

AI' Role in Novel Target Discovery for High Unmet Needs Through Multi-Omics Integration

Namshik Han



UNIVERSITY OF
CAMBRIDGE



YONSEI UNIVERSITY



CardiaTec^{BIO}



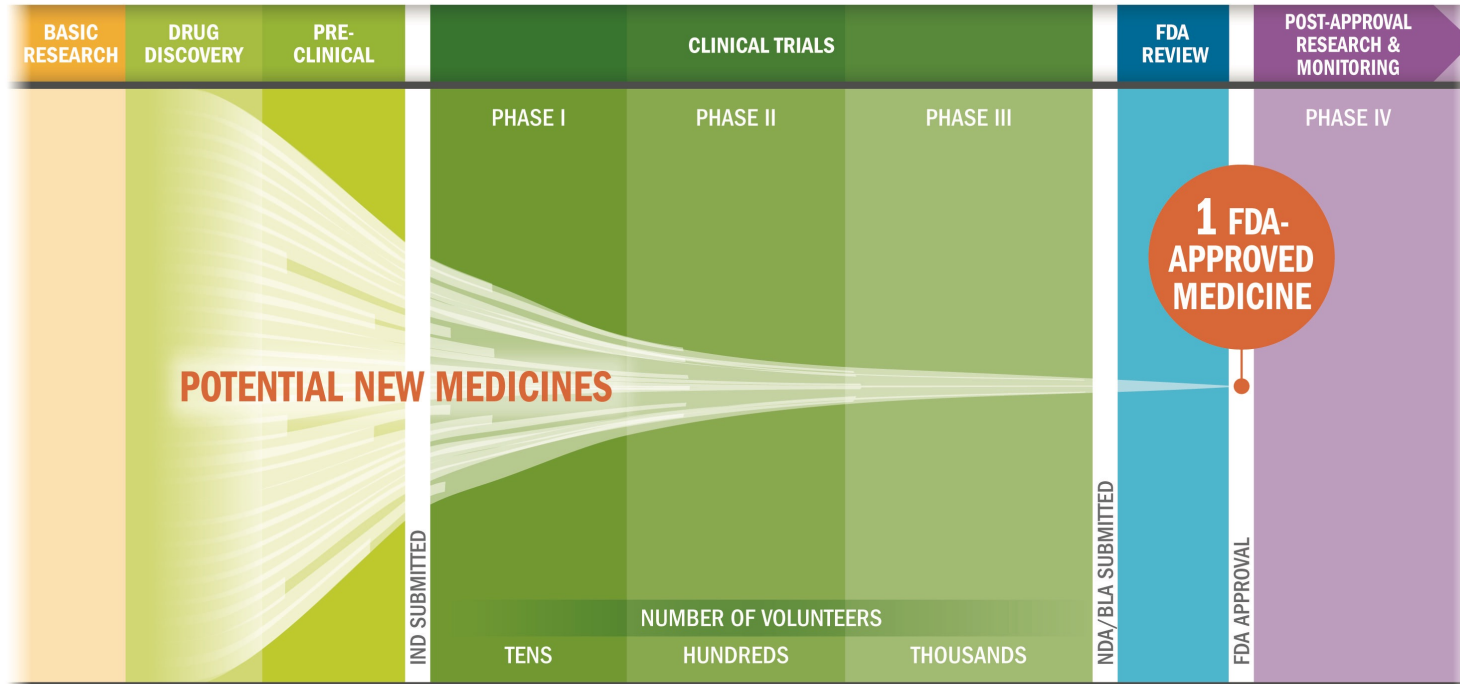
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Challenges in Drug Discovery

THE BIOPHARMACEUTICAL RESEARCH AND DEVELOPMENT PROCESS

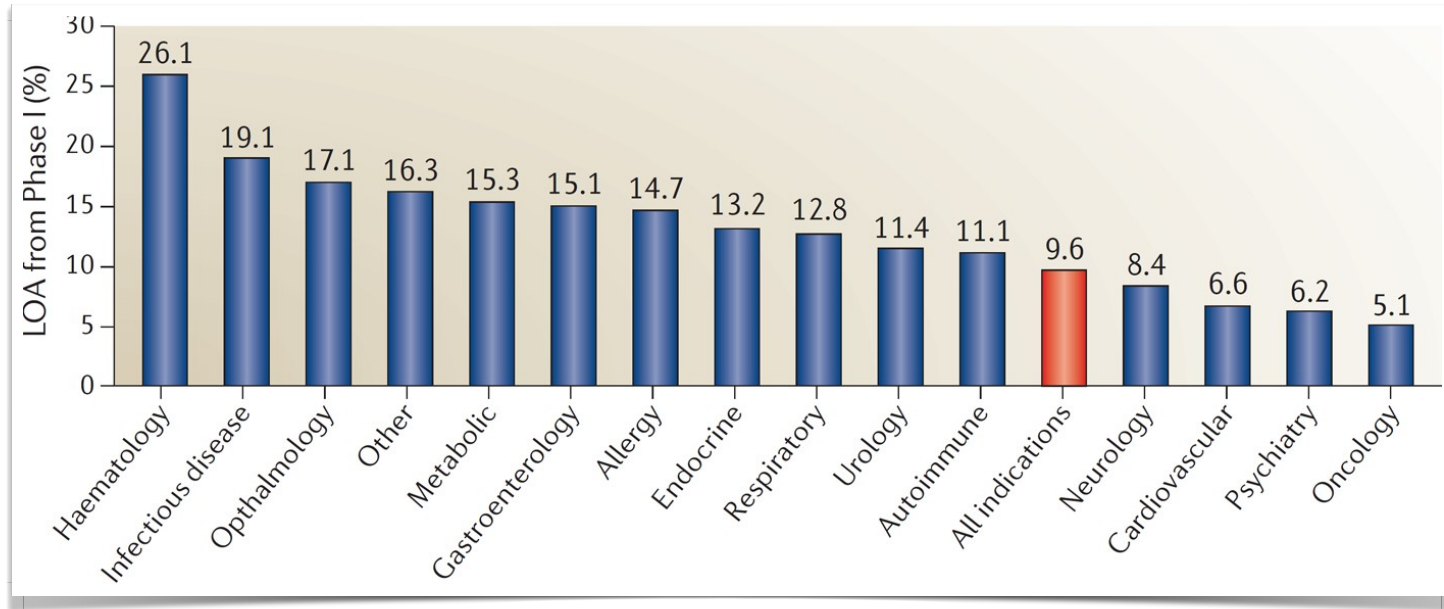
From drug discovery through FDA approval, developing a new medicine takes at least 10 years on average and costs an average of \$2.6 billion.* Less than 12% of the candidate medicines that make it into Phase I clinical trials will be approved by the FDA.



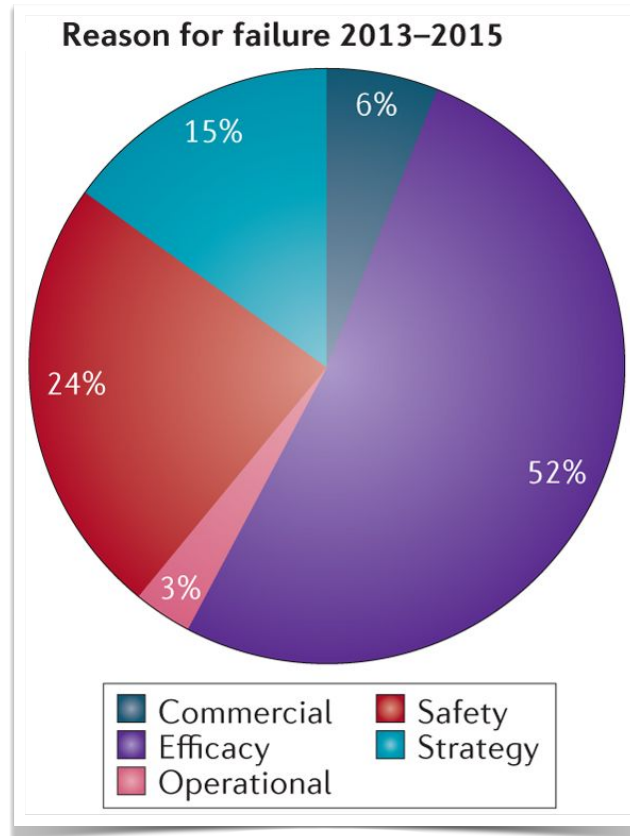
Key: IND: Investigational New Drug Application, NDA: New Drug Application, BLA: Biologics License Application

