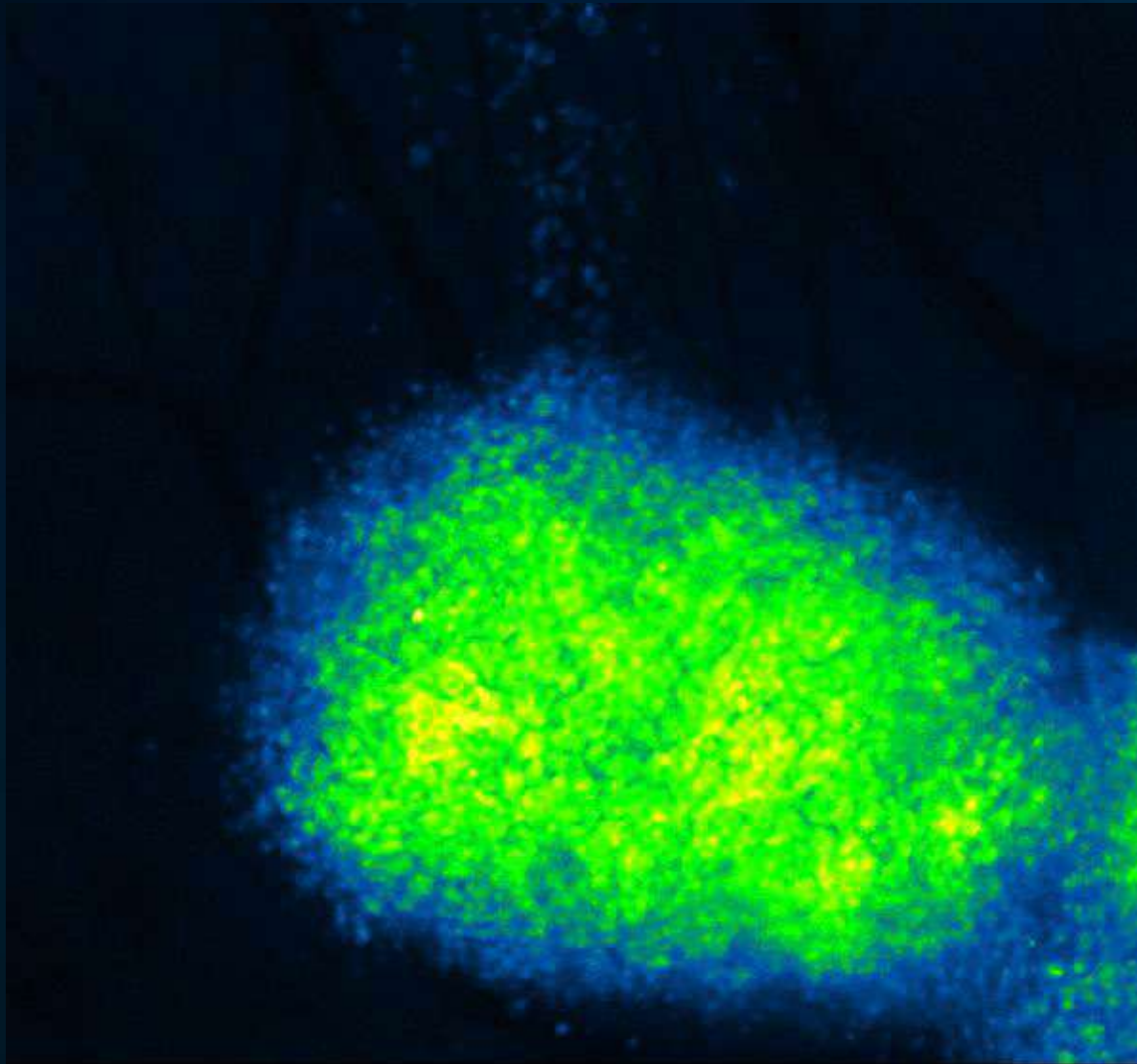


Immediately after injection

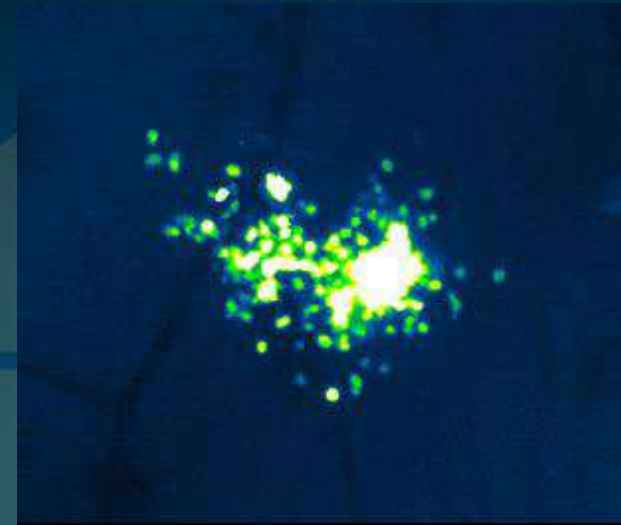
24 hours later



Intravital imaging of tumour growth and metastasis

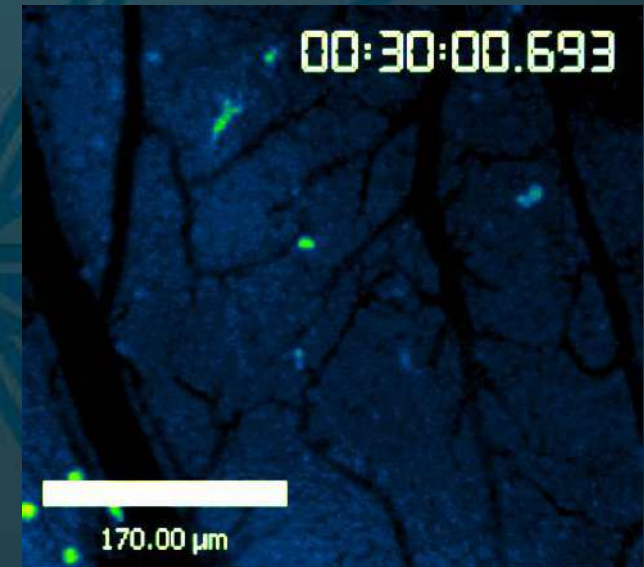


4 mm tumour growing over 4 days



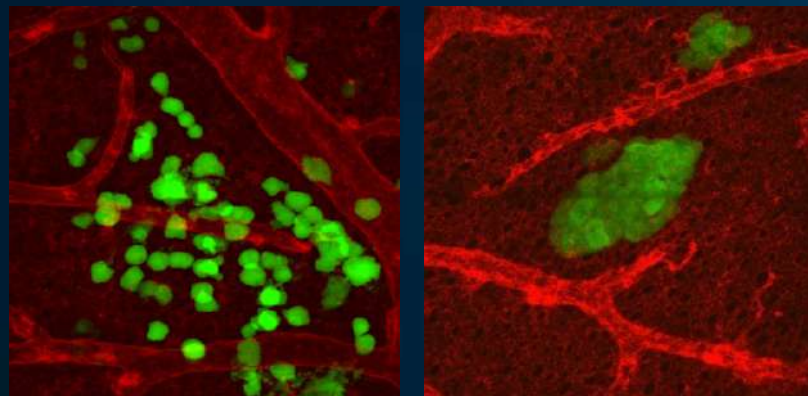
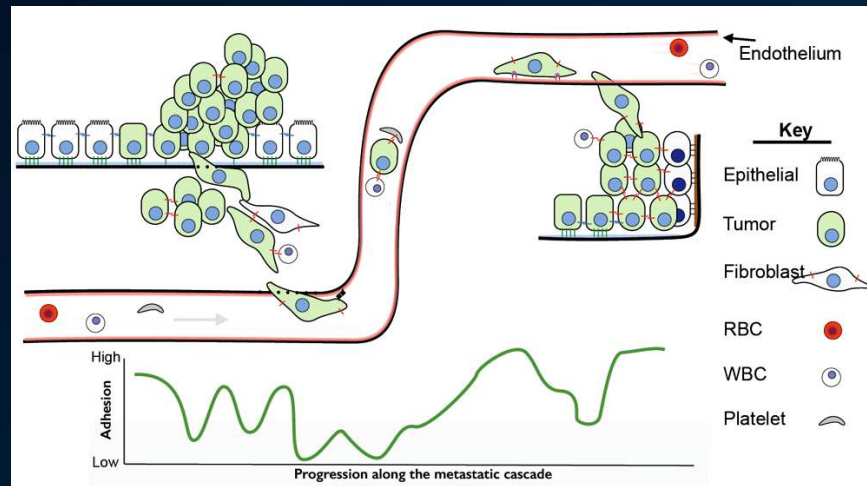
Small metastasis
of 200-300 cells
(undetectable!)

Individual cancer
cells
(completely
undetectable!)



170.00 μm

We have identified biomarkers associated with cancer metastasis



Biomarkers related to metastasis were discovered using advanced microscopy approaches

nature COMMUNICATIONS

ARTICLE

DOI: 10.1038/44447-09-04743-2 OPEN

Quantitative in vivo whole genome motility screen reveals novel therapeutic targets to block cancer metastasis

Konstantin Stoletov¹, Lian Willetts¹, Robert J. Pazroski¹, David J. Bond¹, Srijan Raha¹, Juan Jovel^{2,3}, Benjamin Adam⁴, Amy E. Robertson¹, Francis Wong¹, Emma Woolner¹, Deborah L. Sosnowski¹, Tarek A. Bismar⁵, Gene Ka-Shu Wong^{2,3,6}, Andries Zijlstra⁷ & John D. Lewis¹

Cell Reports Article

Invadopodia Are Required for Cancer Cell Extravasation and Are a Therapeutic Target for Metastasis

Hon S. Leong,¹ Amy E. Robertson,¹ Konstantin Stoletov,² Sean J. Leith,¹ Curtis A. Chin,¹ Andrew E. Chien,¹ M. Nicole Hague,² Amber Ablack,² Katia Cammine-Simmon,² Victor A. McPherson,¹ Carl O. Postenka,² Eva A. Turley,^{2,4} Sara A. Courtenidge,⁵ Ann F. Chambers,² and John D. Lewis^{1*}

¹Translational Prostate Cancer Research Group, London Regional Cancer Program, 790 Commissioners Road East, London ON N6A 4L6, Canada

Cancer Cell Article

The Inhibition of Tumor Cell Intravasation and Subsequent Metastasis via Regulation of In Vivo Tumor Cell Motility by the Tetraspanin CD151

Andries Zijlstra,^{1,2,*} John Lewis,^{1,4} Bernard DeGryse,² Heidi Stuhlmann,^{1,5} and James P. Quigley^{1,6}

¹Department of Cell Biology, The Scripps Research Institute, 10550 North Torrey Pines Road, La Jolla, CA 92037, USA

