# INTERPRETATION OF NGS TUMOR DATA









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DavidTamborero







#### **OBJECTIVES OF THE SLIDE SET**

- Review general concepts for tumor NGS data interpretation
- Explain how to analyse NGS data of an individual case
- Hands-on with resources supporting the task



Introduction

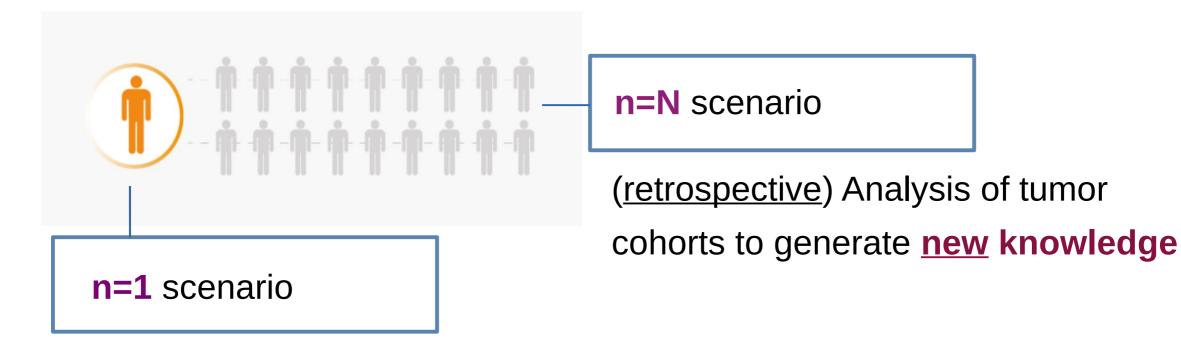
Functional relevance of variants

O Clinical actionability of variants

O Resources for variant interpretation



# From cohorts to individual tumors



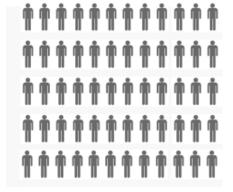
(<u>prospective</u>) Analysis of a tumor individual based on <u>current</u> knowledge

#### EXAMPLE OF COHORT VS INDIVIDUAL ANALYSIS









> Analyse **patterns** of mutations across (large) cohorts of sequenced tumors



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Discovery of ~ 80 unique mutational signatures

( capturing the different mutational processes that [tumor] cells undergo)





> Identify which of these **known**mutational processes have occurred
in this tumor individual





# From cohorts to individual tumors



n=N scenario

<u>(retrospective)</u> Analysis of tumor

cohorts to generate <u>new</u> knowledge

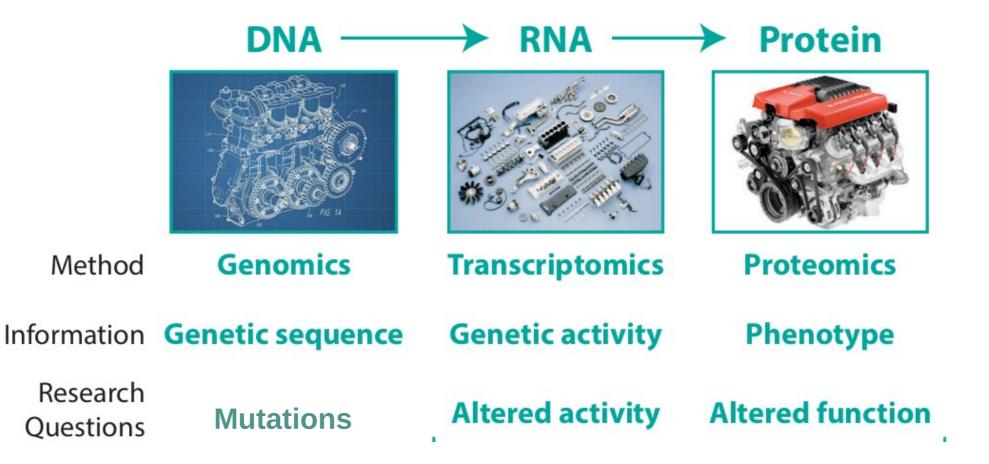
**n=1** scenario

(<u>prospective</u>) Analysis of a tumor individual based on <u>current</u> knowledge

