

SITC 2019

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& Convention Center

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Society for Immunotherapy of Cancer

Predicting Tumor Response to Immunotherapies

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SITC, November 9th 2019

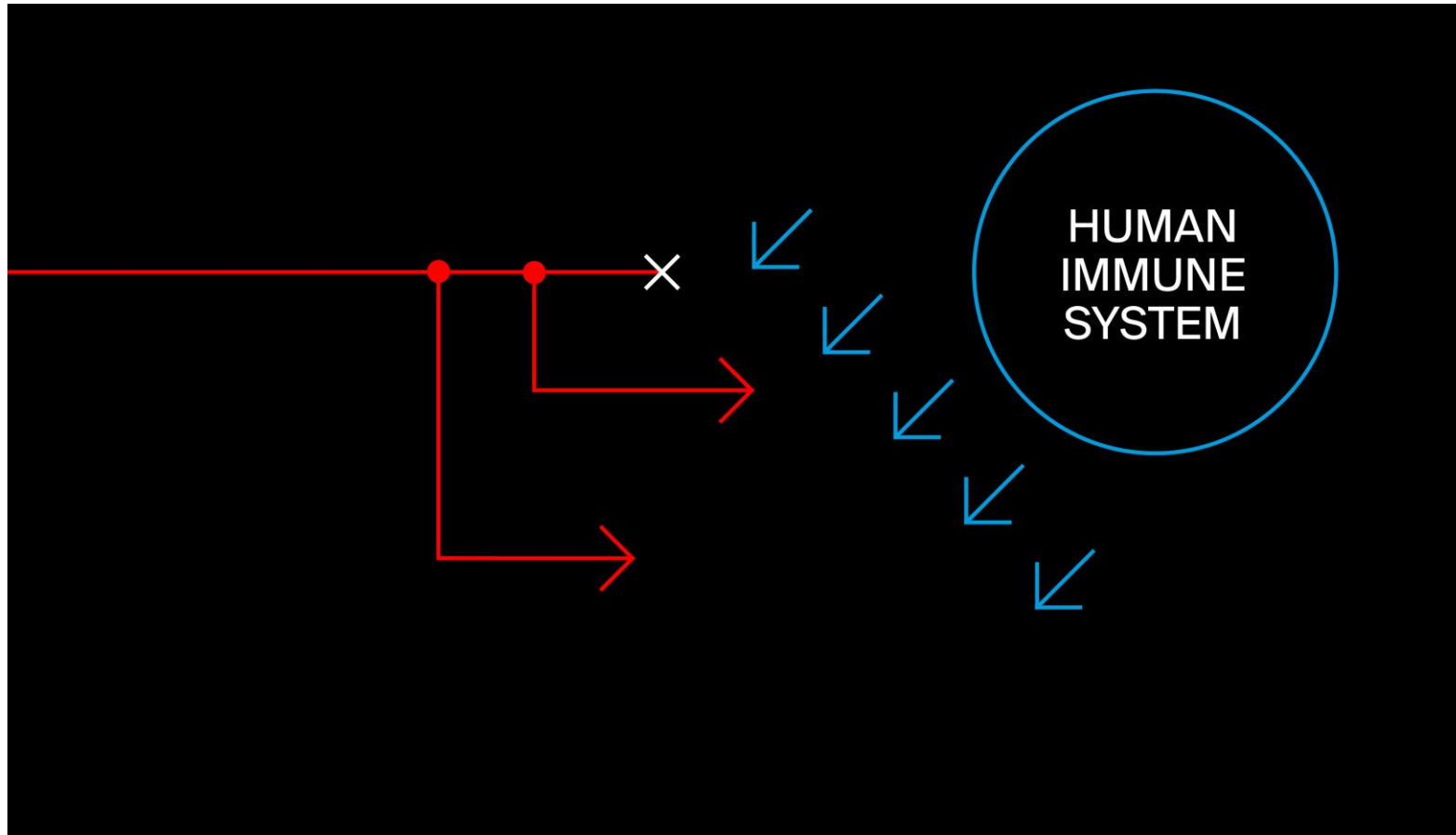
Disclosure

Honorarium from: Merck, Bristol-Myers Squibb

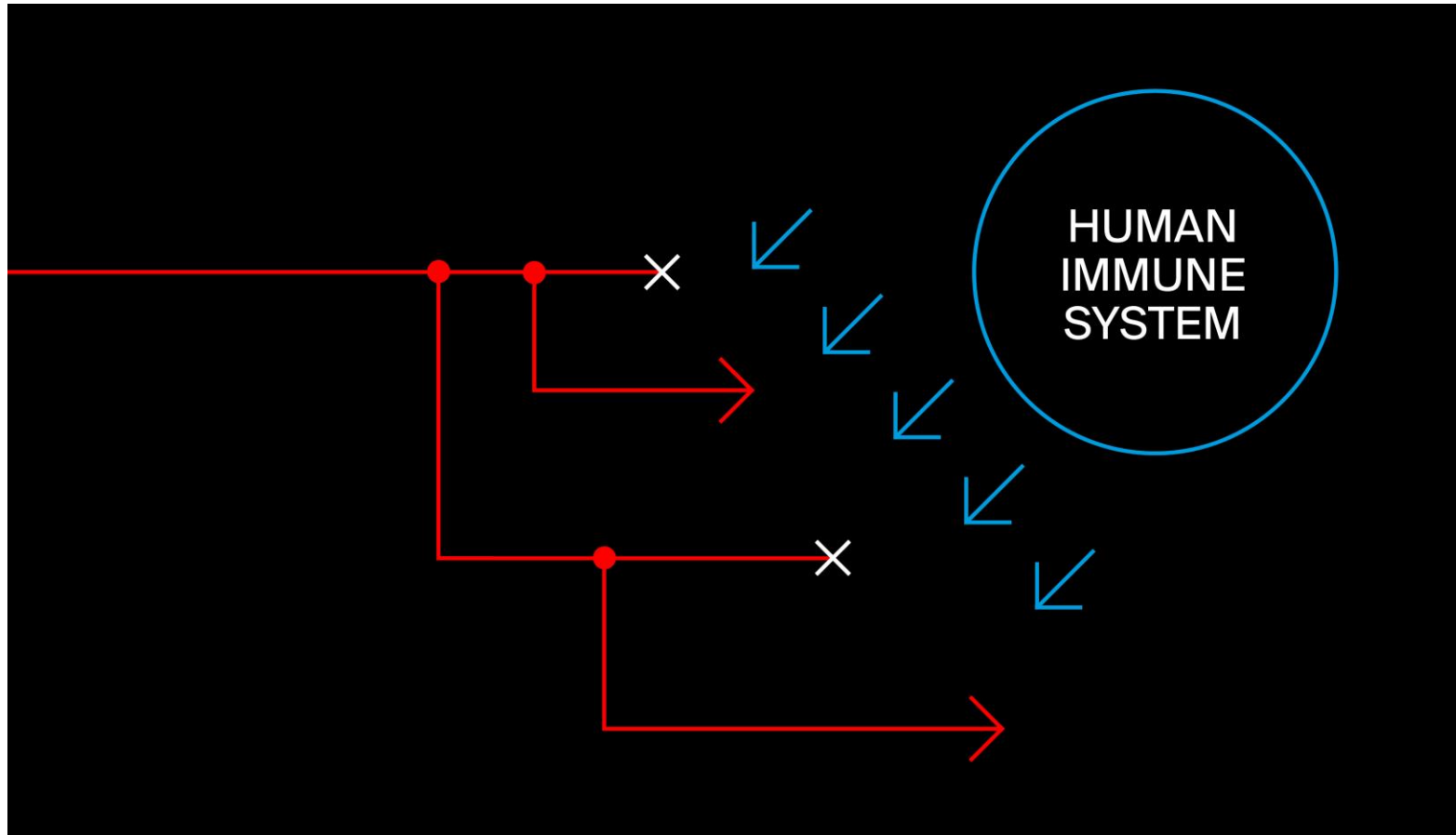
Pathogen/cancer arms race with the immune system