²BFCM, FIRCO Institute of Molecular Oncology, Via Adamello 16, 20139 Milan, Italy

³Present address: Department of Pathology, Vanderbilt University, Nashville, TN 37232-2561, USA.

⁴Present address: London Regional Cancer Program, University of Western Ontario, London, Ontario N6A 4L6, Canada.

⁵Present address: Department of Cell and Developmental Biology, Weill Cornell Medical College, New York, NY 11065, USA.

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DOI: 10.1016/j.ccr.2008.01.031

nature COMMUNICATIONS

ARTICLE

Received 8 Dec 2016 | Accepted 24 Feb 2017 | Published 24 Apr 2017 DOI: 10.1038/ncomms15559 OPEN

LPP is a Src substrate required for invadopodia formation and efficient breast cancer lung metastasis

Elaine Ngan^{1,2}, Konstantin Stoletov³, Harvey W. Smith^{1,4}, Jessica Common^{1,2}, William J. Muller^{1,2,4}, John D. Lewis³ & Peter M. Siegel^{1,2,4,5}

Molecular and Cellular Pathobiology Cancer Research

Integrin-Free Tetraspanin CD151 Can Inhibit Tumor Cell Motility upon Clustering and Is a Clinical Indicator of Prostate Cancer Progression

Trenis D. Palmer¹, Carlos H. Martinez⁴, Catalina Vasquez², Katie E. Hebron^{1,2}, Celestial Jones-Paris¹, Stelma A. Arnold¹, Susanne M. Chan⁴, Veru Chalasani¹, Jose A. Gomez-Lemus⁵, Andrew K. Williams¹, Joseph L. Chin¹, Giovanna A. Giannico¹, Tatiana Ketova¹, John D. Lewis², and Andries Zijlstra^{1,2}

The Prostate 72:825–833 (2012)

Ghrelin Receptor as a Novel Imaging Target for Prostatic Neoplasms

Chen Lu,^{1,4} Mark S. McFarland,² Rae-Lynn Nesbitt,^{1,3} Andrew K. Williams,^{1,4} Susanne Chan,⁵ Jose Gomez-Lemus,⁵ Anna Maria Autran-Gomez,^{1,4} Ali Al-Zahrani,^{1,4} Joseph L. Chin,^{1,4} Jonathan I. Izawa,^{1,4} Leonard G. Luyt,^{2,6} and John D. Lewis^{1,3,4}

¹Translational Prostate Cancer Research Group, London Regional Cancer Program, London, Ontario, Canada

Open

ORIGINAL ARTICLE

The interaction between caveolin-1 and Rho-GTPases promotes metastasis by controlling the expression of alpha5-integrin and the activation of Src, Ras and Erk