#### HIGH-THROUGHPUT DIAGNOSTIC ASSAYS







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#### DNA



Method

**Genomics** 

Information Genetic sequence

Research Questions

**Mutations** 

> Single nucleotide mutations, indels, copy number alterations and/or fusions

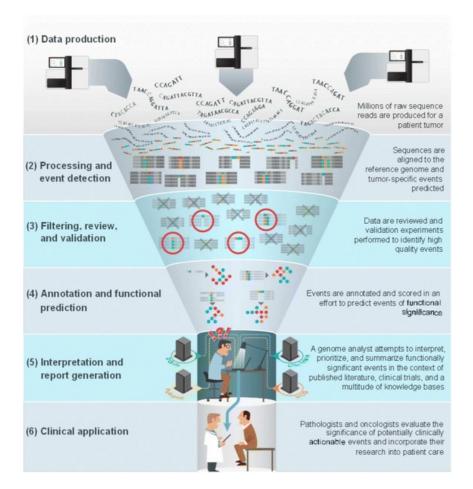
> Use in the clinics: cancer gene panels and WGS/WTS\*

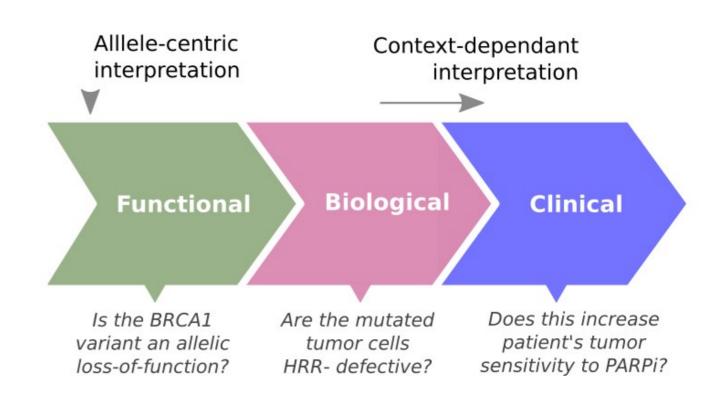
\* Hodder et al, Nature Medicine 2024 (PRAINING) et al, BMJ 2024 (sarcomas)

#### n=1 INTERPRETATION OF GENE MUTATIONS









Good et al. Genome Biology 2014



O Introduction

Functional relevance of variants

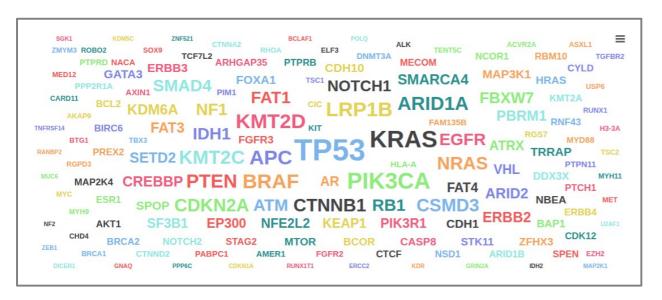
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#### ~600-700 CANCER GENES DESCRIBED