Highest Attrition Rate

Likelihood of approval (LOA) from Phase I

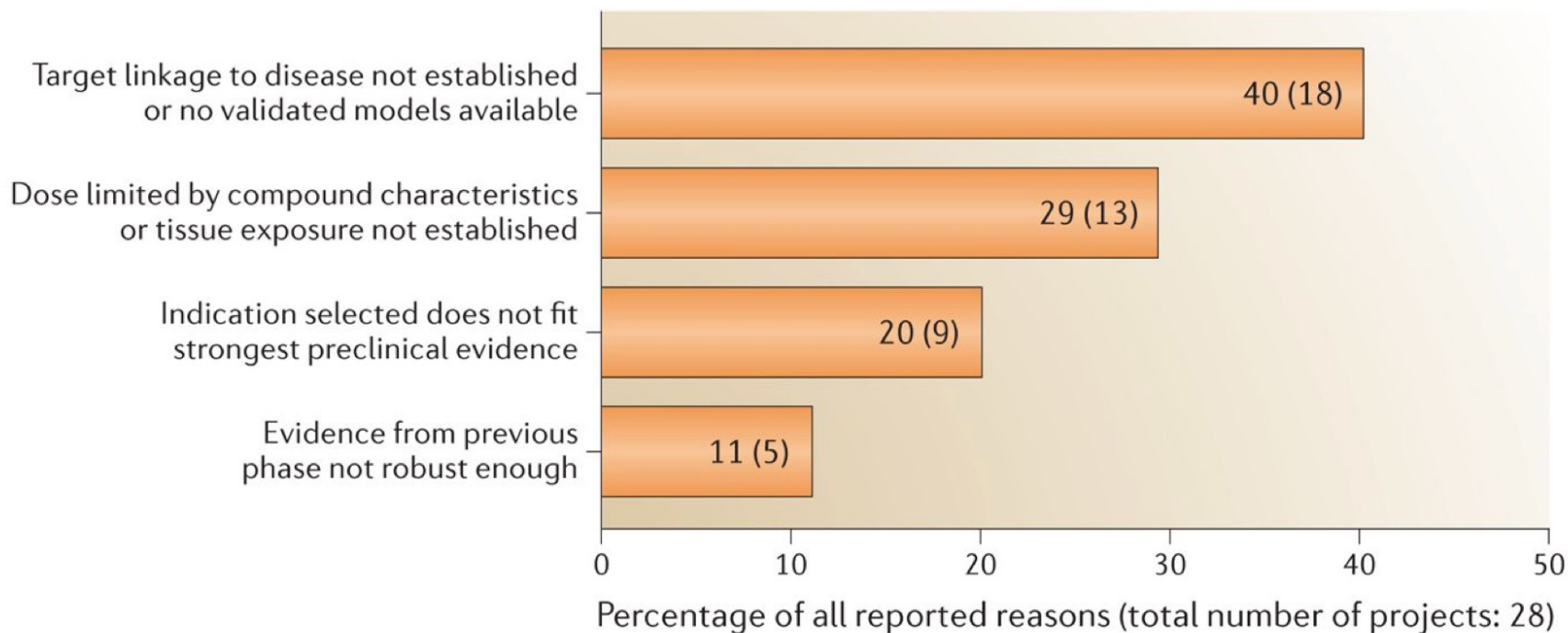


Scientific Reasons



Target-Disease Association

Reasons for lack of clinical efficacy



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Better drugs, faster: The potential of AI-powered humans

By Emma Woollacott
Technology of Business reporter

1 August 2017 | Business

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BBC News, 2017

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AI-powered drug discovery captures pharma interest

Eric Smalley

Nature Biotechnology **35**, 604–605 (2017) | doi:10.1038/nbt0717-604
Published online 12 July 2017

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Nature Biotechnology, 2017

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By Jane Wakefield
Technology reporter

🕒 30 January 2020



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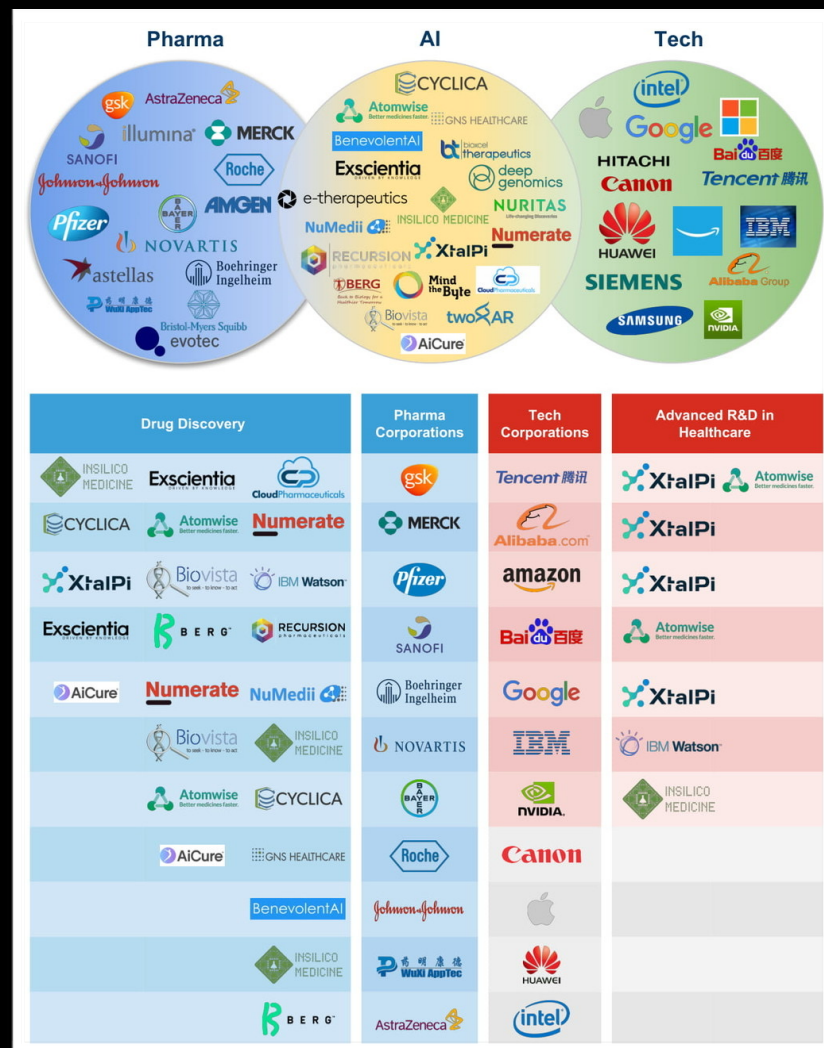
The drug was much quicker to market than ones developed in more traditional ways

A drug molecule "invented" by artificial intelligence (AI) will be used in human trials in a world first for machine learning in medicine.

It was created by British start-up Exscientia and Japanese pharmaceutical firm Sumitomo Dainippon Pharma.

The drug will be used to treat patients who have obsessive-compulsive disorder (OCD).

Typically, drug development takes about five years to get to trial, but the AI drug took just 12 months.





BCG

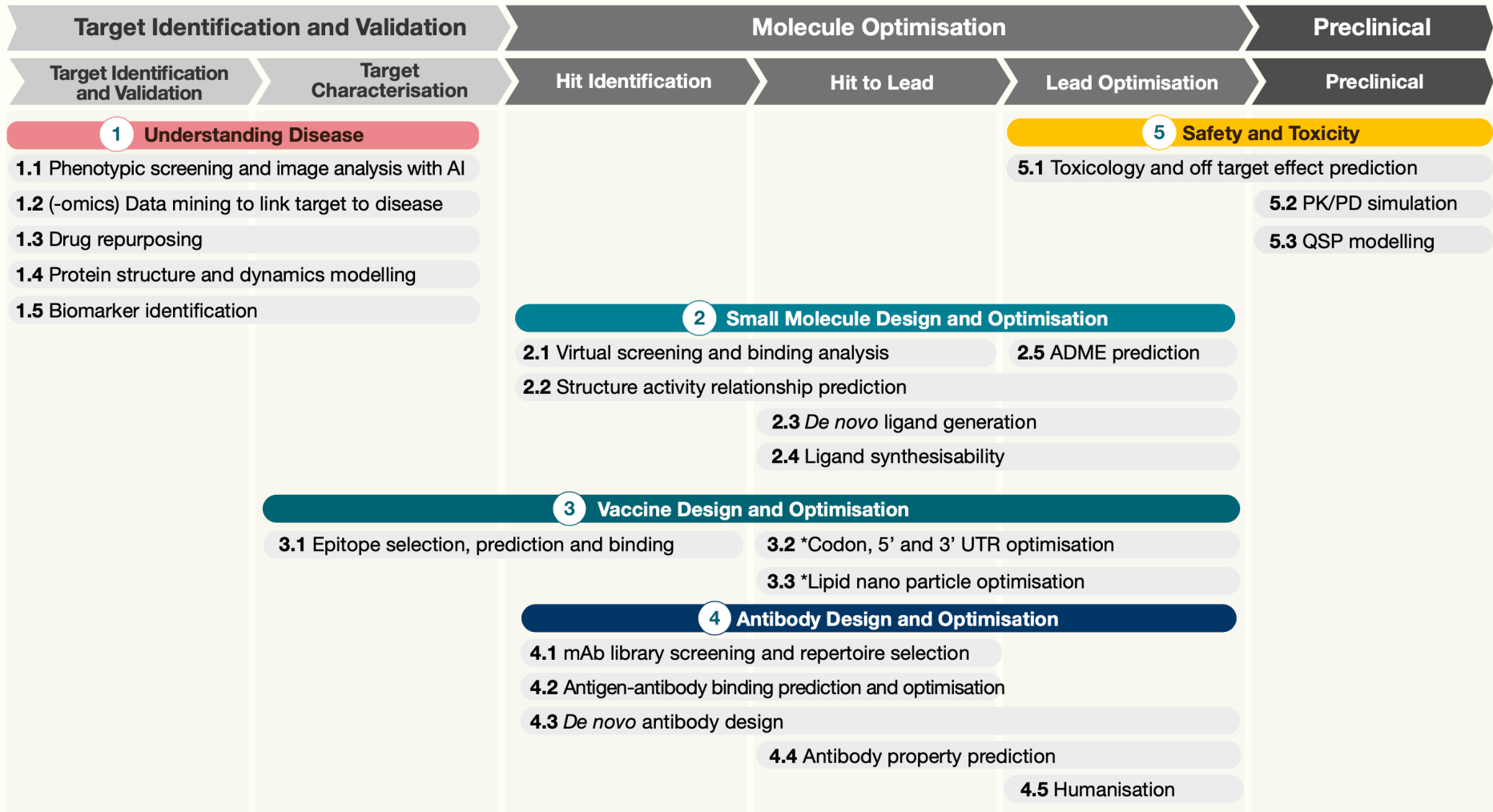
Unlocking the potential of AI in Drug Discovery