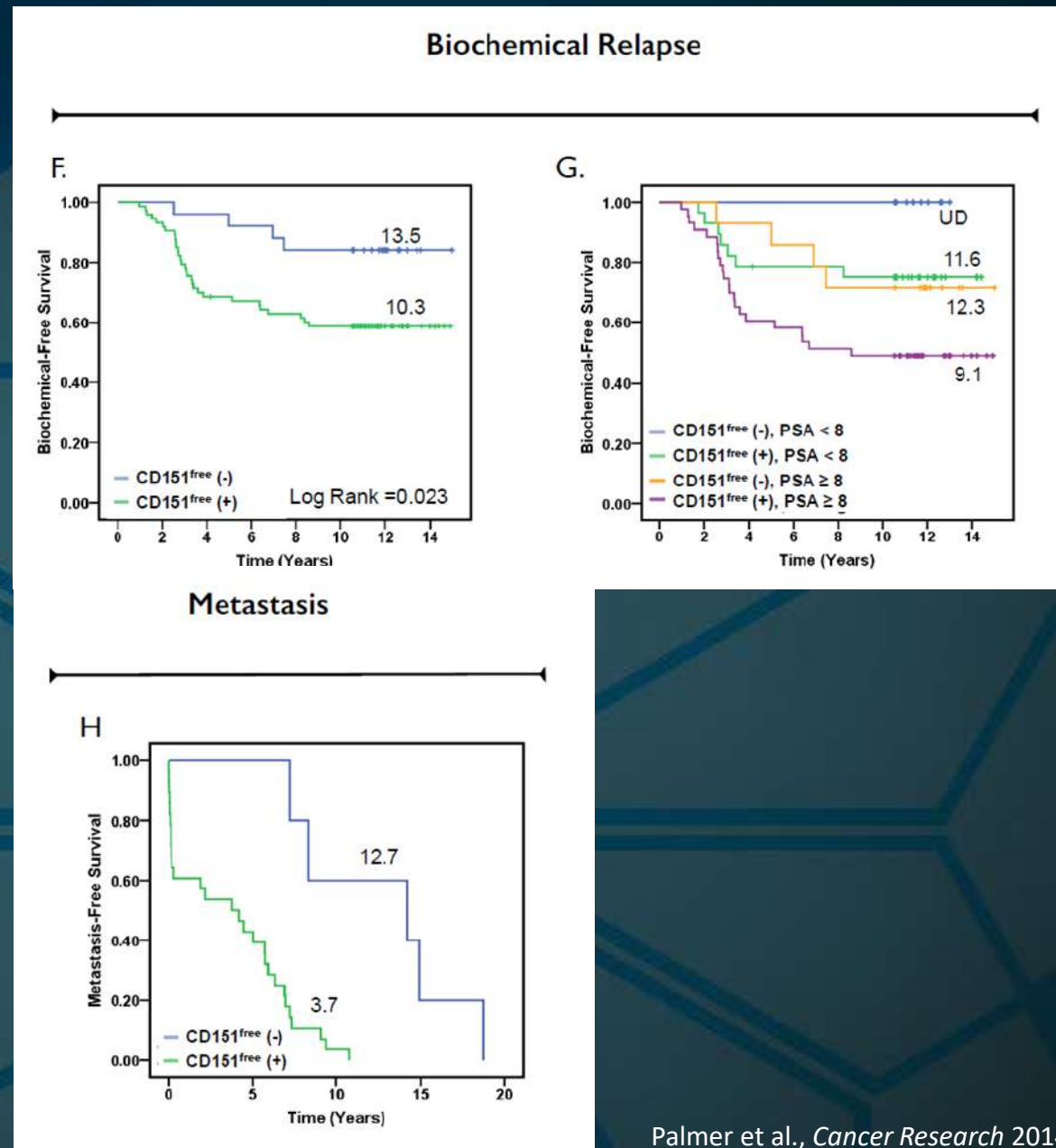
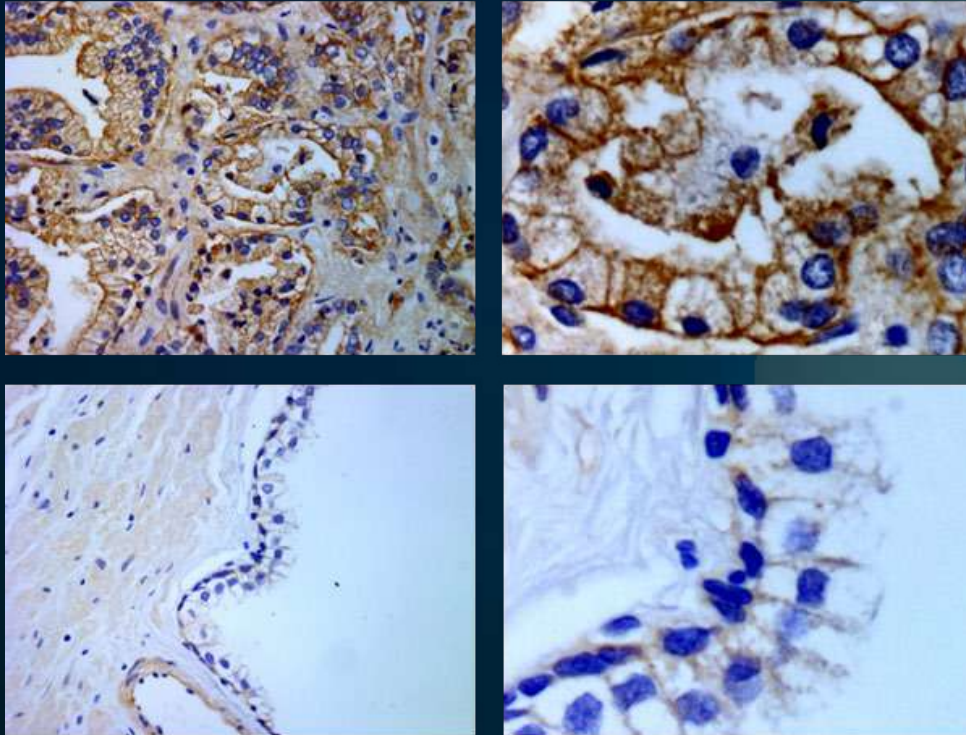
E. Arpaia^{1,2,6}, H. Blaser^{1,6}, M. Quintela-Fandino^{1,3,6}, G. Duncan¹, HS Leong⁴, A. Ablack⁴, SC Nambiar⁴, EF Lind¹, J. Silvester¹, CK Fleming¹, A. Rufini⁵, MW Tusche⁴, A. Brüstle⁴, PS Ohashi^{1,2}, JD Lewis⁴ and TW Mak^{1,2}

CD151^{free} predicts prostate cancer recurrence and metastasis

138 prostate cancer surgery patients

Follow up: 12.1 years
Recurrence: 34 cases
Metastasis: 38 cases

mAb 1A5



Detection of disease alone is not sufficient

What is the need for intervention?



Most new tests identify gene mutations that are simply **associated** with disease, not driving the disease itself

- Will identified mutations be clinically significant?
- Will a potential disease indication ever need intervention?