#### > Most recurrently mutated cancer genes:



Intogen website, 2025

#### **ONCOGENES**



> WT function **promotes** cell growth, division and survival

#### **TUMOR SUPPRESSORS**



> WT function **inhibits**cell cycle and **ensures**apoptosis and DNA
damage

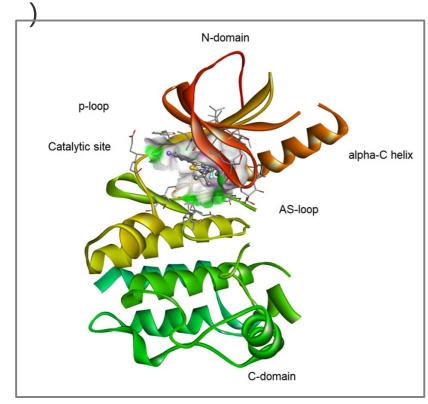
#### **HOW A MUTATION CAN BE TUMORIGENIC?**





> Example: BRAF p.V600E (gain-of-function)

(chr7:g.140453136A>T



BRAF V600E with a ligand bound (PDB ID: 4XV2)

- > BRAF is a kinase that in WT is phosphorylated via RAS-regulated (extracellular) growth signaling
- > The mutation codes for a glutamic acid, which mimics the effect of adding phosphate groups
- > The protein is 'tricked' to flip into a constitutively active form
- > Cells with this mutant allele are in overactive proliferation/survival signaling state

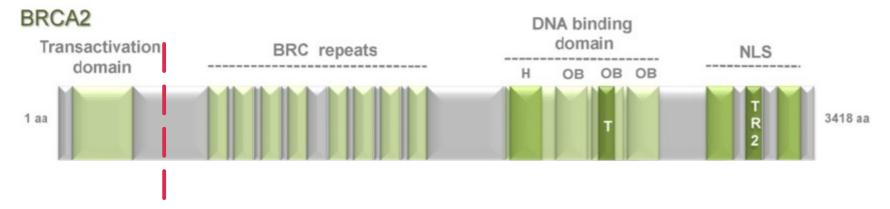
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> Example: BRCA2 p.Gln366Ter (loss-of-function)

(chr13:g.32906712C>T

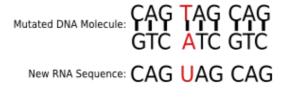


#### Normal



Normal Amino Acid Sequence: Gln -Gln -Gln

#### **Nonsense Mutation**



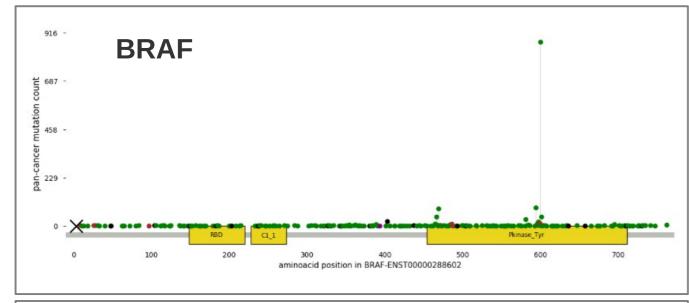
New Amino Acid Gln stop

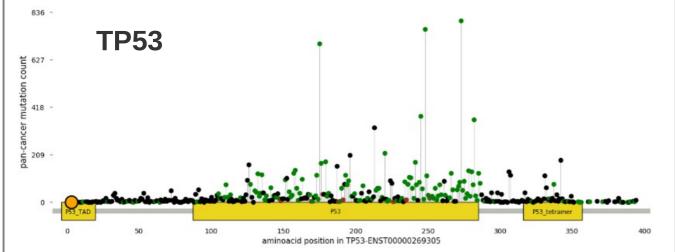
- > The mutation creates an aberrantly premature stop codon
- > Only a small mRNA segment is transcribed, preventing normal gene function.