Current status, barriers and
future opportunities

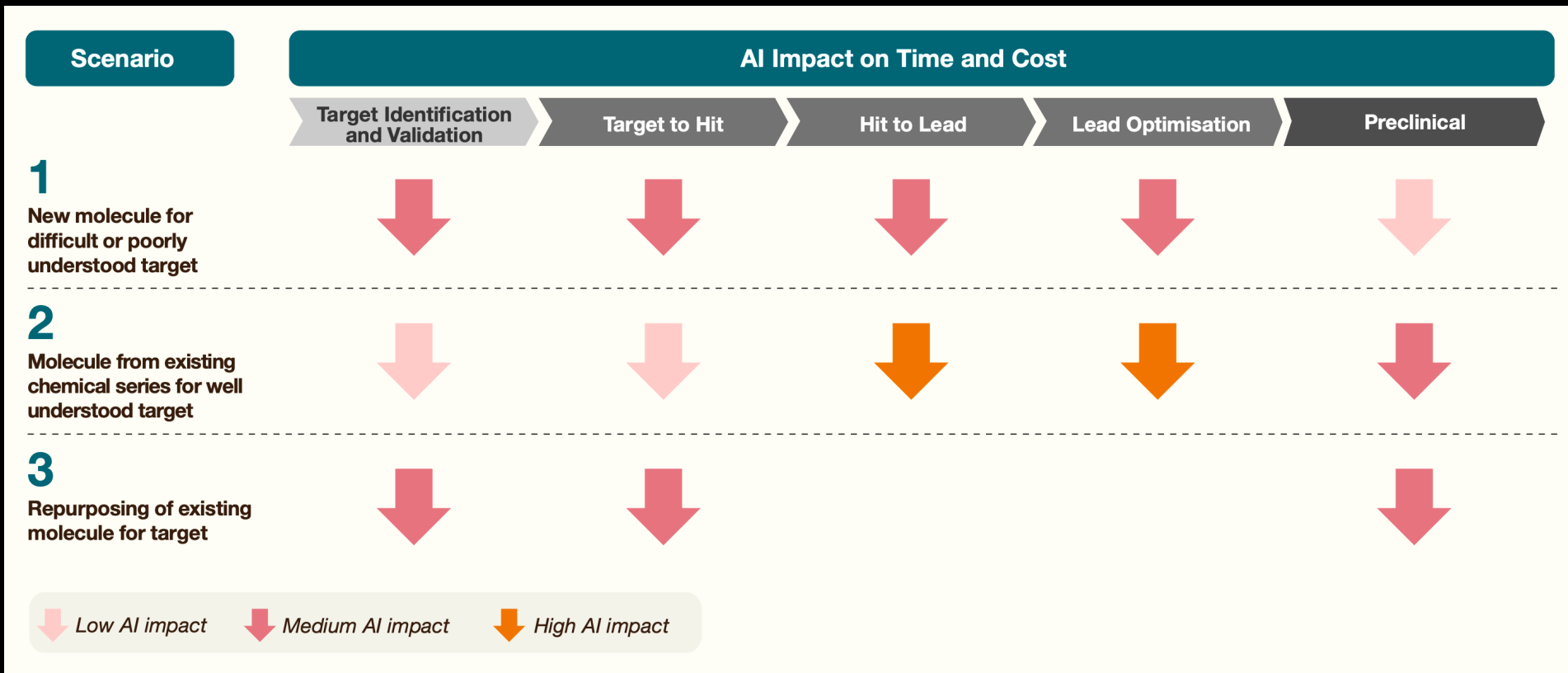


Key Pain Points

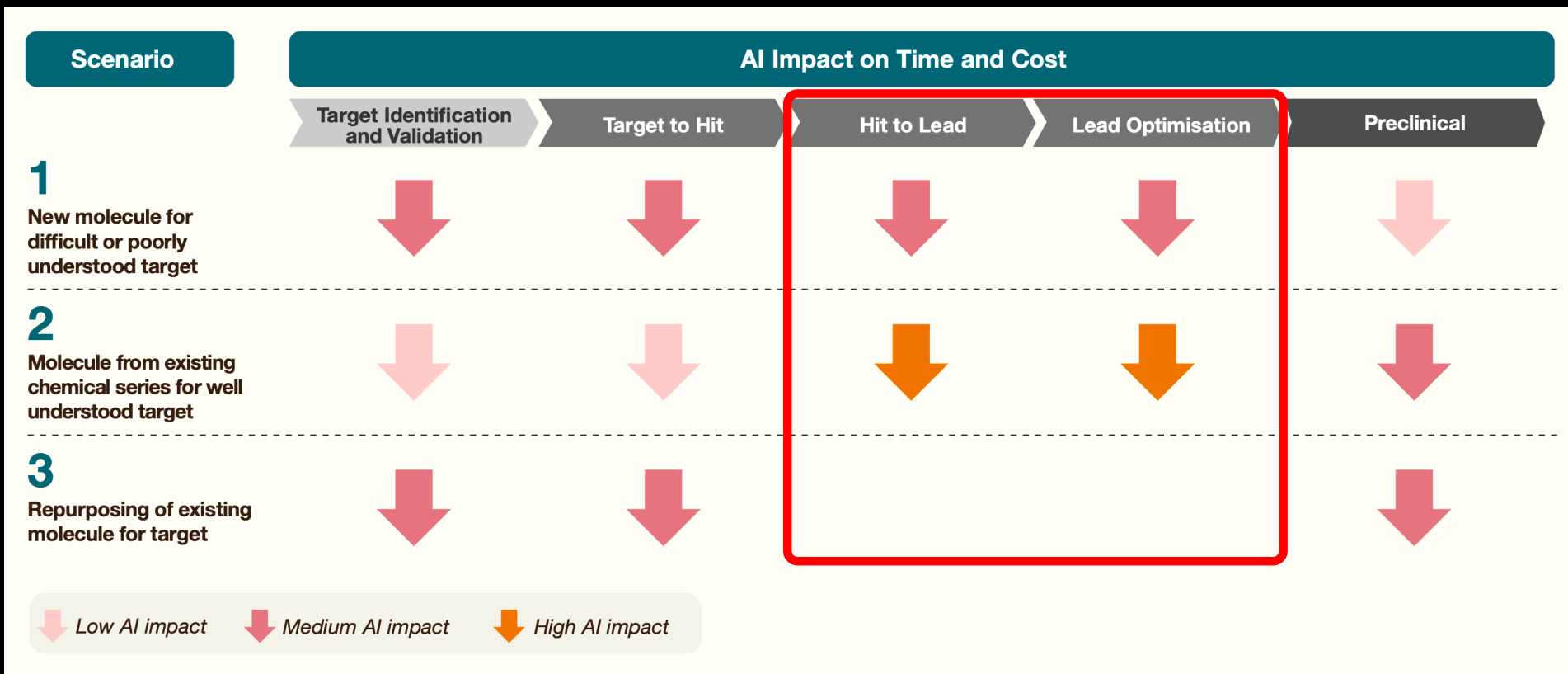
Typical Pain Points	Target Identification and Validation		Molecule Optimisation			Preclinical
	Target Identification and Validation	Target Characterisation	Hit Identification	Hit to Lead	Lead Optimisation	Preclinical
Modality and Discovery Strategy Agnostic Pain Points	Complex deconvolution of experimental and clinical signals to identify credible targets		Low translatability of experimental assays to <i>in vivo</i> systems		Lengthy optimisation cycles across numerous properties	Low translatability of animal models
	Fragmented, multi-modal data sets inconsistently collected and harmonised					
	Challenging to fully characterise proteins/protein-protein interactions					
Small Molecules Pain Points			Long design-make-test cycles across discovery journeys		Lack of access to ADMET data	
			Unable to fully explore chemical space			
Vaccines Pain Points		Challenging to identify optimal immunogenic antigen sequences			Sub-optimal mRNA translation efficiency	
		Time consuming to identify conserved regions manually			Lack of understanding of delivery systems	
Antibodies Pain Points	Difficult to identify relevant/superior epitopes		Challenging to select correct candidate across experimental outputs/human repetoires		Lengthy humanisation of unfamiliar scaffolds	
	Challenging to design and optimise novel antibodies				Long lead times to improve mAb preclinical properties	