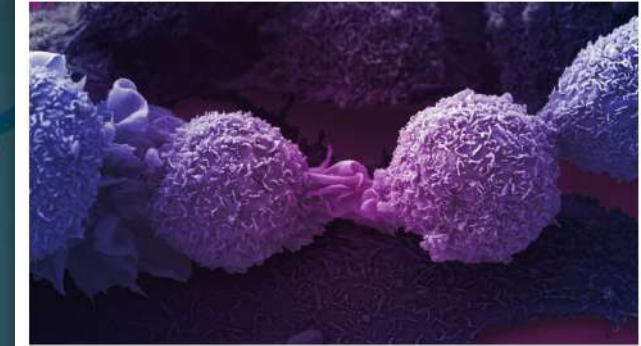
“Genomic testing is revolutionizing disease screening”



Gene screening exacerbates the problem of over diagnosis without providing clear guidance on the need for intervention, which dramatically increases over-treatment, reduces patient quality-of-life and increases healthcare costs

Blame rising cancer overdiagnosis on ‘irrational exuberance’ for early detection

By H. Gilbert Welch Oct. 2, 2010



Lung cancer cells
ANNE WESTON/FRANCIS CRICK INSTITUTE/WELLSOME

Future Medicine Ltd
Personalized Medicine
Volume 15, Issue 5, September 2018, Pages 343-346
<https://doi.org/10.2217/pme-2018-0041>

Editorial



Too much of a good thing?
Overdiagnosis, or overestimating risk in
preventive genomic screening

Karen M Meagher¹ and Jonathan S Berg²

¹Department of Social Medicine, University of North Carolina at Chapel Hill, Chapel Hill, NC 27599-7240, USA

²Department of Genetics, The University of North Carolina at Chapel Hill, Chapel Hill, NC, 27599-7264, USA

*Author for correspondence. +1 919 966 7043; jonathan_berg@med.unc.edu

Bullard and Chiolero *Public Health Reviews* (2015) 368
DOI: 10.1186/s40985-015-0012-1

Public Health Reviews

COMMENTARY

Open Access

Screening and overdiagnosis: public health implications

Jean-Luc Bulliard^{1*} and Arnaud Chiolero^{1,2}



OVERVIEW of APCaRI

The Alberta Prostate Cancer Research Initiative (APCaRI) is a multi-disciplinary team of:

- Prostate cancer researchers
- Physicians & Clinical research scientists
- Patients
- Industry partners and
- Not for profit networks
- Hospitals & Diagnostic labs



We are working together!




Researchers and Clinicians to positively impact the outcomes and quality of life of those living with prostate cancer by accelerating the translation of new research ideas from the laboratory to the clinic.



Our network of clinical teams






The Northern Alberta Urology Clinic

Dr. Adrian Fairley
APCaRI Co-PI

Dr. Michael Chetner

Dr. Gerald Todd

Ruth Fazio
Research Coord.

Yulya Kuchytka and Homaira H-Mehrabani
Clinical Data Entry Clerks/

+ a team of Urologists, MOAs and Clinic Administrator... all committed to the success of the study






Northern Alberta Urology Centre – Edmonton



Cross Cancer Institute – Edmonton




The Cross Cancer Institute

Dr. Scott North

Dr. Nawaid Usmani

Dr. Michael Kolinisky

Dr. Brita Danielson

Dr. Matthew Purlanant

Dr. Teena Makowichuk

+ a team of Medical and Radiation Oncologists and the CCI laboratory!!

The Tom Baker Cancer Centre





Dr. Jackson Wu

Dr. Dean Ruether

Dr. Daniel Heng




Dr. Harvey Quon

Dr. Winson Cheung

+ a team of Medical and Radiation Oncologists




Tom Baker Cancer Centre – Calgary



Prostate Cancer Centre – Calgary

The Prostate Cancer Centre


Dr. Bryan Donnelly
APCaRI Co-PI

Dr. Eric Hyndman

Dr. Geoffrey Gorto





Lorael Mendoza
Clinical Research Coordinator

Gabriela DaSilveira & Madison Turk
Clinical Data Entry Clerks & Lab Assistant

+ a team of Nurses, Lab Assistants, Clinical Research personnel, Urologists, MOAs and PCC Director... all committed to the success of the study!!

APCaRI 06

Bladder Cancer Dx test

Recruitment: Dec 2021

APCaRI 01 & 03

Registry & Biorepository

3875 participants recruited in AB

APCaRI 02 & 04

Single Cell Genomics
Familial Prostate Cancer

18 participants recruited in AB

A Study of Men With Advanced Prostate Cancer in Canada (GURC)

Document the course of advanced prostate cancer: disease progression, real-world treatment, and patient management

30 participants recruited in ED

TRUE^{NTH}
A NOVEMBER INITIATIVE



GLOBAL REGISTRY
PROSTATE CANCER OUTCOMES

Improve outcomes and quality of life of men with prostate cancer and their families.

4978 participants recruited in AB

APCaRI Studies

Metformin Active Surveillance Trial (MAST) Study

Effect of Metformin to avoid PCa Progression

32 participants recruited in ED



IRONMAN

International Registry for Men with Advanced Prostate Cancer.

EMPRO

50 participants recruited in ED

APCaRI 05

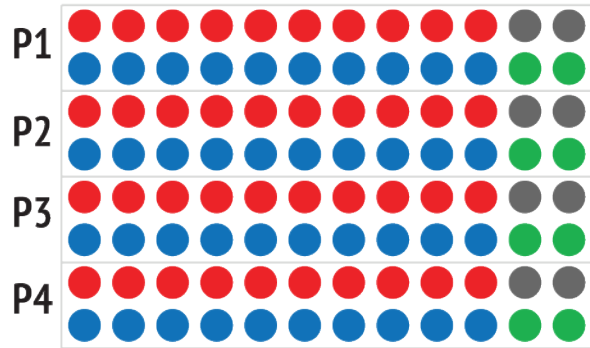


ClarityDX Validation Study

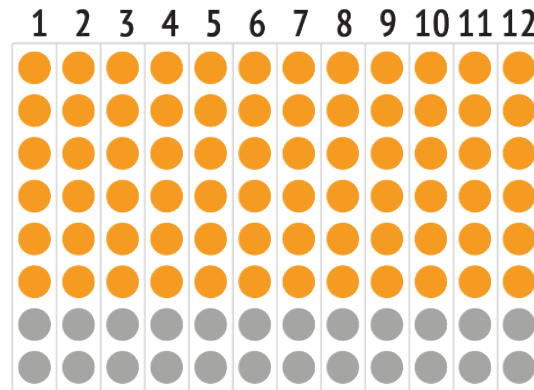
1556 patients recruited in prospective clinical study in AB