#### **MUTATIONS OBSERVED IN ~25,000 TUMORS**









- > Tumor mutations are **fixed** based on:
  - (a) how 'likely' are they to occur ([somatic] <u>mutational processes</u>)
  - (b) the <u>selective advantage</u> that they confer for clonal expansion

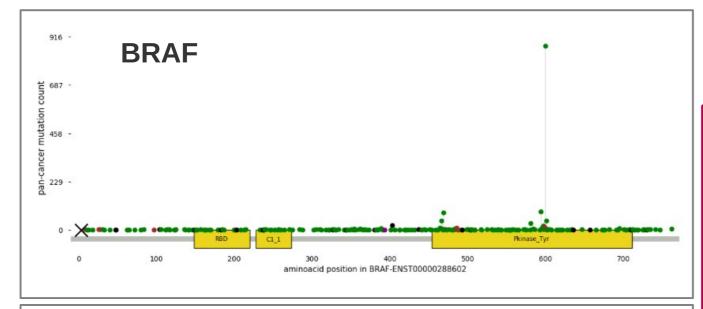


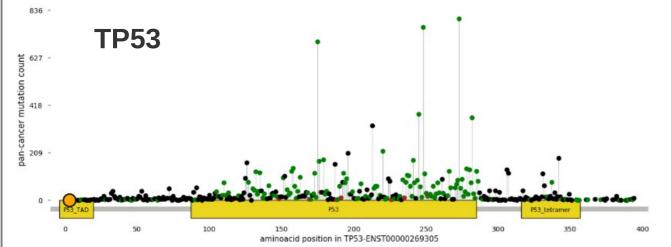
Public MTBP website, 2025

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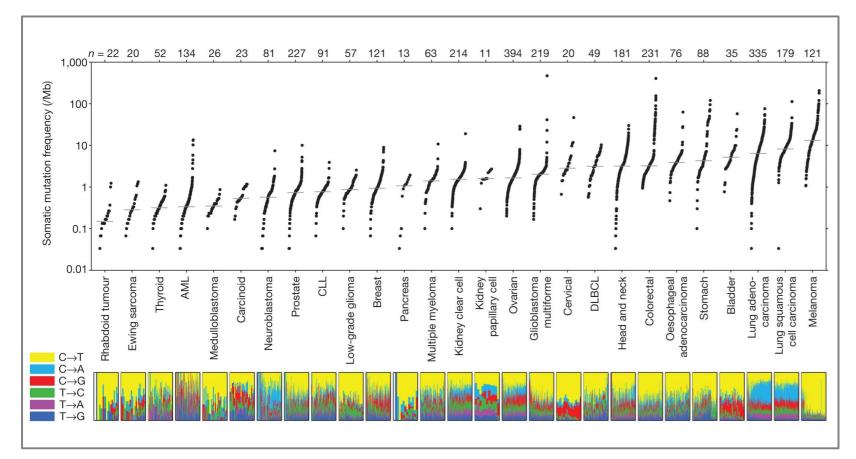
Public MTBP website, 2025

#### **HOW LIKELY IS A MUTATION TO OCCUR?**





> Frequency and type of somatic mutations in first TCGA/ICGC release (~3,000 tumors):



- > Mutational processes in a cell are the result of:
  - (A) <u>Endogenous</u> and <u>exogenous</u> factors
  - (B) Performance of the **DNA** damage repair

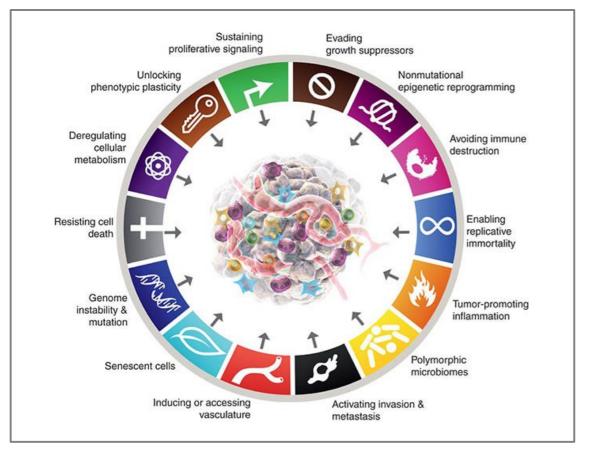
Lawrence M, Nature 2013

#### WHAT IS A (CLONAL) SELECTIVE ADVANTAGE? Karolinska Institutet



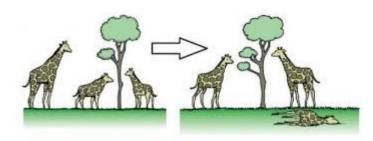


> Tumorigenic mutations make the cell to acquire tumor hallmarks:



- > These are otherwise **normal processes** that occur at the wrong place and/or time
  - > They provide an edge to the malignant cells to outcompete others during **clonal** expansion





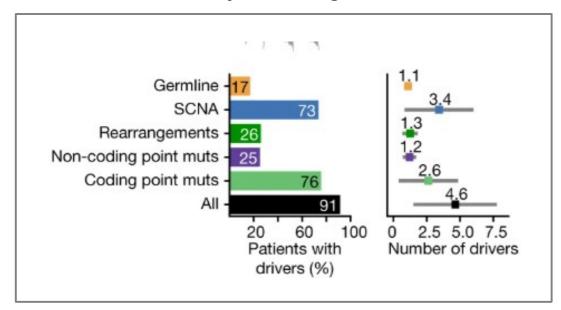
Hanahan D. Cancer Disc 2022

### ARE ALL THE MUTATIONS THAT OCCUR IN CANCER GENES TUMORIGENIC?



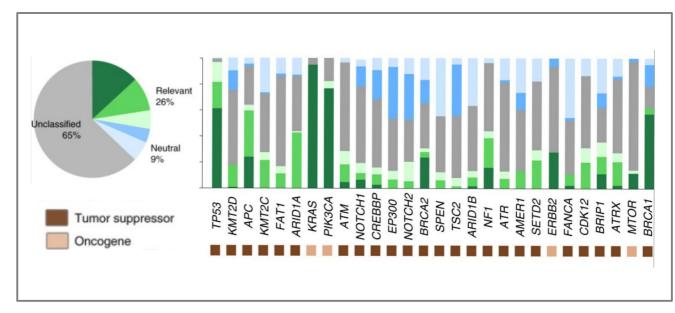


> Tumors bear a mean of ~4 genomic alterations likely tumorigenic:



PCAWG, Nature 2020

> ~25% of the tumor mutations observed <u>in cancer</u> genes have evidence of being tumorigenic:



Tamborero D, Nature Cancer 2022

#### INTERPRETATION OF GENE MUTATIONS

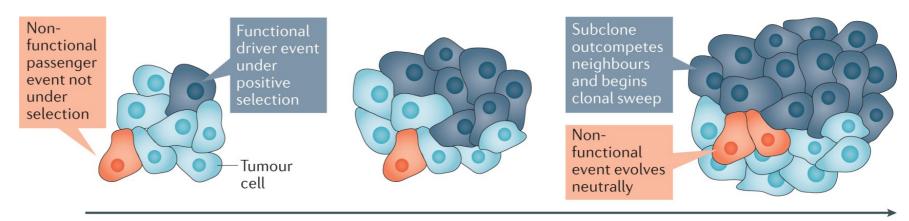




> Importance of classifying mutations observed in a tumor by their **functional relevance**, as not all of them (<u>even in cancer genes</u>) are **tumorigenic** 



#### **Driver** vs **Passenger** mutations



Time



O Introduction

O Functional relevance of variants

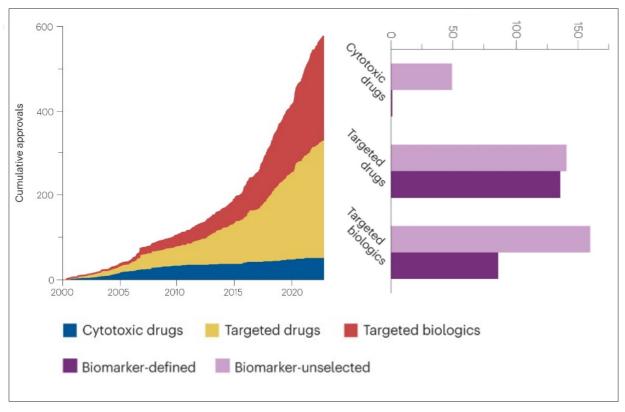
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#### PRECISION CANCER ONCOLOGY