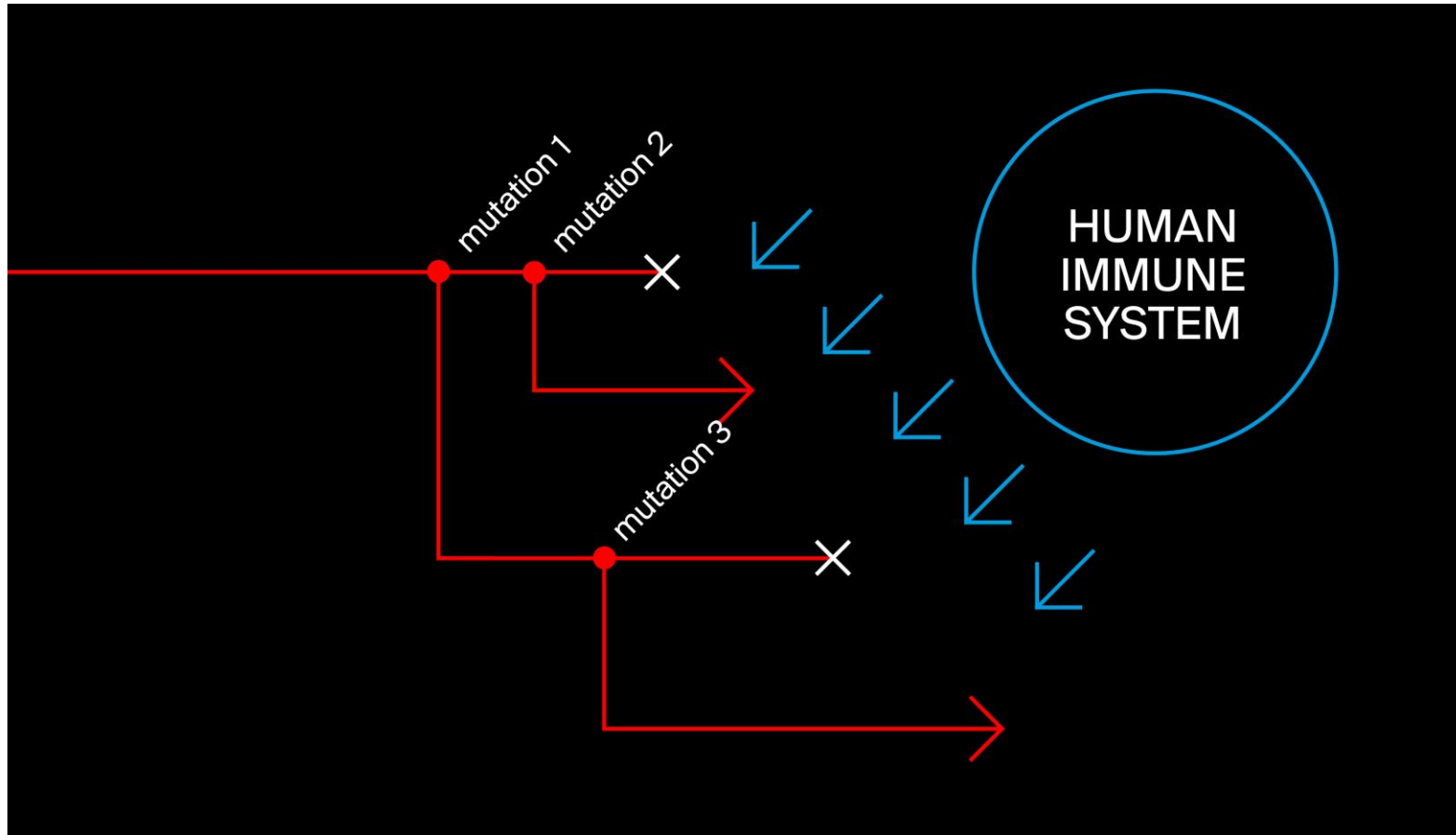
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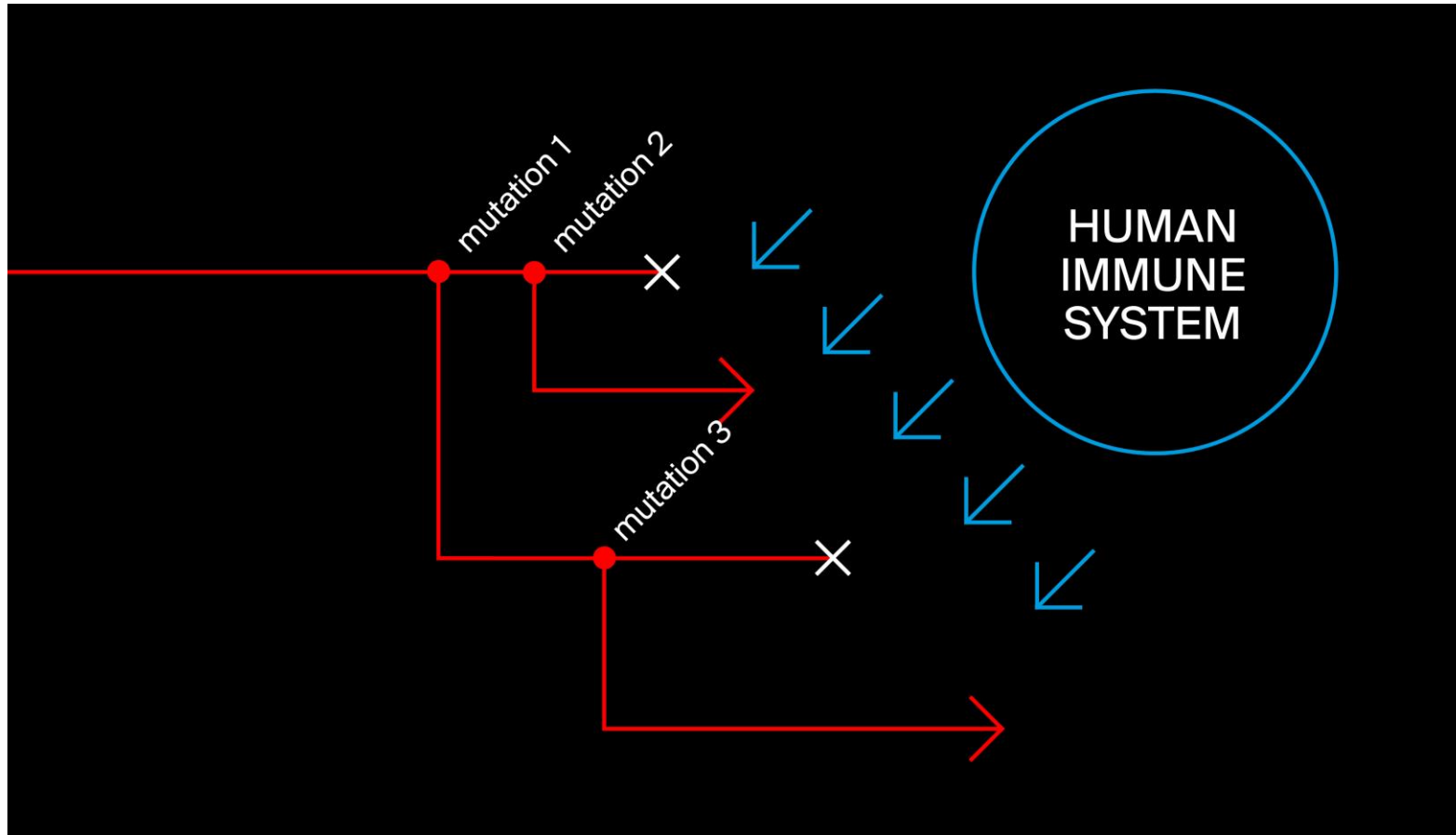