APCaRI Biorepository: a prospective pre-diagnosis cohort in Alberta

Blood: **Serum**, **plasma**,
buffy coat and **red blood cells**



Urine and semen



2D Barcode System

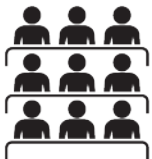


Storage at



CANADIAN BIOSAMPLE REPOSITORY

Samples collected and processed – 64 aliquots per patient



2700
participants enrolled



>150,000
sample aliquots



2 hours
from arm to freezer
in 2h -80°C



1544 fields
demographic and
clinical outcome data



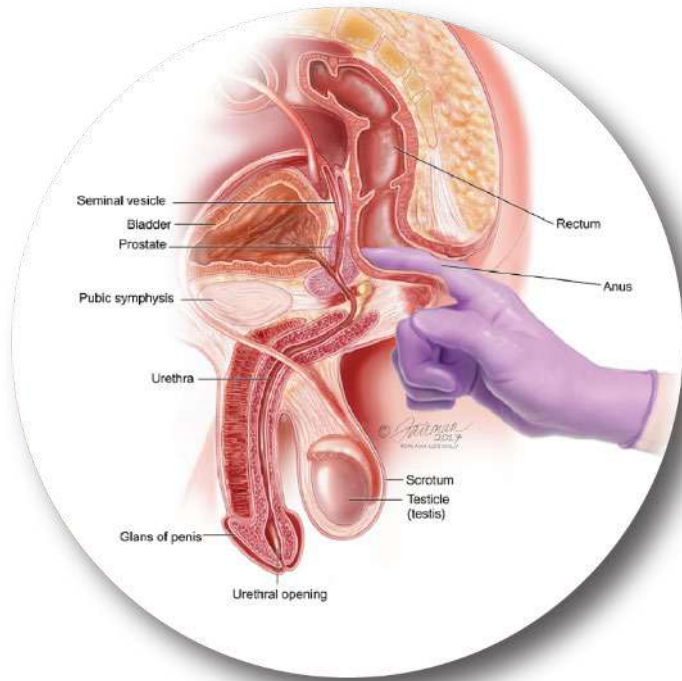
3266
Aliquots distributed
for biomarker validation



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<http://apcari.ca/our-research/biorepository/>

Screening for prostate cancer causes unnecessary harm

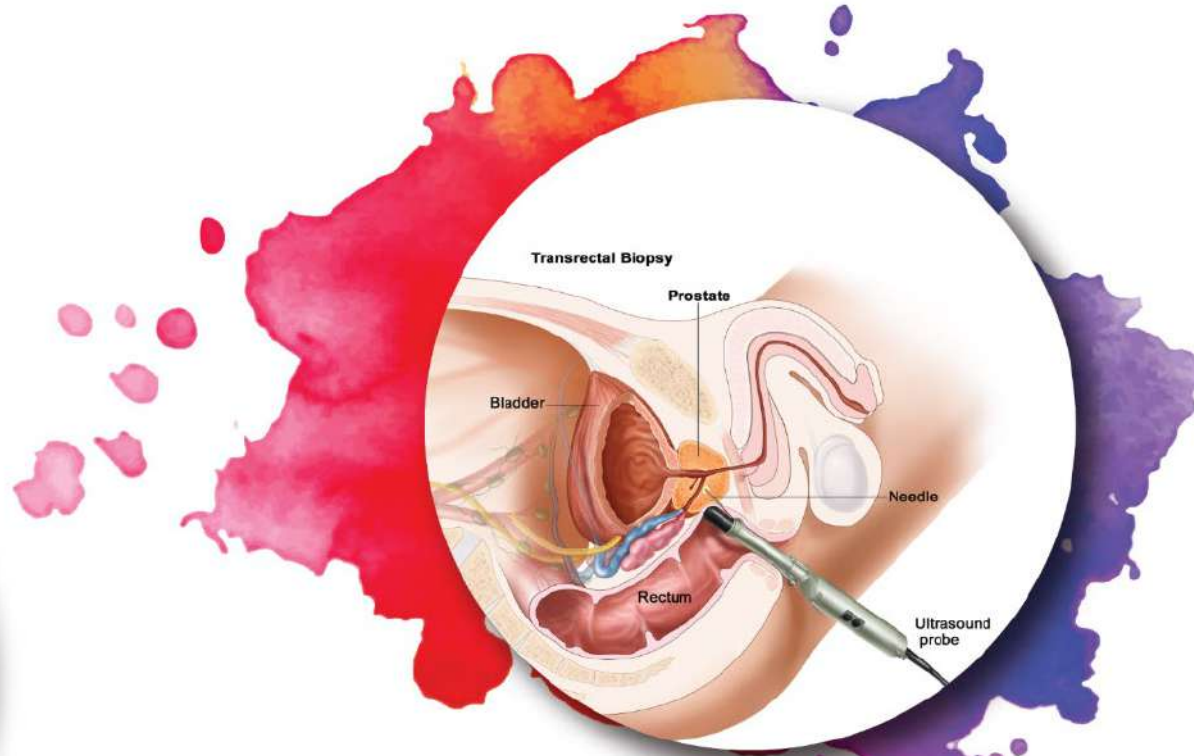


Screening

- PSA Blood test

30 million PSA tests every year in North America

80% of PSA tests are false positives.



Diagnosis

- Biopsy

12 needles through the prostate

- Pain, Discomfort, Infection
- Chance of life-threatening sepsis

Most are either negative or indolent prostate cancer

Current Dilemmas in Prostate Cancer Diagnosis

Over detection

- The positive predictive value of PSA (4-10 ng/mL) is low (25-40%)¹
- ~50% to 70% of newly diagnosed cancers are low risk;^{1,2} these are unlikely to benefit from early treatment

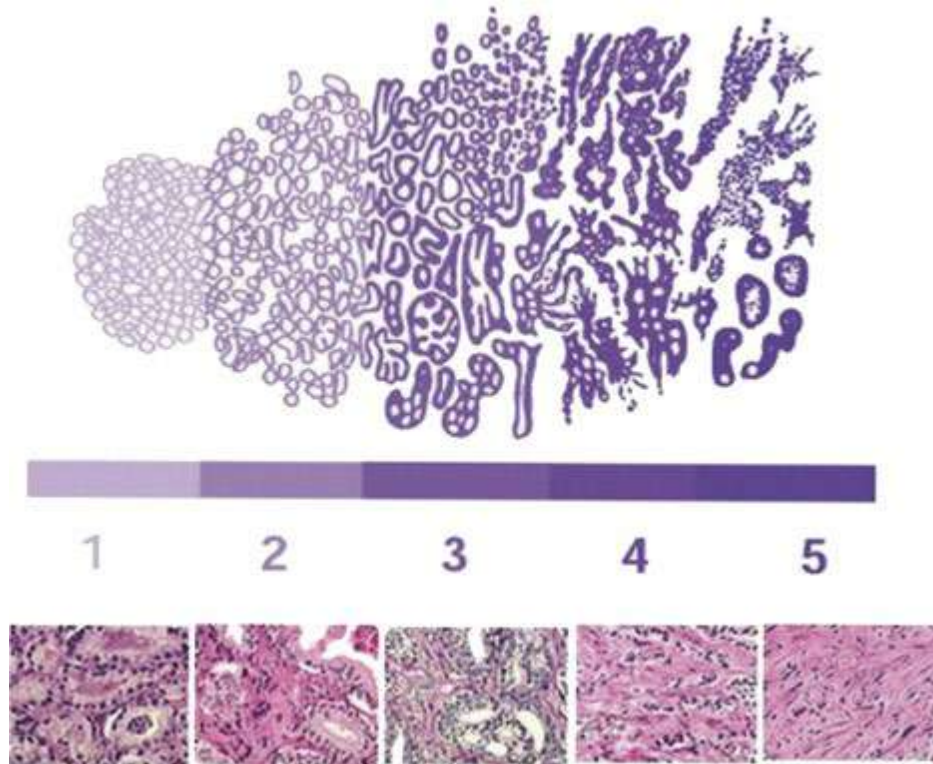
Unnecessary biopsies

- 65-75% of initial and 10-35% of repeat biopsies are negative^{1,3}
- Fear of missing significant cancer often results in a repeat biopsy, which is often also negative