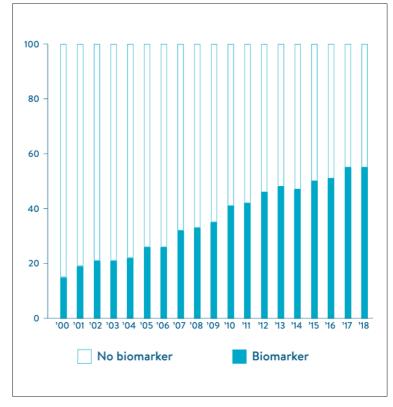


#### > FDA drug approvals (2000-2022)



Scott et al Nat Rev Drug Disc 2023

#### > Clinical Trials (2000-2018)



Vadas et al L.E.K. Consulting 2018

#### APPROVED DRUG GENOMIC BIOMARKERS





Drug Name	Biomarker(s)	Cancer Type(s)	Regulatory Approval
Imatinib	BCR-ABL fusion	Chronic myelogenous leukemia (CML), gastrointestinal stromal tumors (GIST)	FDA (2001), EMA (2002
Gefitinib	EGFR exon 19 deletion, L858R mutation	Non-small cell lung cancer (NSCLC)	FDA (2003), EMA (2009
Erlotinib	EGFR mutations	NSCLC, pancreatic cancer	FDA (2004), EMA (2005
Lapatinib	HER2 overexpression	HER2-positive breast cancer	FDA (2007), EMA (2008
Alpelisib	PIK3CA mutation	HR-positive, HER2-negative breast cancer (with fulvestrant)	FDA (2019), EMA (2020)
Selpercatinib	RET fusion, RET mutation	NSCLC, medullary thyroid cancer, other RET-altered cancers	FDA (2020), EMA (2021
Entrectinib	NTRK fusion, ROS1 fusion, ALK fusion	Solid tumors (tissue-agnostic)	FDA (2019), EMA (2020
Sotorasib	KRAS G12C mutation	NSCLC, colorectal cancer	FDA (2021), EMA (2022
Krazati (adagrasib)	KRAS G12C mutation	Colorectal cancer (with cetuximab)	FDA (2024), EMA (2025
Vorasidenib	IDH1 or IDH2 mutation	Grade 2 gliomas (astrocytoma or oligodendroglioma)	FDA (2024), EMA (2024

~190 prescriptions (genomic alteration + cancer type\* + drug) approved for **standard-of-care** 

\*note also pancancer approvals

Examples of genomic markers for FDA/EMA approved therapies

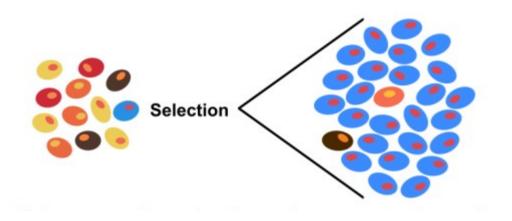
#### PRIMARY AND SECONDARY RESISTANCE







> Drugs that target a specific **molecular mechanism** that the malignant cells are "addicted" to pursue the tumor collapse (sparing healthy cells)



- > If this is not true for a significant number of tumor cells → **no clinical response**
- > If this is not true for a (small) subset of cells
  - → **clinical relapse**