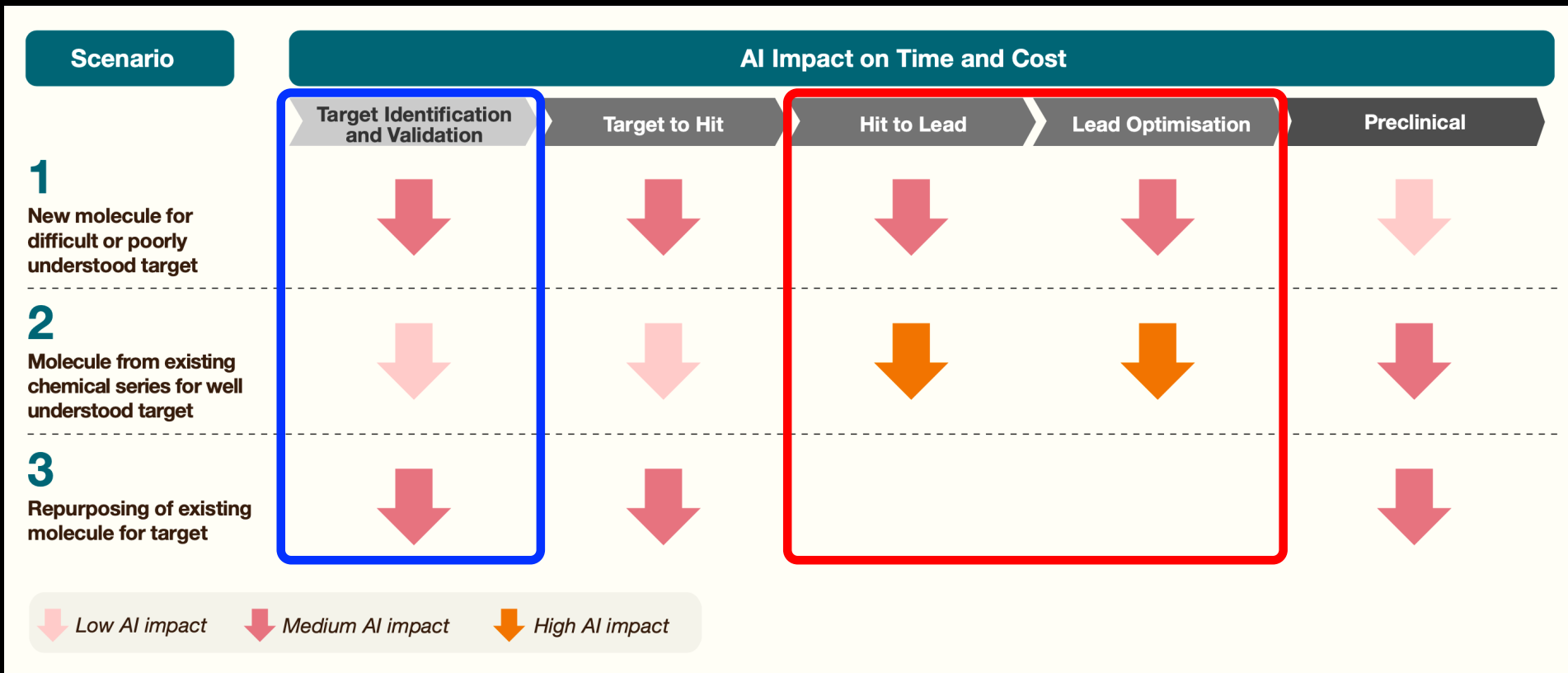
AI Impact on Time & Cost



AI Impact on Time & Cost



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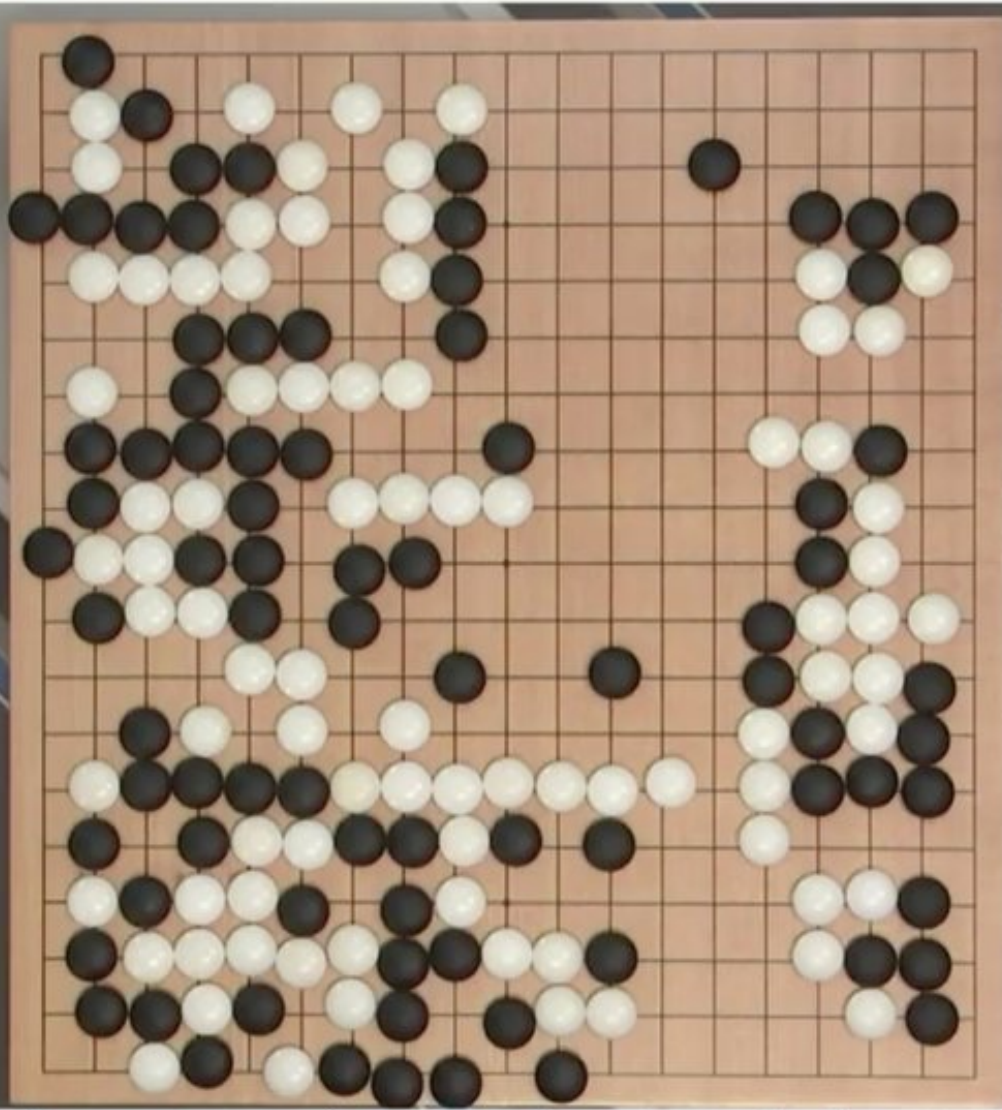
AI Impact on **Unmet Medical Needs**



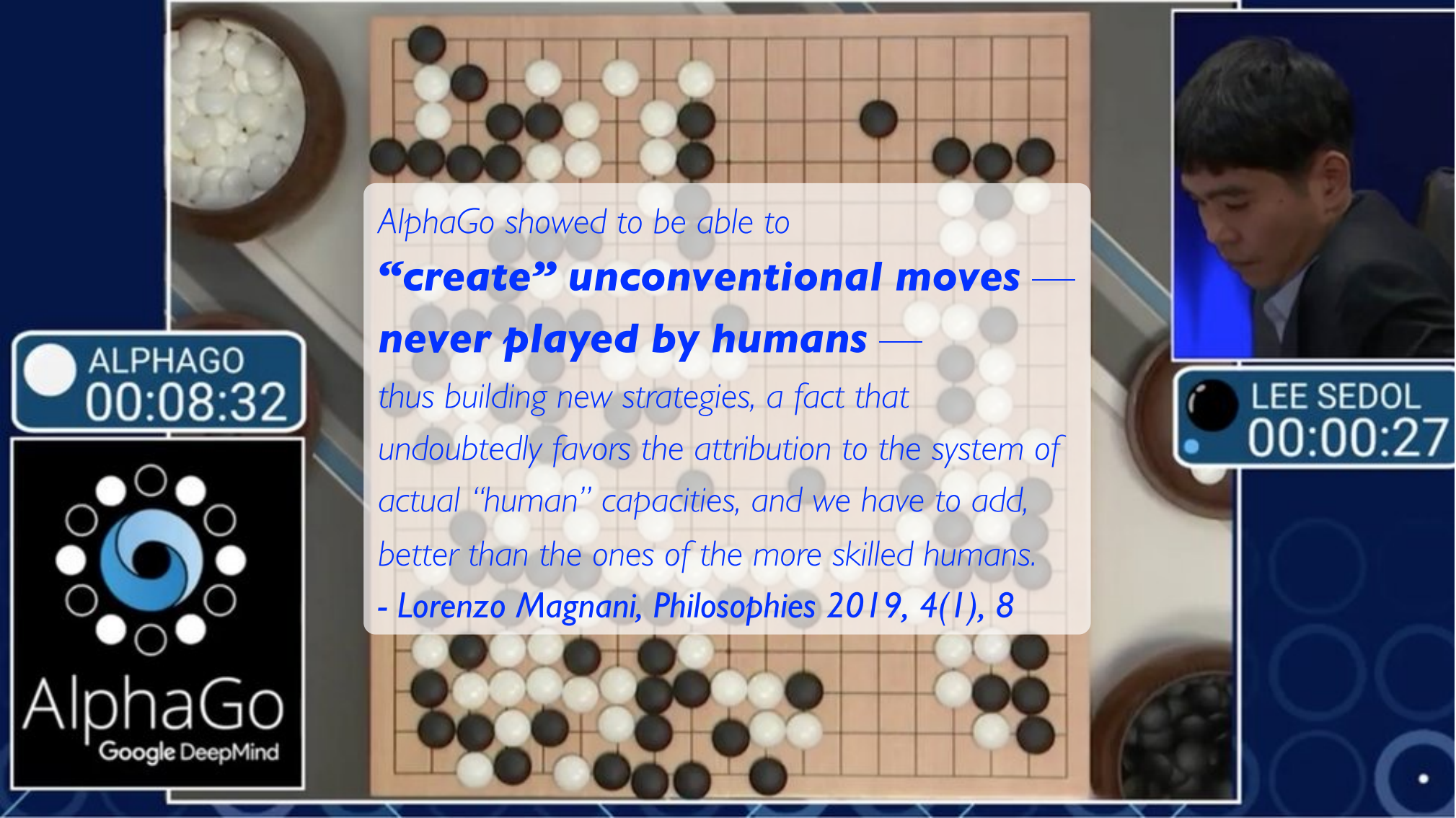
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AlphaGo
Google DeepMind