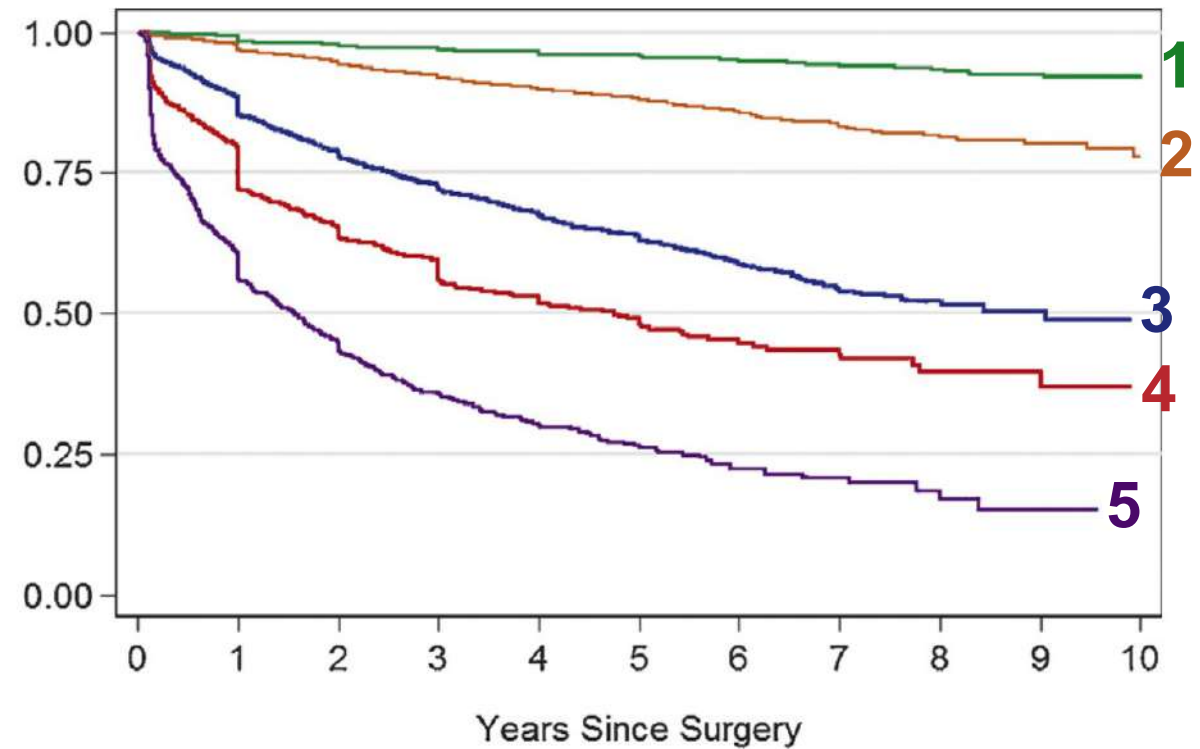
Burden of biopsy

- Prostate biopsy may be associated with pain, anxiety and complications⁴⁻⁶

Prostate cancer is a heterogeneous disease: some forms are lethal, others are not



Men with Gleason Grade Group 3-5 prostate cancers
have significantly worse outcomes



Significant need for an accurate, non-invasive test to detect patients that have aggressive prostate cancer
To determine if patients need a biopsy

ClarityDX: combining biomarkers and machine learning

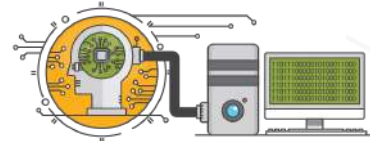
for the prediction of clinically significant PCa