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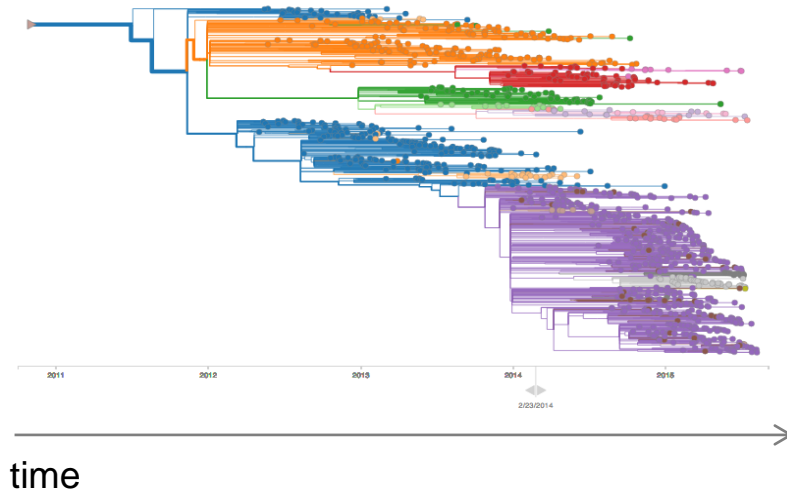
Pathogen/cancer arms race with the immune system