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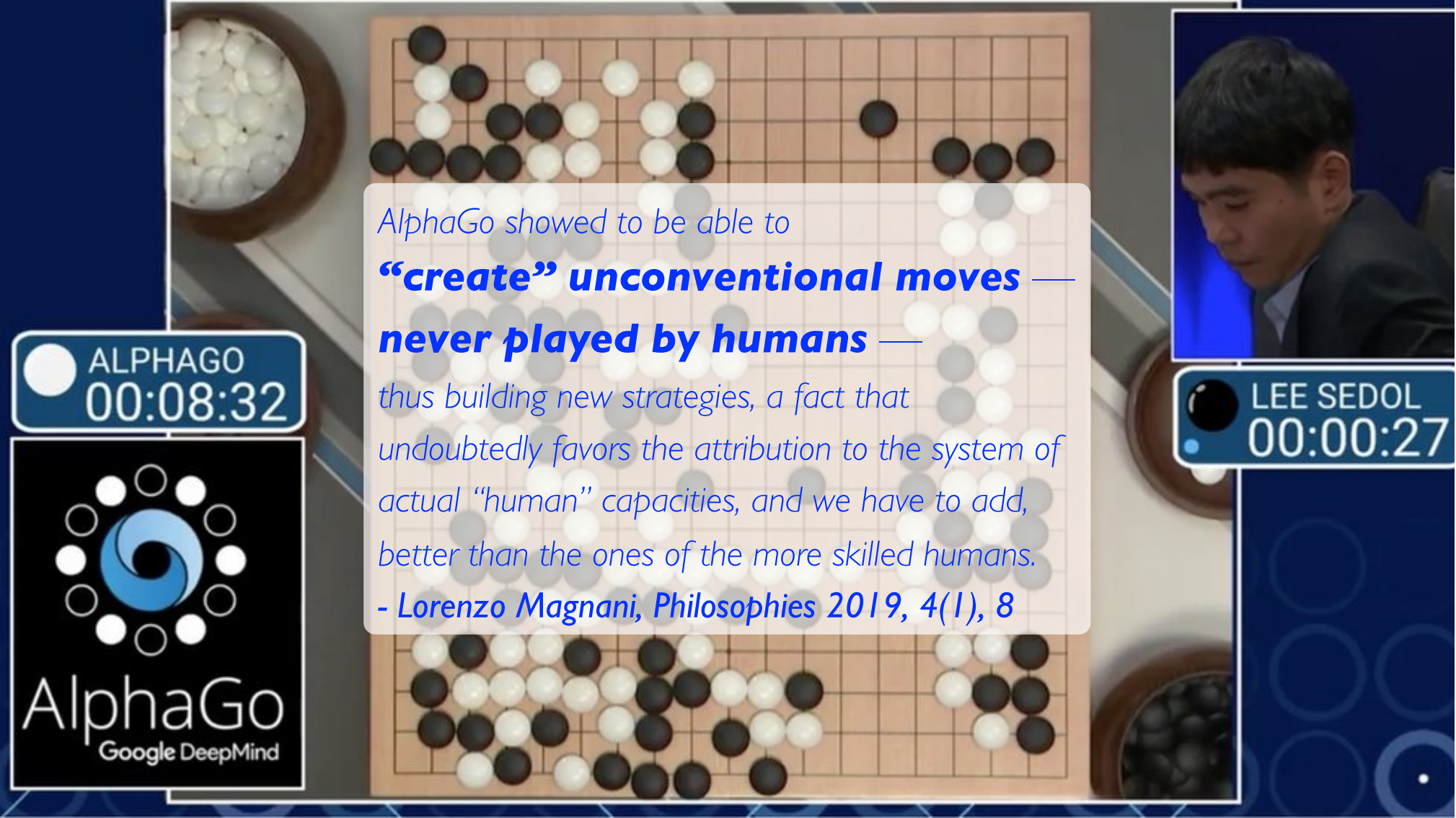
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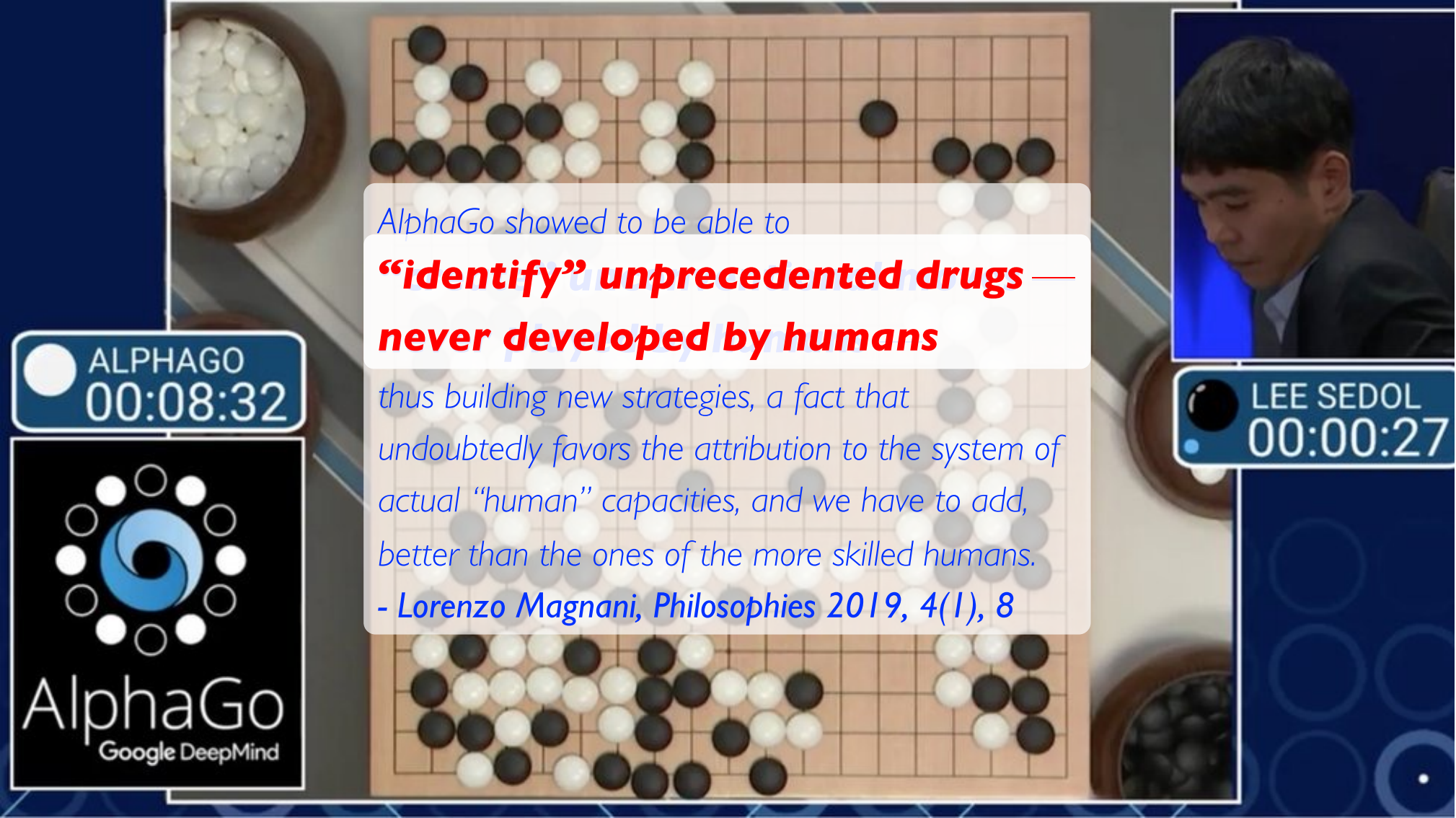
AlphaGo showed to be able to
**“create” unconventional moves —
never played by humans —**

*thus building new strategies, a fact that
undoubtedly favors the attribution to the system of
actual “human” capacities, and we have to add,
better than the ones of the more skilled humans.*