Clinical data

- Age
- Digital rectal exam findings
- Previous negative biopsies



Machine Learning



Machine Learning

Biomarker data

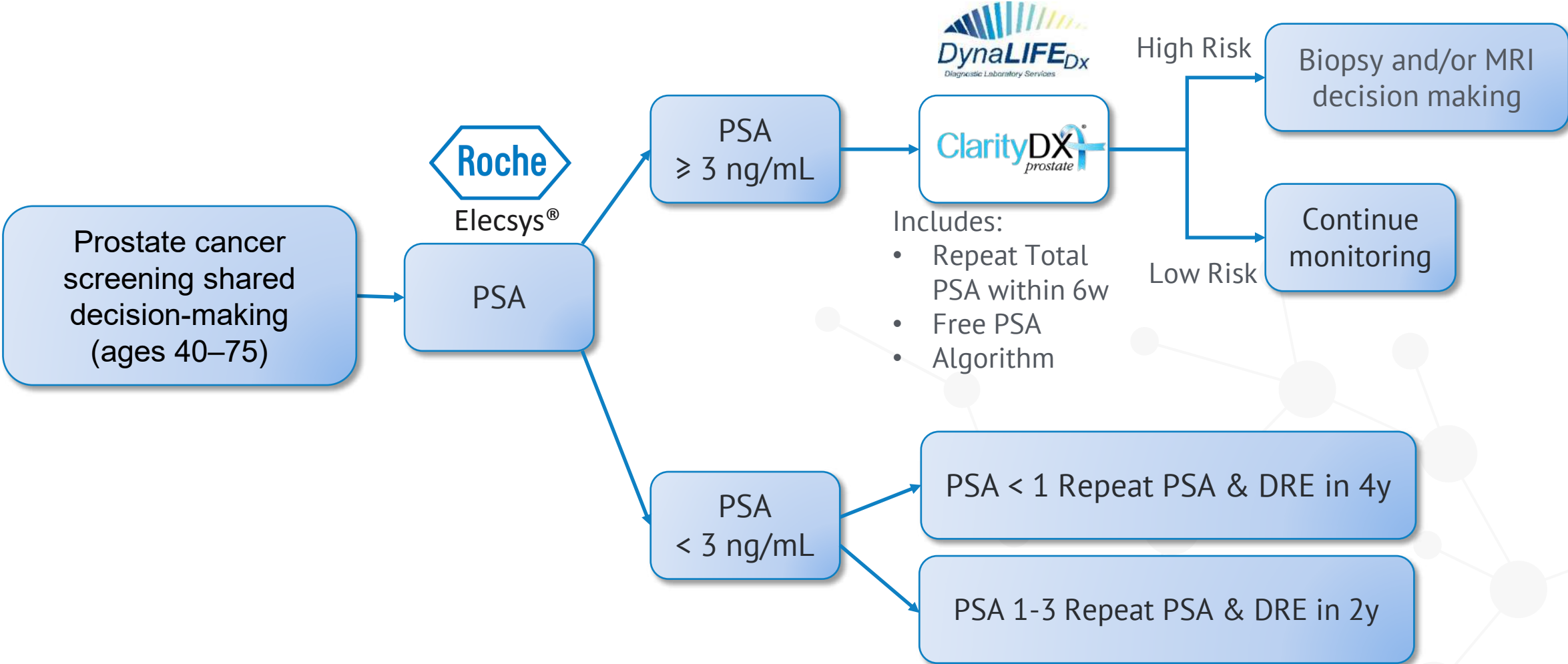
- Total PSA
- Free PSA

Risk score

Probability of having GG \geq 2 PCa



ClarityDX Prostate® as a reflex test for the PSA test



ClarityDX Prostate Clinical Study Protocol (APCaRI-05)

Male patient, 40-75yo, with abnormal PSA (≥ 3 ng/mL) & referred for prostate

$n= 1,584$

Identified by urologist / CRC - refer to CRC/research nurse

- Urology clinics
- Pre-biopsy (Diagnostic Radiology Clinics)

- 3 Sites Recruiting
 - Kipnes Urology Center
 - Prostate Cancer Center
 - Johns Hopkins University

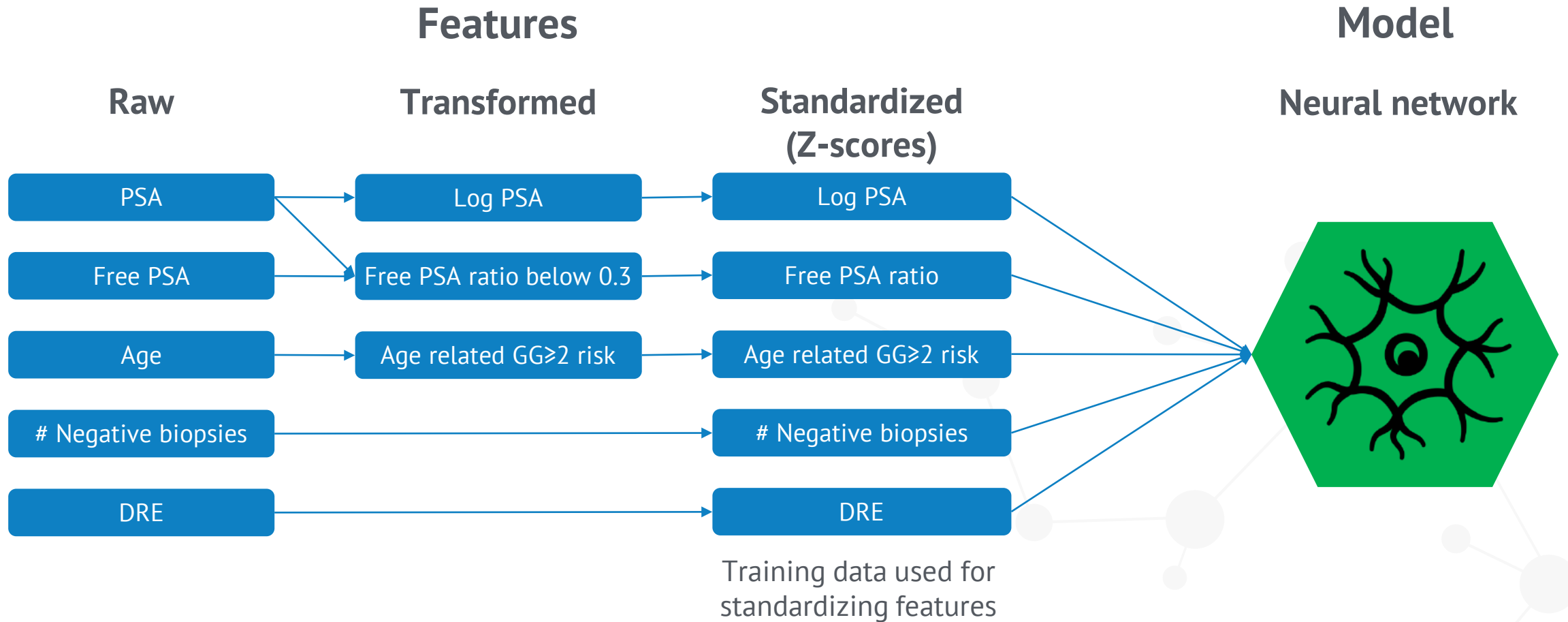
- Informed Consent
- Biospecimens are collected
- Intake Surveys (eligibility, medications)
- Demographic and clinical data



Biopsy is performed

Performance Characteristics

ClarityDX Prostate feature engineering and model

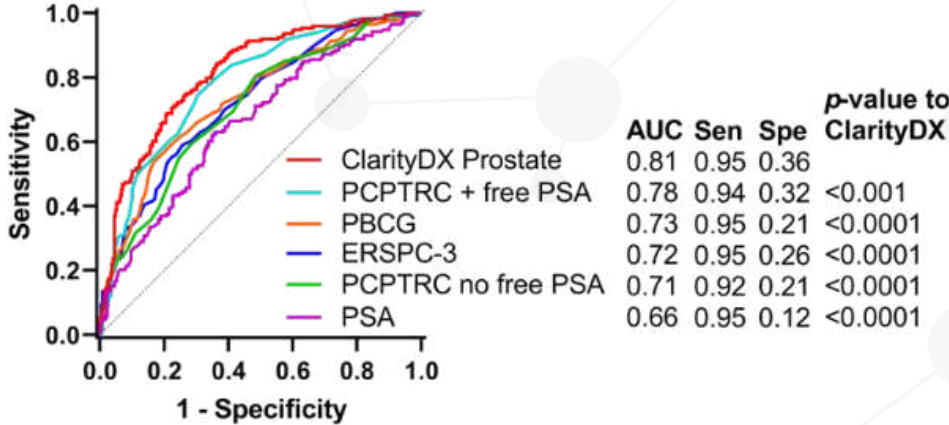


ClarityDX Prostate test has superior clinical performance

	GG \geq 2 PCa				
	AUC	Sensitivity	Specificity	NPV	PPV
ClarityDX Prostate	0.81	95	36	92	49
PCPTRC	0.71	92	21	81	43
PSA	0.66	95	12	78	41

- Model trained with data from 967 participants from KUC and JHU
- Model evaluated with data from 442 participants from PCC

Predicting grade group \geq 2 prostate cancer
Validation cohort (n = 442)



ClarityDX Prostate is the most accurate model for the prediction of clinically significant PCa