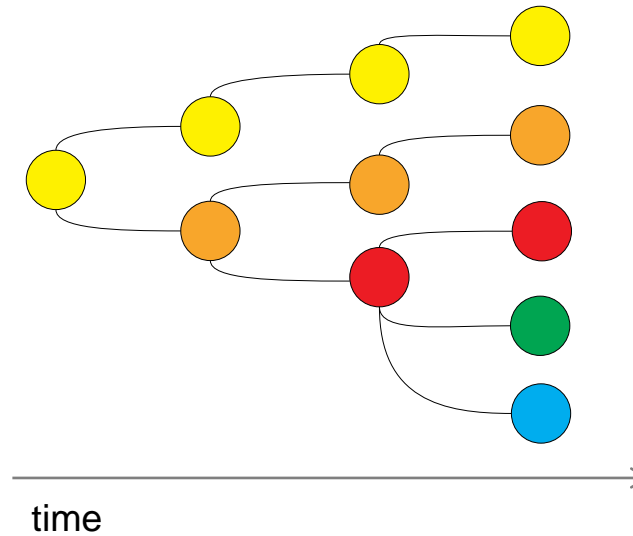
Can we quantify the immune interactions to predict the evolution?

Immune-fitness models for evolutionary predictions

Influenza



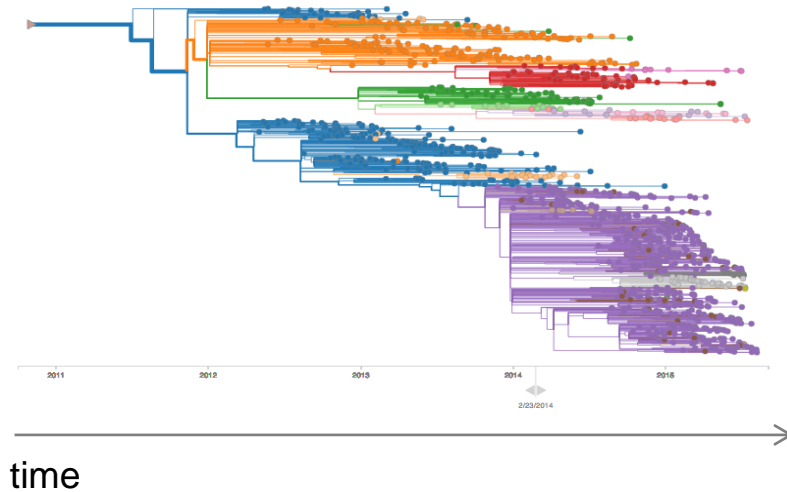
Cancer + immunotherapy



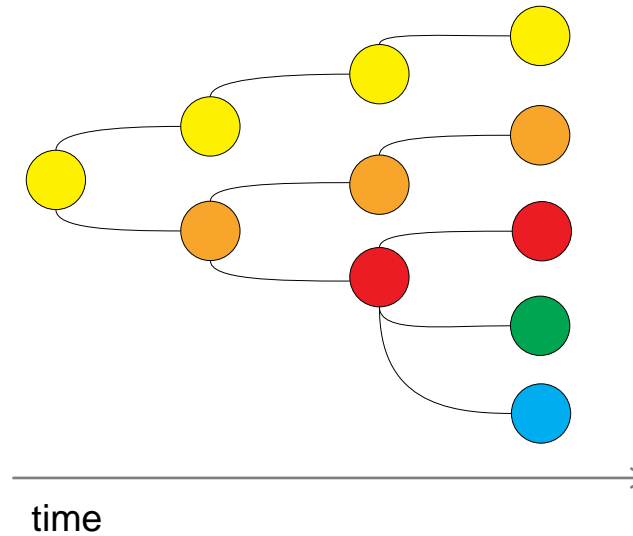
- High population heterogeneity
- Strong **immune** selection

Immune-fitness models for evolutionary predictions

Influenza



Cancer + immunotherapy



[Łuksza&Lässig, Nature 2014]

Evolutionary predictions based on a **fitness model** are currently used to support the influenza vaccine selection by the WHO.

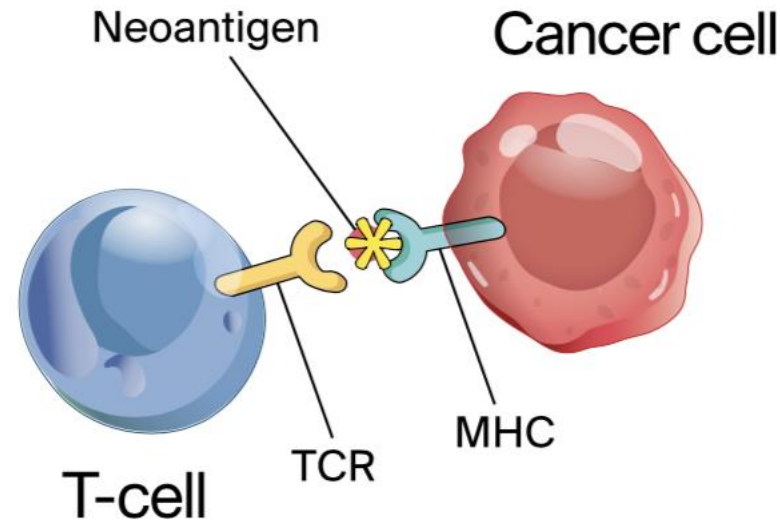
Neoantigen fitness model for tumors

The learning problem:

Predict tumor response to immunotherapy
from genetic data

Learning problem challenge 1: molecular complexity

MHC-presented neoantigens are potentially **immunogenic**:
recognized by the host's T-cells



[figure adapted from Sarkizova&Hacohen, News&Views, Nature 2017]

Learning problem challenge 1: molecular complexity



Fitness of a cancer cell is decreased due to recognition of presented neoantigens.