- Lorenzo Magnani, Philosophies 2019, 4(1), 8



LEE SEDOL
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ALPHAGO
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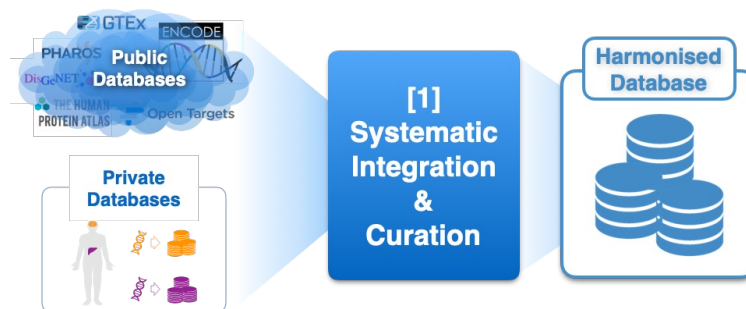
**“identify” unprecedented drugs —
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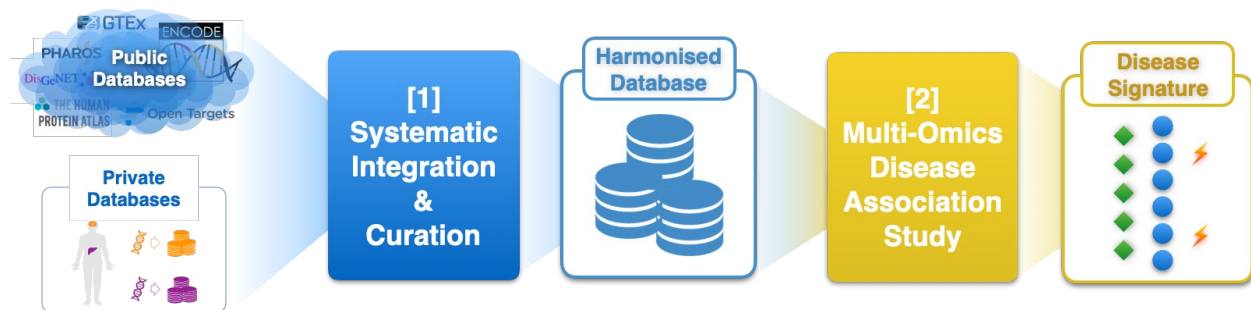
- Lorenzo Magnani, Philosophies 2019, 4(1), 8

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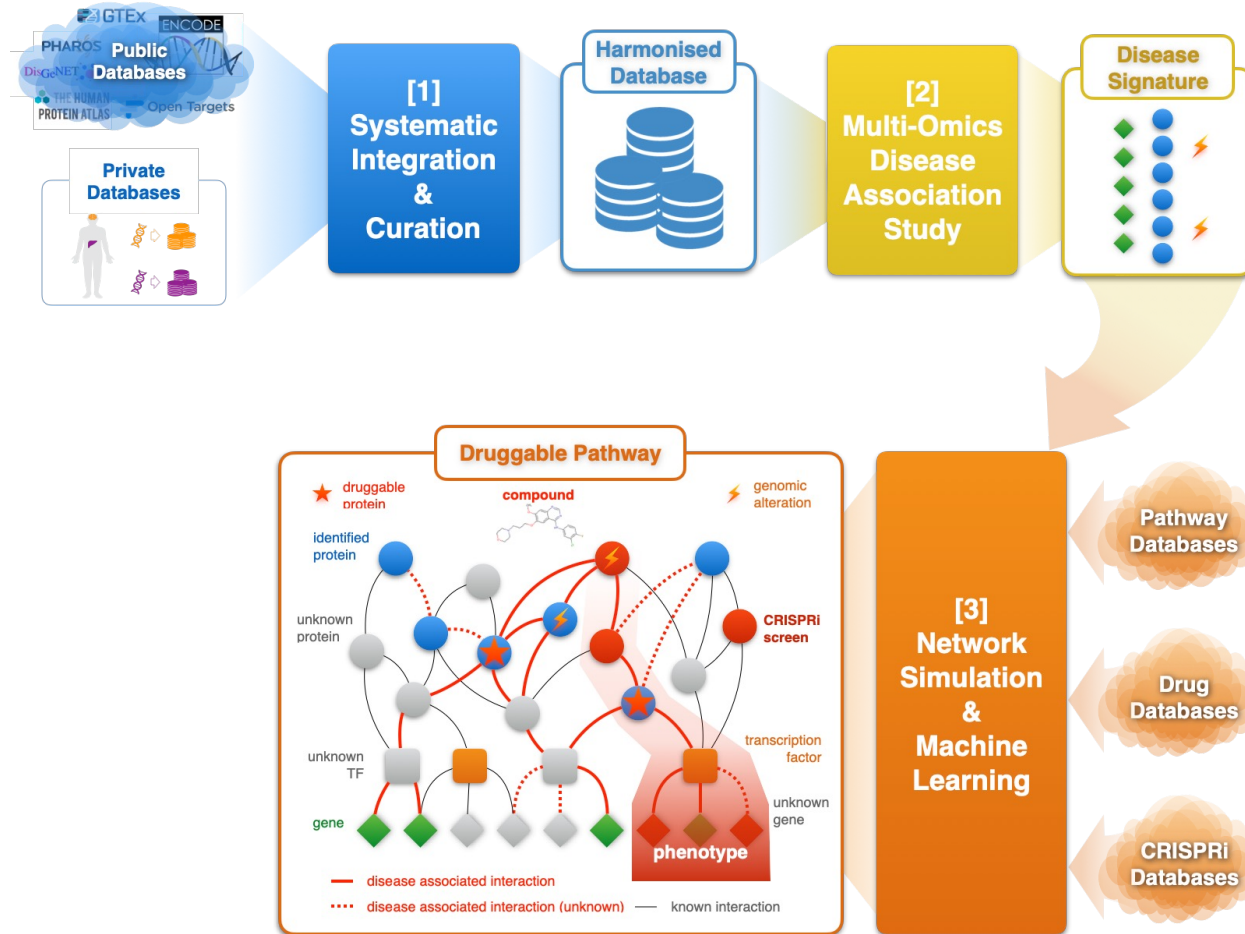
in silico Drug Discovery Pipeline



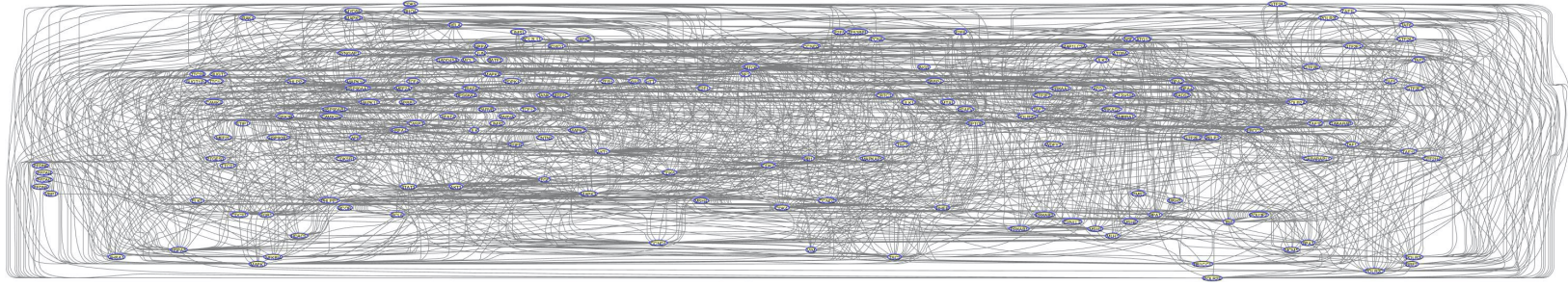
in silico Drug Discovery Pipeline



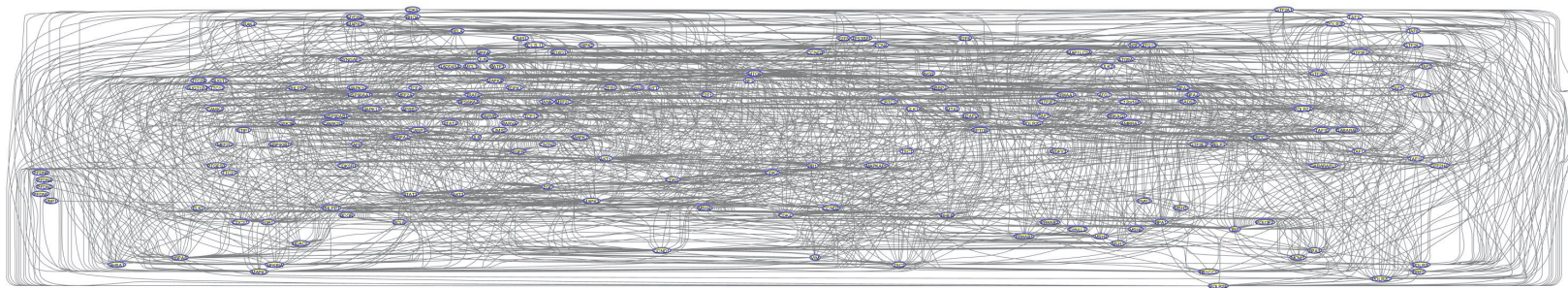
in silico Drug Discovery Pipeline



Network Analysis



Network Analysis



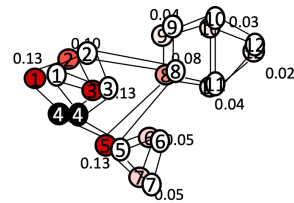
$$\vec{r}_i = c\tilde{W}\vec{r}_i + (1-c)\vec{e}_i$$