#### THERAPY-RESISTANT TUMOR CLONES







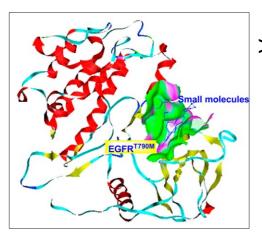




> Relapse to Vemurafenib in BRAF V600E melanoma

Wagle et al JCO 2011

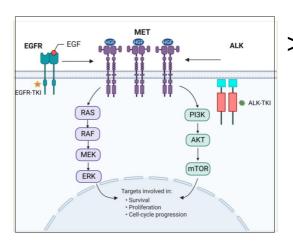
#### **On-target resistance**



> EGFR T790M prevents binding of 1<sup>st</sup> generation EGFR inhibitors

Song et al JMC 2016

#### **Off-target resistance**



> MET amplification as parallel driver to EGFR / ALK signaling

Qin et al Cancers 2023

#### **ACTIONABILITY OF (DRIVER) MUTATIONS**



- > They can be biomarkers of drug response
  - > Confer **sensitivity** or **resistance** to a certain therapy
    - > Some are already approved for <u>SoC use</u>, others are tested in ongoing <u>clinical trials</u>
    - > In some institutions, genomics are also used for n=1 off-label prescriptions
- > They can also be used as diagnostic and/or prognostic markers

> Others: secondary/incidental findings (including clonal hematopoiesis)



O Introduction

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Resources for variant interpretation

#### n=1 VARIANT INTERPRETATION





## From cohorts to individual tumors



Michael Ramirez

#### **KNOWN VARIANT EFFECTS**



- > Based on existing pre-clinical and clinical data
  - > "In vitro Ba/F3 cell studies show FGFR2 p.Cys62Tyr mutation to be activating as measured by increased transformation activity compared to wildtype."
  - > "RNA studies show that BRCA2 p.Val211Leu disrupt the normal splicing of the encoded transcript."
  - > "Germline TP53 p.Arg337Cys mutations have been demonstrated to cause Li-Fraumeni syndrome."
  - > "ST7 (e2)-MET (e2) fusion have been observed in several patients relapsed to ALK-inhibition therapy."

#### **KNOWLEDGEBASES**





- > Several international efforts to gather existing knowledge on variant effects
  - > **Scope** (e.g. germline vs tumor somatic)
  - > Data model (e.g. format, classification framework)
  - > <u>Curation and distribution</u> (e.g. community vs in-house, license model)

















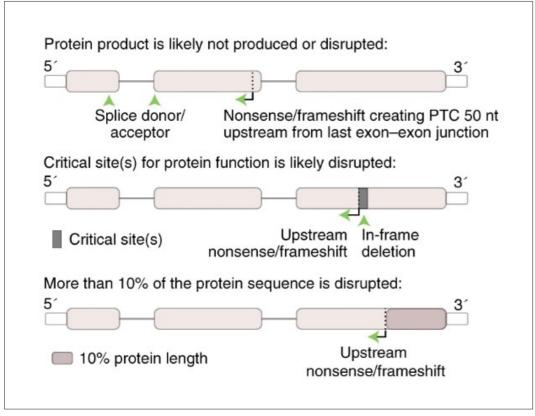


#### **VARIANT EFFECT ASSUMPTIONS**





> Some variants have been not characterized but their effect can be (safely) assumed



> Examples of mutations that likely disrupt the protein WT function

D Tamborero, Nature Cancer 2022

#### **VARIANT PREDICTIONS**





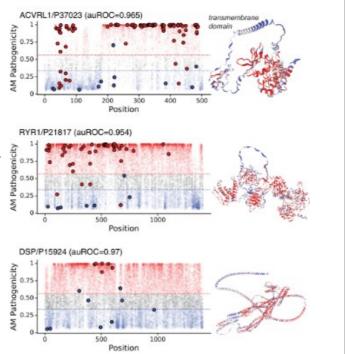
> When the variant effect is not known and cannot be assumed, computational predictions provide the lowest level of evidence