Goal: quantify the likelihood of these events using genomics patient data

Learning problem challenge 1: molecular complexity



- **Likelihood of presentation** based on binding affinities inferred with the netMHC algorithm, trained on (abundant) MHC assay data

Probability of neoantigen TCR-recognition

Our solution: let's copy how others do it:

Compare tumor neoantigens to pathogens

Tumor neoantigen:	PPSAR R GPL
Human Herpes Virus (HHV)-8:	PPSGQ R GPV

- Positive examples from the IEDB database of validated T-cell assays for microbial epitopes

Probability of neoantigen TCR-recognition

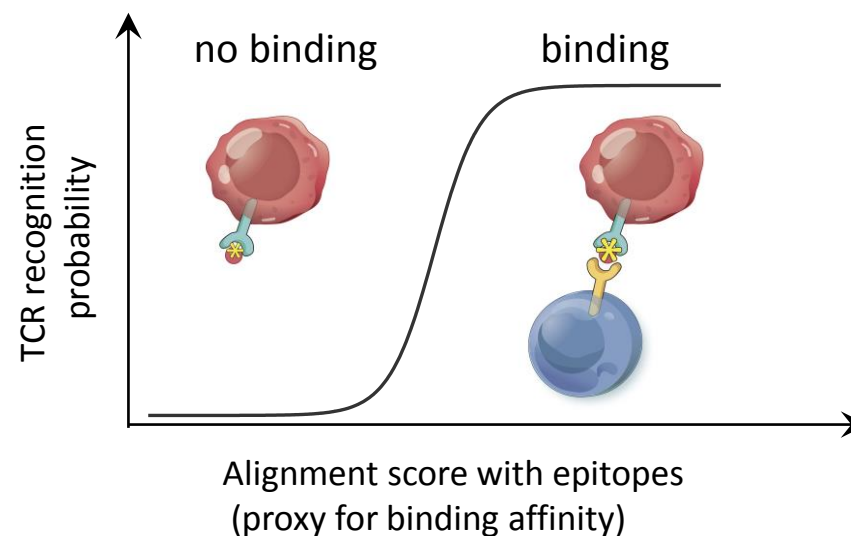
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Compare tumor neoantigens to pathogens

Tumor neoantigen:	PPSAR R GPL
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- Positive examples from the IEDB database of validated T-cell assays for microbial epitopes

And use a biophysical model:



Learning problem challenge 1: molecular complexity

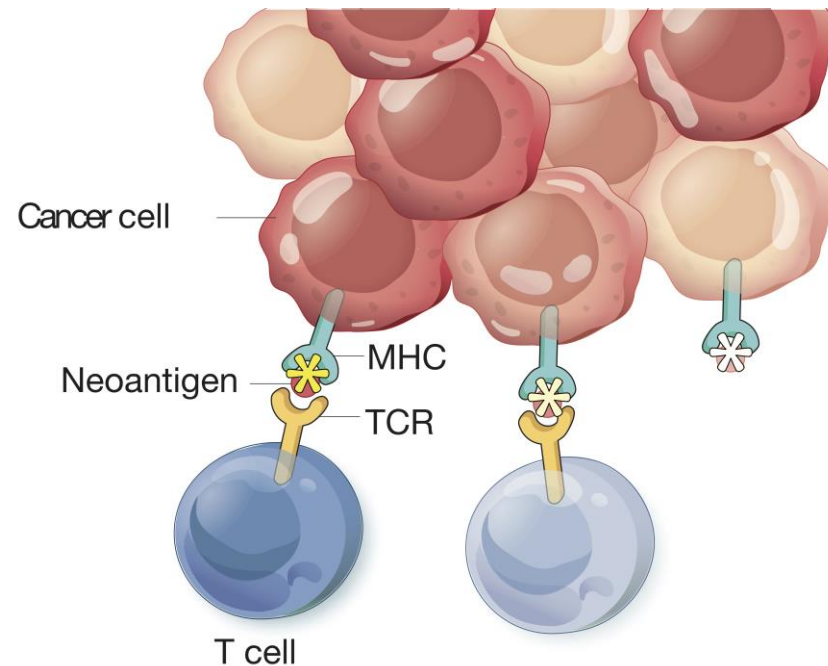


Fitness of a cancer cell is decreased due to recognition of presented neoantigens:

$$F = F_0 - (\text{MHC presentation probability} \times \text{TCR recognition probability})$$

Learning problem challenge 2: Tumor heterogeneity

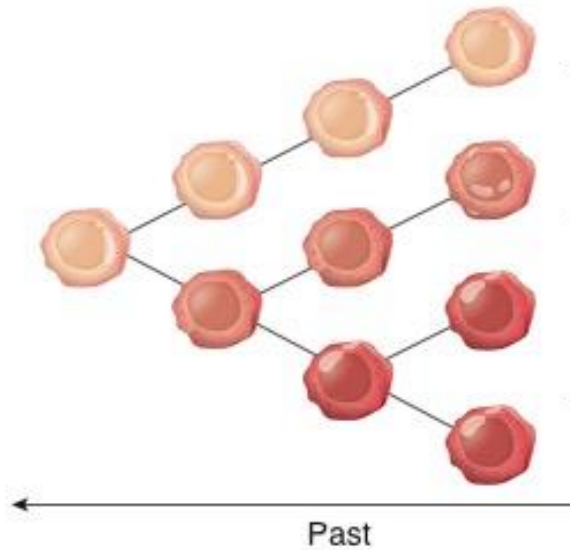
- Tumor cells are genetically heterogeneous
- They potentially have different immune interactions



[figure adapted from Sarkizova&Hacohen, News&Views, Nature 2017]