Ranking vector

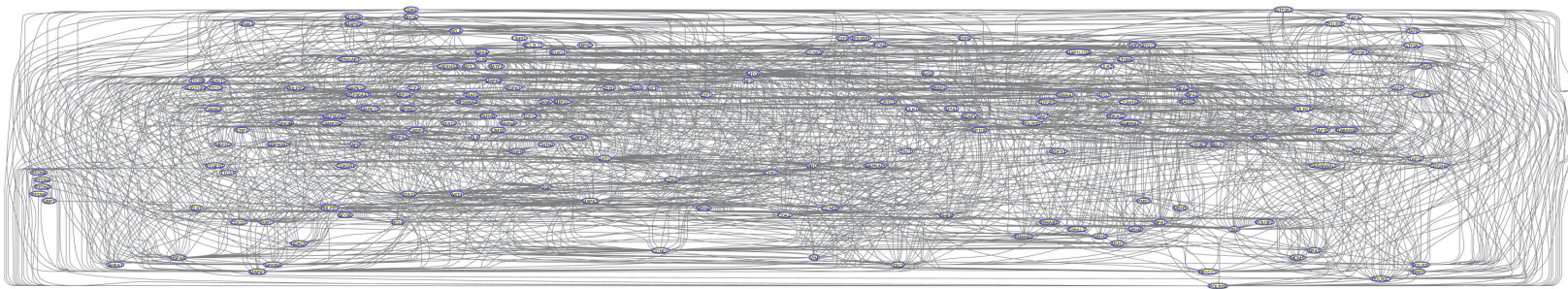
Adjacent matrix

starting vector

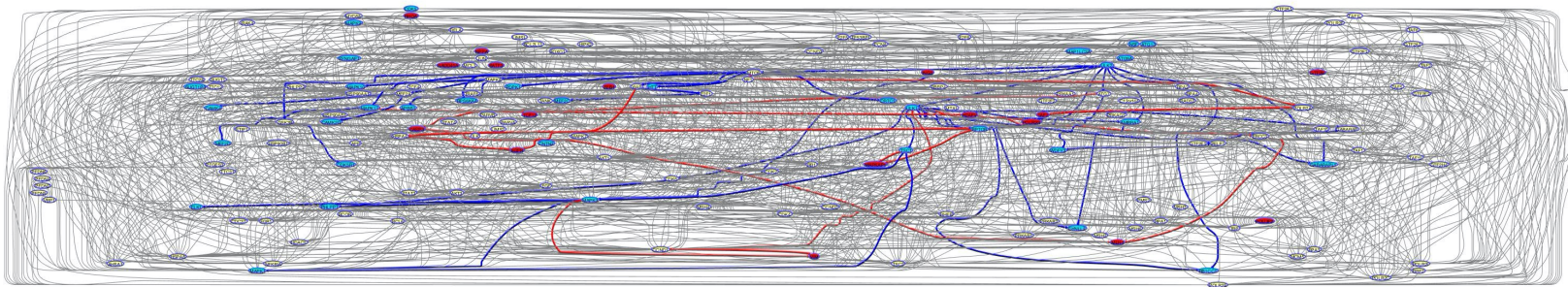
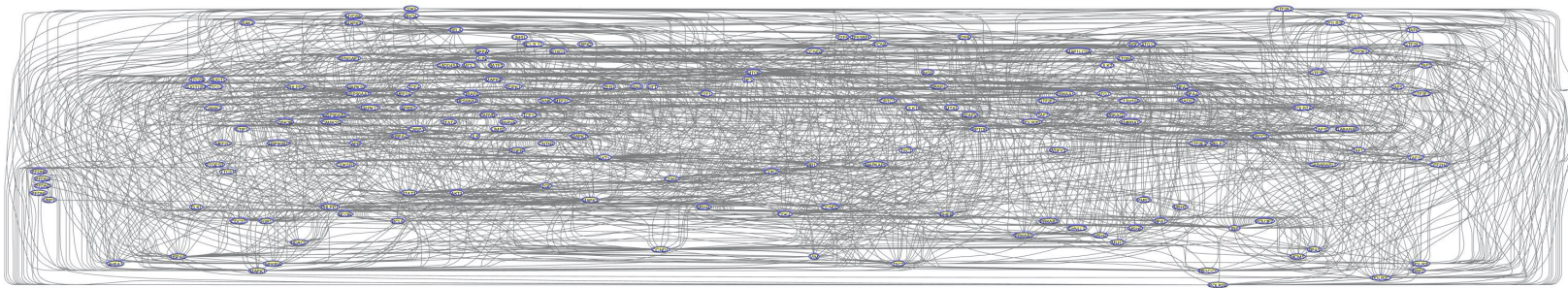
$\begin{pmatrix} 0.13 \\ 0.10 \\ 0.13 \\ 0.13 \\ 0.22 \\ 0.13 \\ 0.05 \\ 0.05 \\ 0.08 \\ 0.04 \\ 0.03 \\ 0.04 \\ 0.02 \end{pmatrix}$ $n \times 1$	$= 0.9 \times$	$\begin{pmatrix} 0 & 1/3 & 1/3 & 1/3 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 1/3 & 0 & 0 & 0 & 0 & 1/4 & 0 & 0 & 0 \\ 1/3 & 1/3 & 0 & 1/3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 1/3 & 0 & 1/3 & 0 & 1/4 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 1/3 & 0 & 1/2 & 1/2 & 1/4 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1/4 & 0 & 1/2 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1/4 & 1/2 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1/3 & 0 & 0 & 1/4 & 0 & 0 & 1/2 & 0 & 1/3 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1/4 & 0 & 1/3 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1/2 & 0 & 1/3 & 1/2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1/4 & 0 & 1/3 & 0 & 1/2 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1/3 & 1/3 & 0 \end{pmatrix}$ $n \times n$	$\begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{pmatrix}$ $n \times 1$
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Network Analysis



Network Analysis

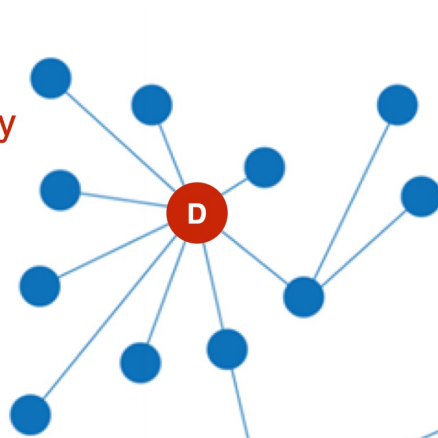


Network Algorithms

Degree Centrality

Number of connections?

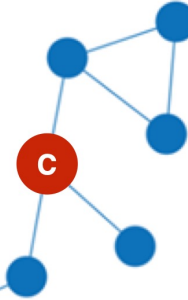
"A" has a high degree



Closeness Centrality

Which node can most easily reach all other nodes in a graph or subgraph?

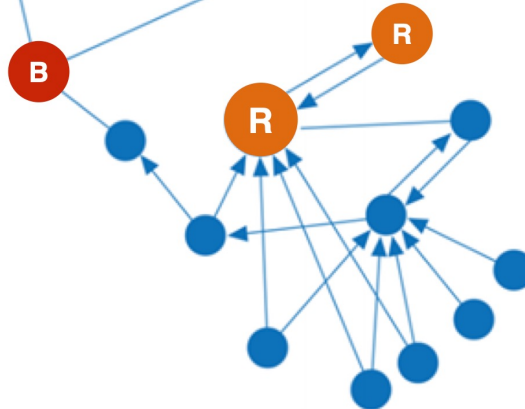
"B" is closest with the fewest hops in its subgraph



Betweenness Centrality

Which node has the most control over flow between nodes and groups?

"C" is a bridge



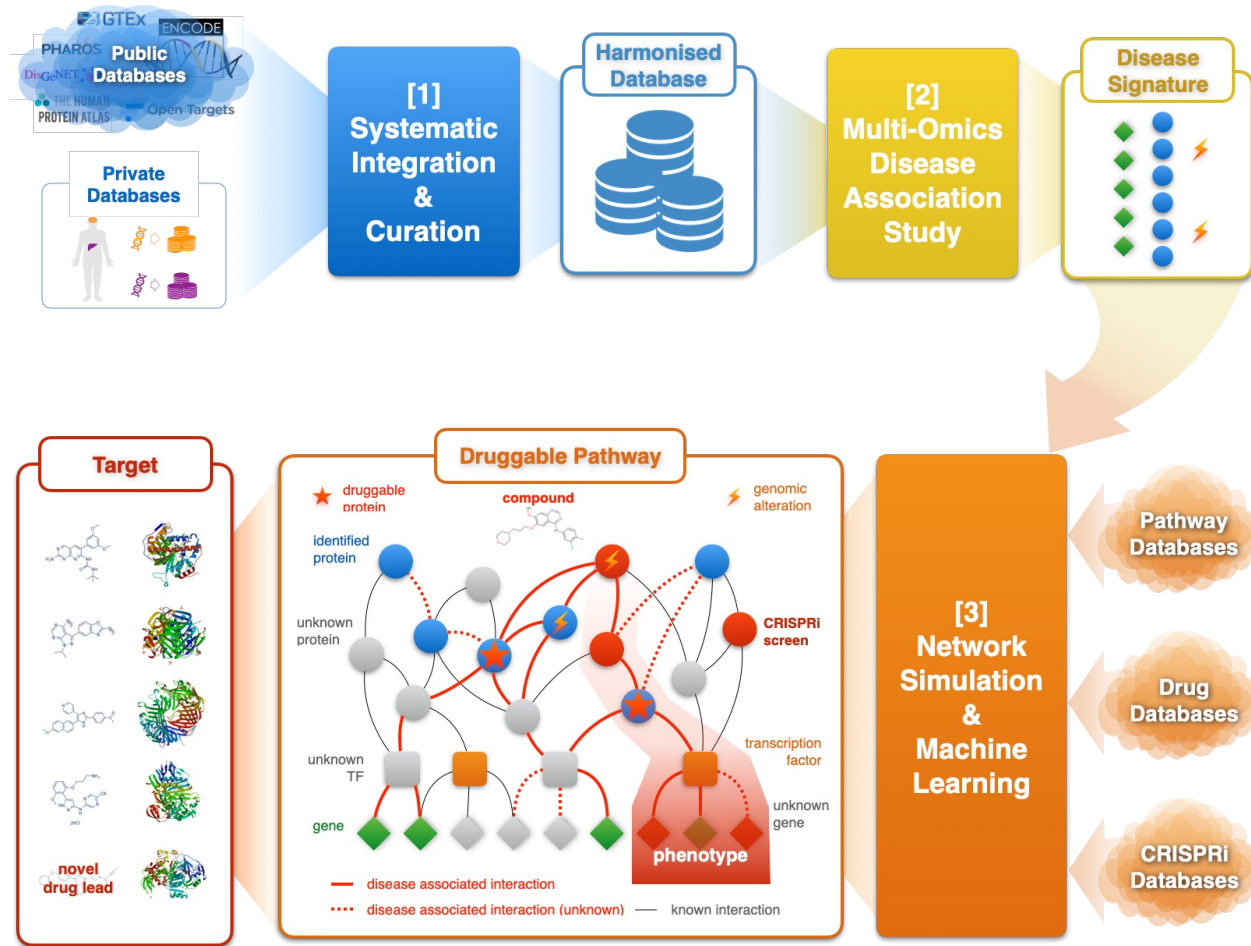
Random Walk with Restart

Which node is the most important?