### Accurate proteome-wide missense variant effect prediction with AlphaMissense





ML-algorithm predicting the impact of a given mutation based on e.g. conformational changes in the affected protein



## FROM VARIANT ANNOTATION TO CLASSIFICATION



> Underlying evidence required to **associate** the variant with a **(clinical) action** 

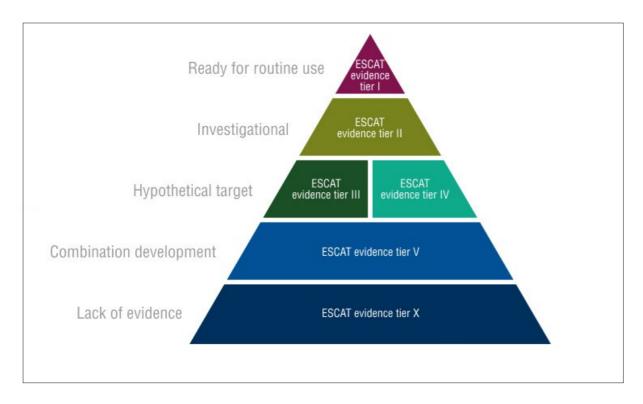
Pathogenic mutation	Pathogenic variants are genetic alterations that are known or strongly suspected to cause disease or a significant increase in disease risk.	Has clinical effect
Likely pathogenic variant	Likely pathogenic variants are genetic alterations that have strong evidence to suggest that they are disease-causing, but may not have been fully validated through research or clinical observation. These variants are often based on indirect evidence, such as family history or population studies.	Has clinical effect
Variant of unknown significance (VUS)	Variants of unknown significance are genetic alteration or mutation that has been identified through genetic testing, but for which there is not enough evidence to determine whether it is pathogenic or benign.	Unknown clinical effect
Likely benign variant	Likely benign variants are genetic alterations that are suspected to be benign but have not yet been fully validated through research or clinical observation.	Has no clinical effect
Benign	Benign variants are genetic alterations that are not known to cause disease or increase disease risk.	Has no clinical effect

> ACMG/AMP classification framework for variant pathogenicity in germline genetic diseases

## FROM VARIANT ANNOTATION TO CLASSIFICATION



> Underlying evidence required to **associate** the variant with a (clinical) action



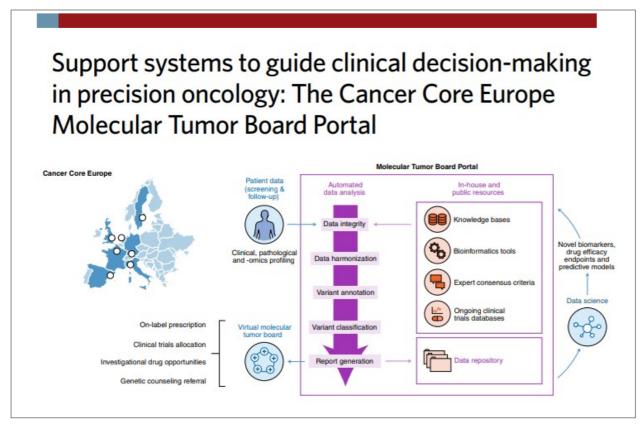
> ESMO classification framework for ranking variant actionability

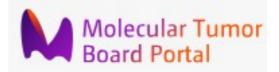
#### **CLINICAL DECISION SUPPORT SYSTEMS**





> Automates the annotation of variants based on multiple resources and their classification according to predefined criteria





D Tamborero, Nature Medicine 2020



### **CONCLUSIONS**



> n=1 interpretation is about using existing knowledge and methods to support the most informed downstream application

> It is key to understand all the **pros/cons** of each **resource** that is employed

> Importance of using computational tools to **scale** the process following a **consistent** workflow





The Personalised Cancer Medicine Program at Karolinska Institutet

#### Cancer Proteogenomics and Clinical Translation











Vetenskapsrådet







