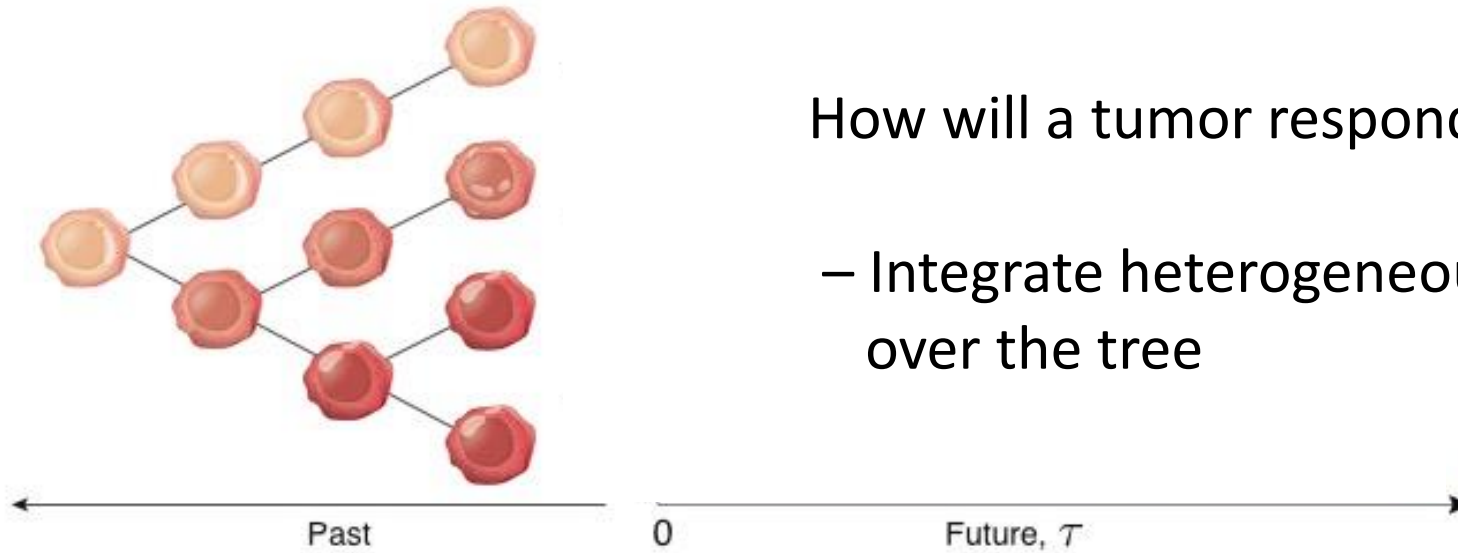
Learning problem challenge 2: Tumor heterogeneity

Tumor is an **evolving population of cancer cells**



Learning problem challenge 2: Tumor heterogeneity

Tumor is an **evolving population of cancer cells**

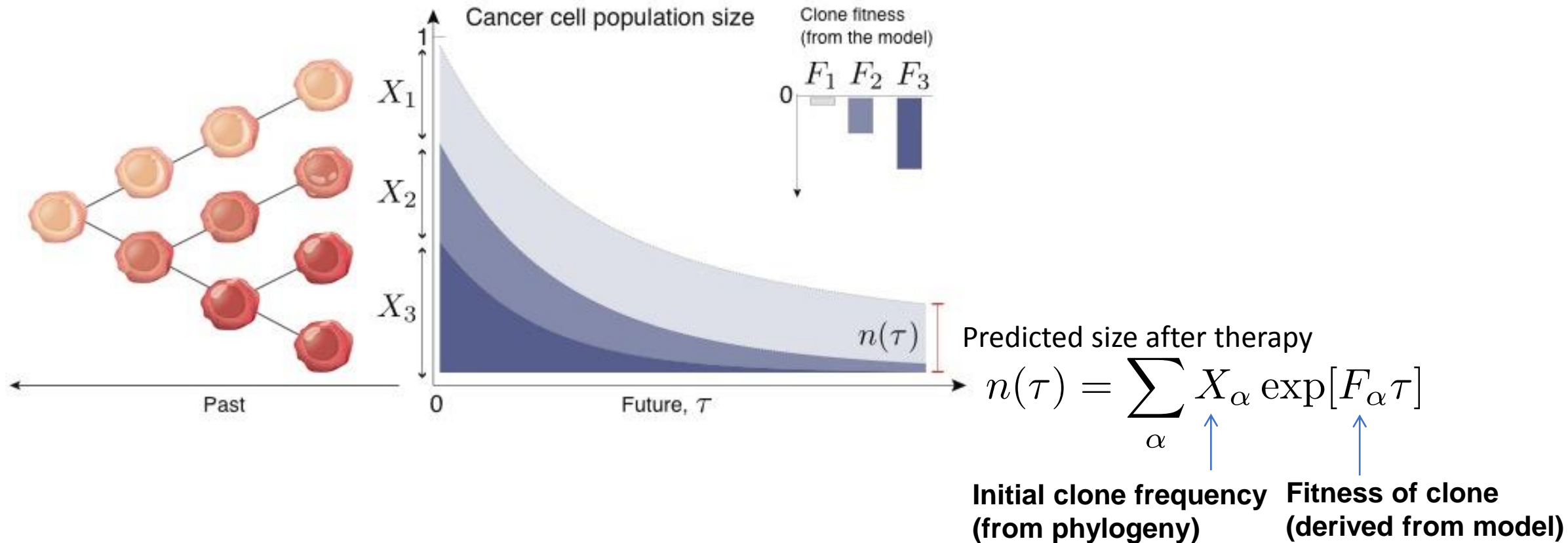


How will a tumor respond to therapy?