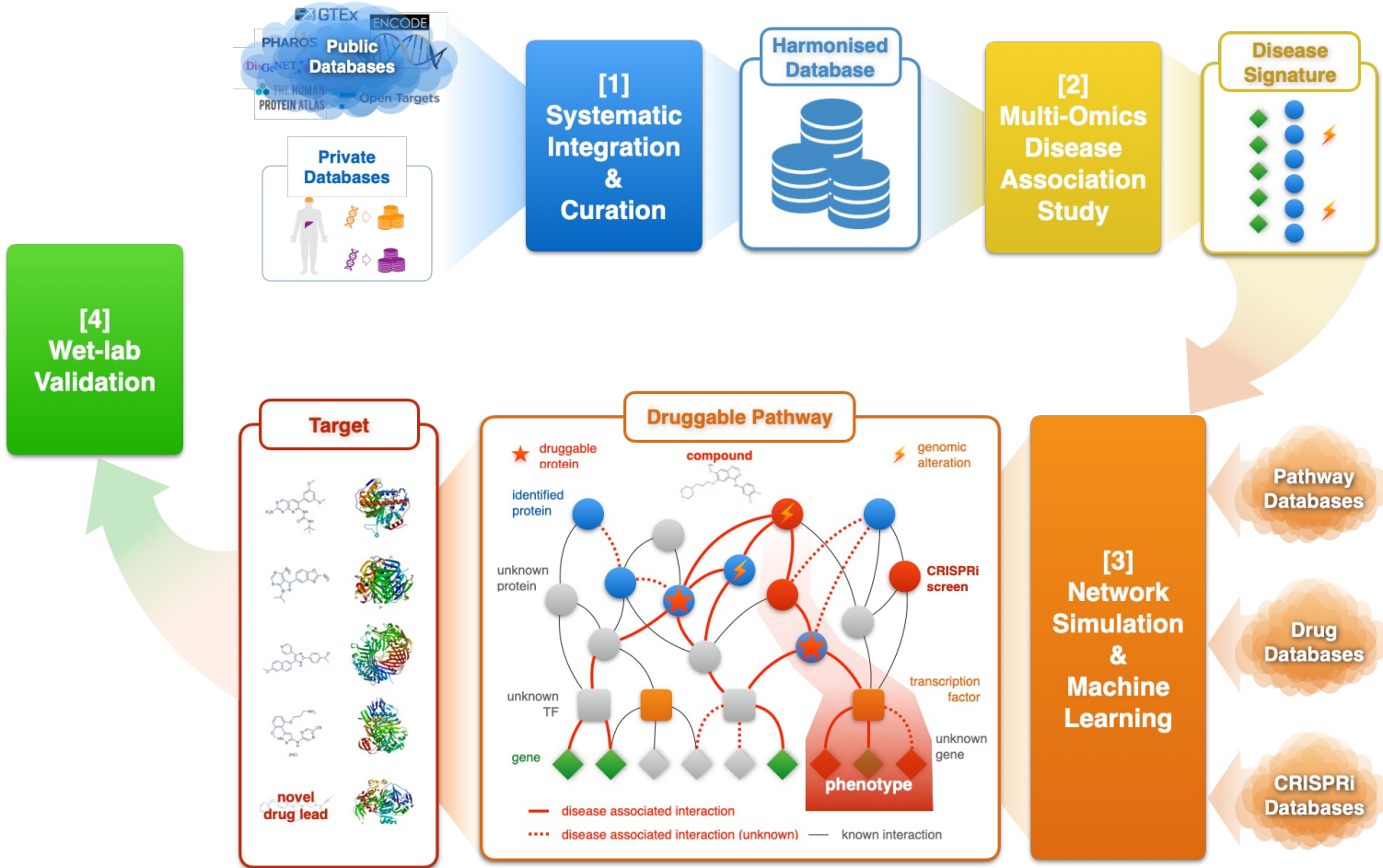
"D" is foremost based on number & weighting of in-links

"E" is next, due to the influence of D's link

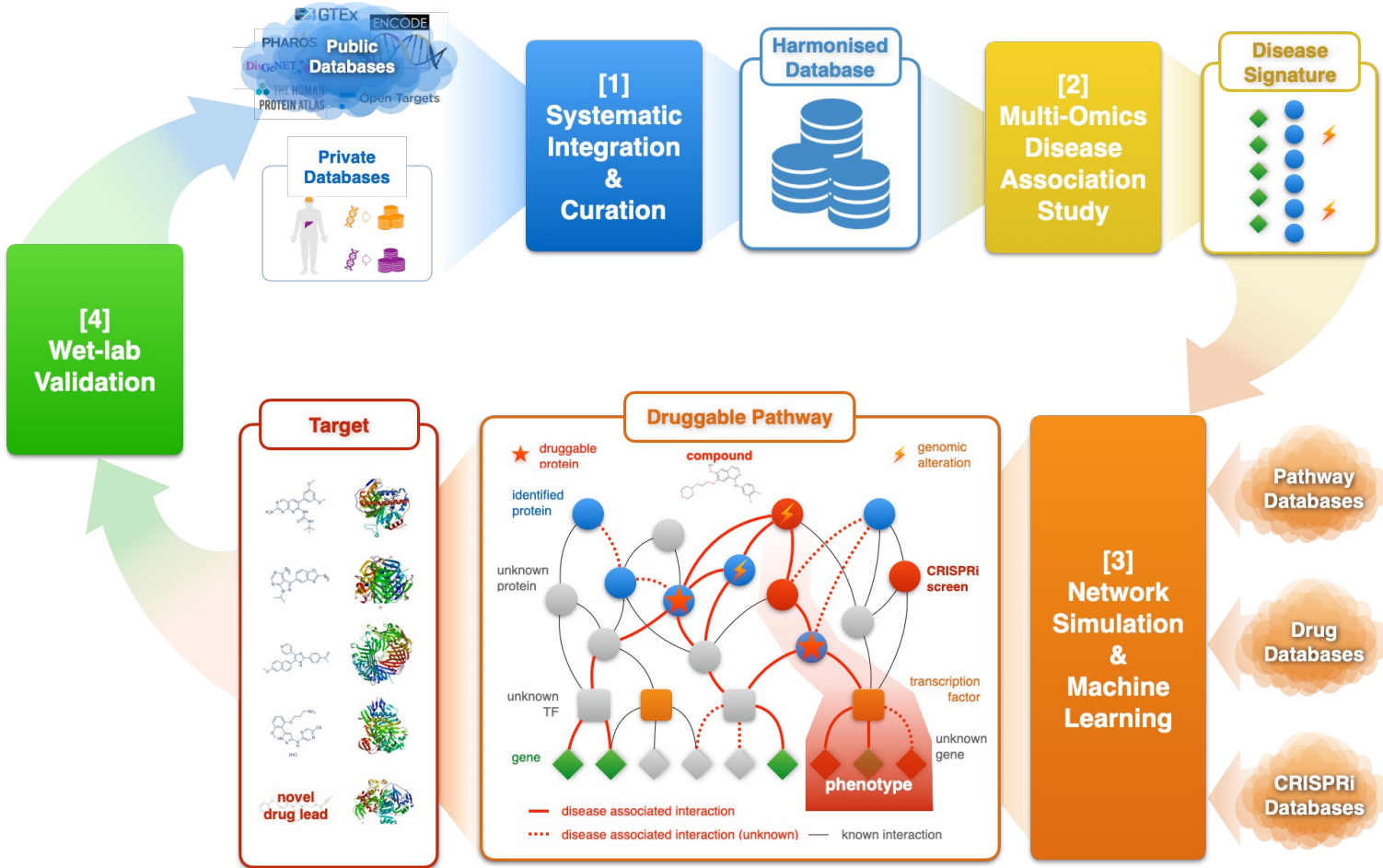
in silico Drug Discovery Pipeline



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in silico Drug Discovery Pipeline



Lab Members, Collaborators & Funders



Industry Partners & Funders

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Chemical Engineering and Biotechnology

David Fairen-Jimenez

IMS Institute of Metabolic Science
Metabolic Research Laboratories

Toni Vidal-Puig

wellcome-MRC
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Bertie Göttgens

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Early Cancer Institute

Harveer Dev

UNIVERSITY OF CAMBRIDGE
Department of Engineering

Yan Yan Sherry Huang

Open Targets

Ellen McDonagh
David Ochoa



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sanger
Institute

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UNIVERSITY OF PORTSMOUTH

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lao-Hveon Lee

Thank you

Namshik Han

nh417@cam.ac.uk

www.milner.cam.ac.uk/machinelearning