	GG ≤ 1 PCa	GG ≥ 2 PCa	p-value	ROC AUC (CI)	Cutoff	Sensitivity, % (CI)	Specificity, % (CI)	PPV, % (CI)	NPV, % (CI)
Patients, n	269 (61%)	173 (39%)							
Prior negative biopsy, n (%)	31 (12%)	3 (1.7%)	<0.0001	0.55 (0.43 - 0.47)		98 (95 - 99)	12 (8.1 - 16)	42 (37 - 47)	91 (76 - 97)
DRE, n (% abnormal)	62 (23%)	82 (47%)	<0.0001	0.63 (0.58 - 0.67)		52 (44 - 59)	74 (67 - 79)	57 (49 - 65)	69 (63 - 75)
Age, yr, median (IQR)	62 (57 - 67)	66 (60 - 69)	<0.0001	0.63 (0.57 - 0.68)	>51.50	95 (91 - 98)	4.8 (2.7 - 7.9)	39 (34 - 44)	59 (36 - 79)
PSA, ng/mL, median (IQR)	6.8 (5.2 - 8.9)	8.5 (6.5 - 12)	<0.0001	0.66 (0.61 - 0.71)	>4.540	95 (90 - 98)	12 (8.4 - 16)	41 (36 - 46)	78 (63 - 89)
PCPTRC, median (IQR)	9.0 (7.0 - 13)	14 (10 - 20)	<0.0001	0.71 (0.66 - 0.75)	>6.500	92 (88 - 96)	21 (17 - 26)	43 (38 - 48)	81 (71 - 89)
ClarityDX Prostate, median (IQR)	24 (15 - 44)	60 (42 - 78)	<0.0001	0.81 (0.77 - 0.85)	>17.80	95 (91 - 98)	36 (31 - 42)	49 (44 - 54)	92 (86 - 96)

ClarityDX Prostate can avoid **36%** of unnecessary biopsies

Threshold	GG ≥2 PCa found (%)	GG ≥2 PCa missed (%)	Biopsies avoided (%)	Unnecessary biopsies avoided (%)
17.8	95	5	24	36

ClarityDX Prostate can save up to \$1.5M per year in Alberta alone

Costs	Avg
Biopsy	\$1,148
MRI	\$733
Complications	\$118
Total	\$2,000
Savings Per Elevated PSA	\$540
Cost of test	\$100
Cost savings per elevated PSA	\$440

Building a real-world data asset enabling AI model training and validation

Individuals



9,000+

Longitudinal patient data:
Prostate cancer records



1,500

Prostate cancer records



2M

Alberta Cancer Registry: access
to all cancer data from Alberta



330,000+

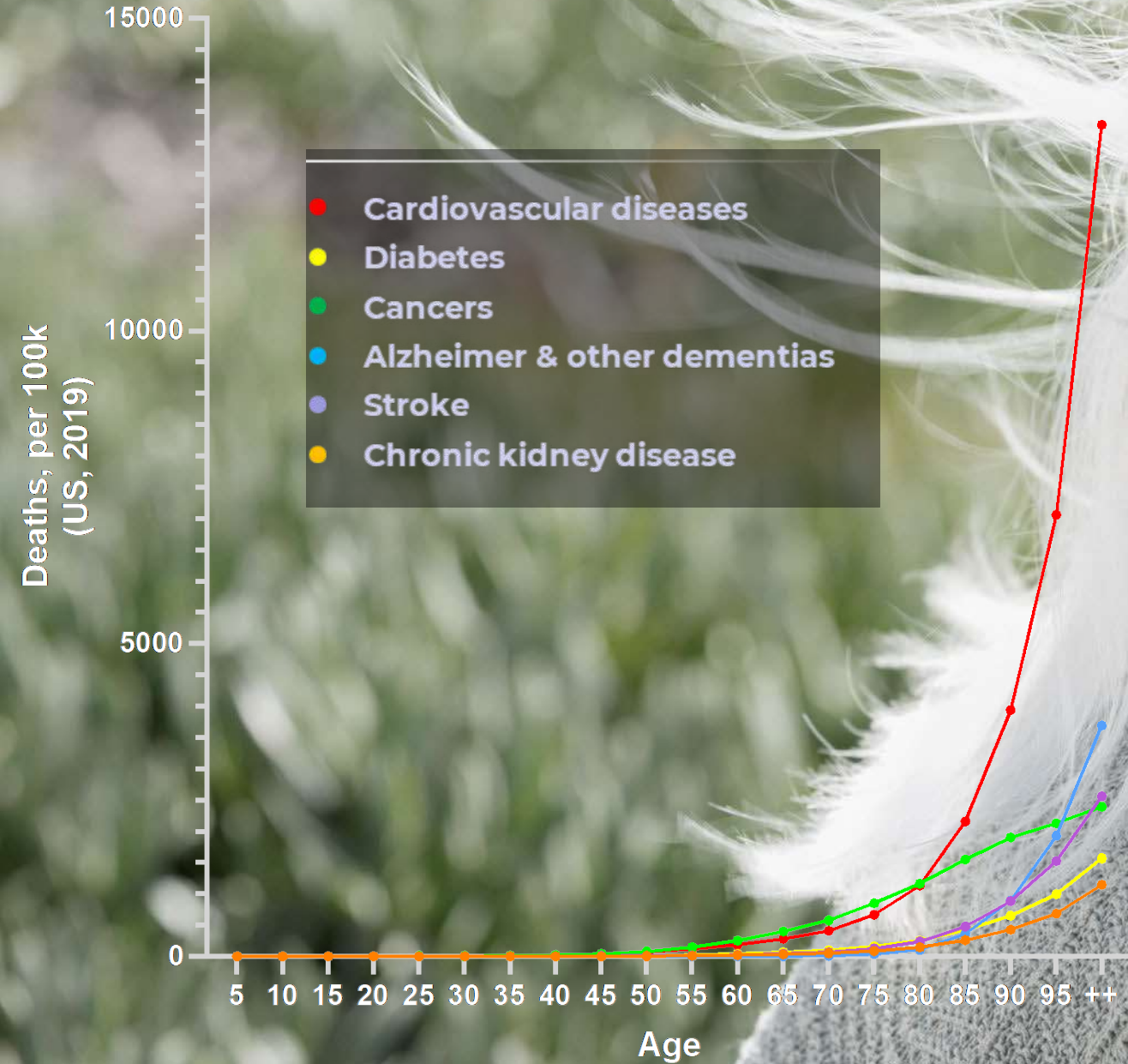
Longitudinal patient data:
Canadians aged 30-74



40,000+

Longitudinal patient data:
men with prostate cancer
globally

What's next?



Acknowledgements

Team

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Study

APCaRI 01&03

APCaRI 05

MAST

APCaRI 01&03

APCaRI 05

IRONMAN

TrueNTH-PCO

TrueNTH-PCO

IRONMAN

APCaRI 01&03

GURC

APCaRI 01&03

IRONMAN

TrueNTH-PCO

APCaRI 01&03

APCaRI 05



MAST

APCaRI 01&03




PROSPeCT



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