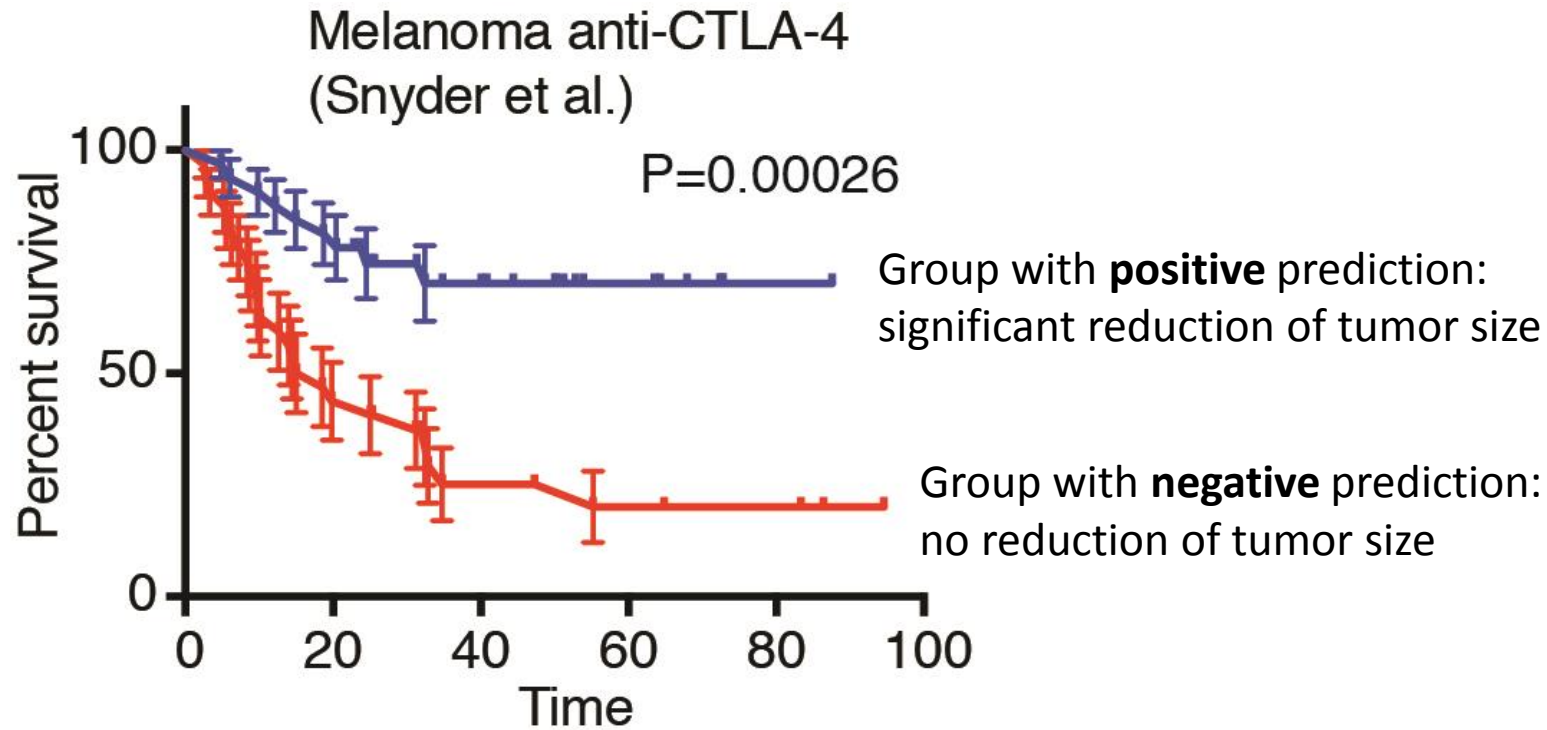
- Integrate heterogeneous fitness effects over the tree

Learning problem challenge 2: Tumor heterogeneity

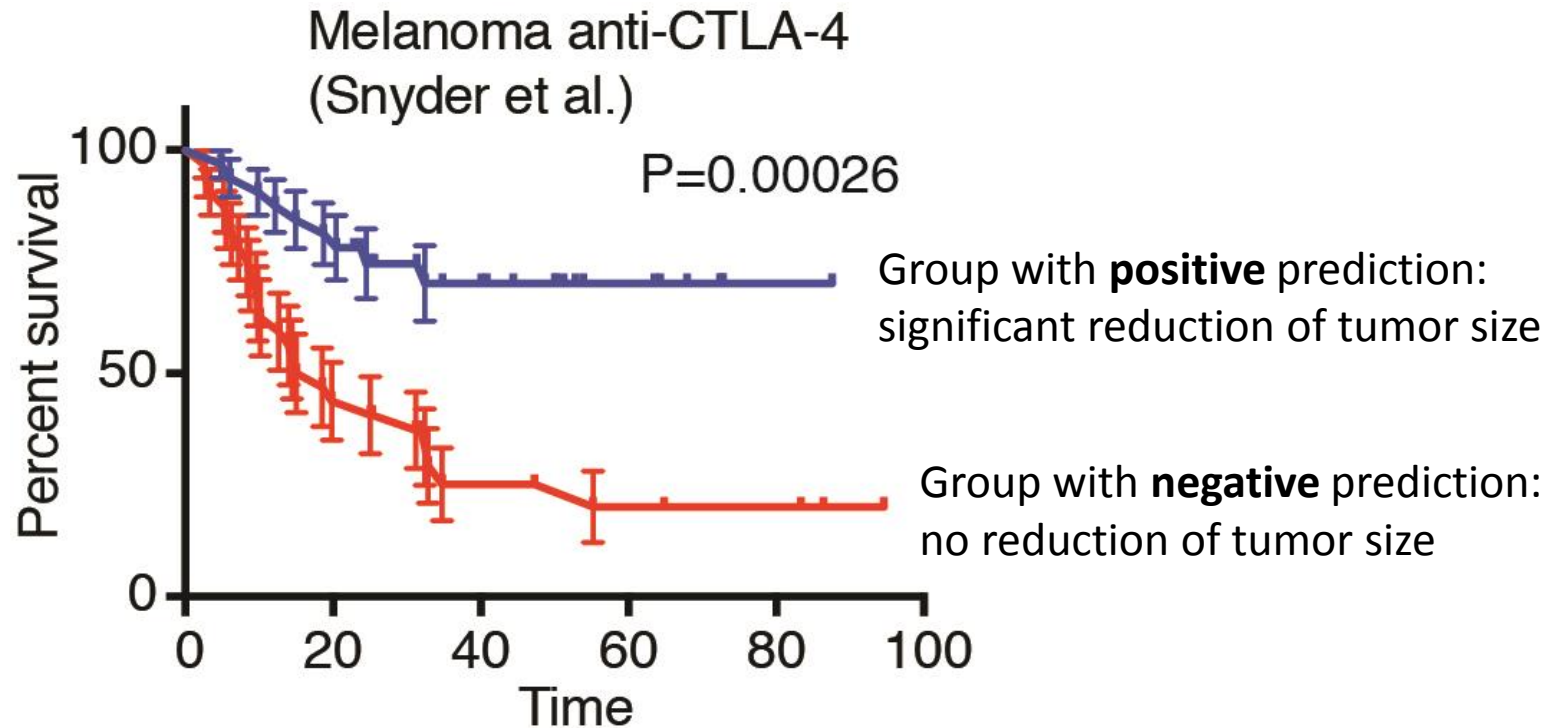
Tumor is an **evolving population of cancer cells**



Model based analysis is predictive of survival



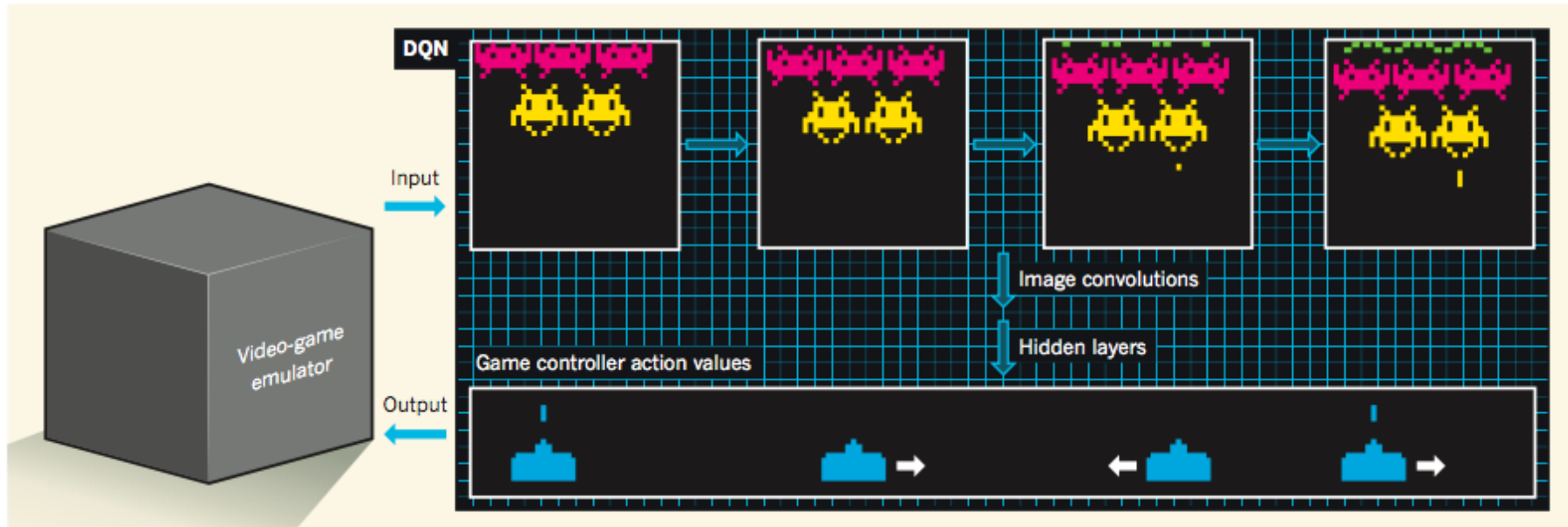
Model based analysis is predictive of survival



Validated on two anti-CTLA4 melanoma cohorts, anti-PD1 lung cohort, and (unpublished): NSLCC cohort, metastatic pancreatic cancer cohort.

Are models still needed?

Deep reinforcement algorithm learned to play 49 vintage computer games **without a priori knowledge** of the games and rules (Mnih et al. Nature 2015)



Go: Silver et al Nature 2017, Go & chess, Silver et al Science 2018:

Are models still needed?

Silver, D. et al. Nature, 2016:

*AlphaGo first studied **30 million positions** from expert games, gleaning abstract information on the state of play from board data, much as other programs categorize images from pixels. Then it played against itself **across 50 computers**, improving with each iteration, a technique known as reinforcement learning.*

- Patient cohorts are not big data
- High molecular & population complexity
- These differences favor constrained **mechanism-informed models** for making evolutionary predictions.

Summary

- We develop **biophysically motivated models of immune interactions**
- Such models allow us to **predict the evolution** of cancer and viral pathogens
- Such predictions can **inform treatment strategies**

Acknowledgments

Cancer and immune system

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Viral evolution

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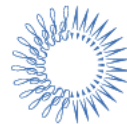
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PEW



AACR

American Association
for Cancer Research