Definition of a driver. Cellular/tissular mechanisms supporting that a driver becomes a target Multiple drivers, mechanisms of resistance

Prof. Dr. Christian Rolfo, MD, PhD, MBAh
Head of Phase I – Early Clinical Trials Unit
Director of Clinical Trials Management Program
Antwerp University Hospital & Center for Oncological
Research (CORE), Antwerp University
Belgium





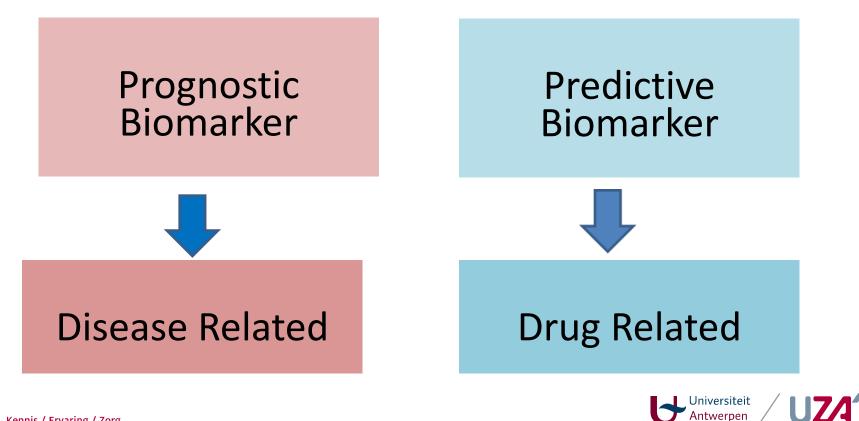
Kennis / Ervaring / Zorg

#### Disclosures

- Novartis International Speaker bureau
- Boeringher Speaker Bureau
- MSD Merck Speaker Bureau
- Oncompass Molecular Profile Steering Committee board Member
- Mylan Biosimilars Advisor for NSCLC
- Guardant Health speaker bureau
- OncoDNA research grant for exosomes



"A biological molecule found in blood, other fluids or tissues, that is a sign of a normal or abnormal process, or a condition or disease"

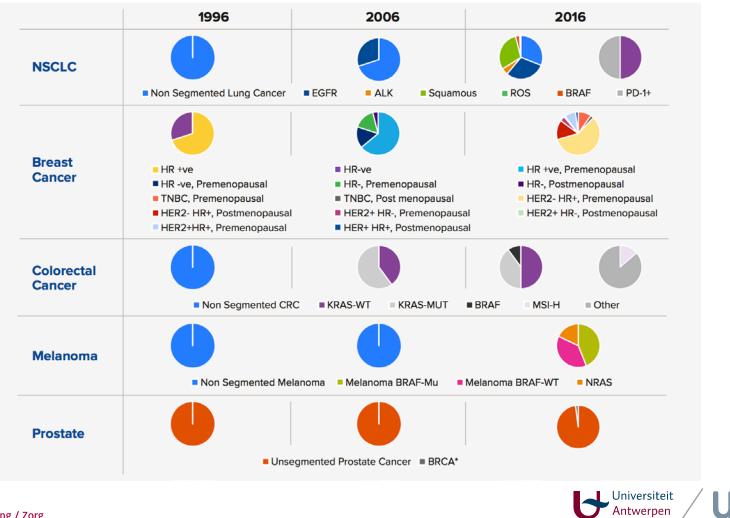


# State of the Science in Biomarker Research

- More than 40,000 papers on cancer biomarkers each year
- Around 4000–5000 on biomarkers for early detection, diagnosis and prognosis
- 99% claims >90% sensitivity and specificity
- But, very few are supported by evidence sufficient for regulatory approval
  - Rigorous standards for validation of clinical relevance in appropriate populations (i.e., in detecting preclinical disease, predicting progression/extent of disease)

### **Cancer treatment over the last past 20 years**

#### Percent of Biomarker-Based Segmentation in Selected Tumor



Kennis / Ervaring / Zorg

Global Oncology Trends 2017: Advances, Complexity, and Cost. QuintilesIMS Institute. June 2017.

## The Hallmarks of a Precision-Oncology Study



Kennis / Ervaring / Zorg

Hyman et al, Cell. 2017 Feb 9;168(4):584-599

UZA

# **Biomarker Development**

#### **Clinical validity:**

The test result shows an association with a clinical outcome of interest.

### Analytical validity:

The test's performance is established to be accurate, reliable, and reproducible.

### **Clinical utility:**

Use of the test results in a favorable benefit to risk ratio for the patient



# **Cancer Biomarkers: Missing the Mark**

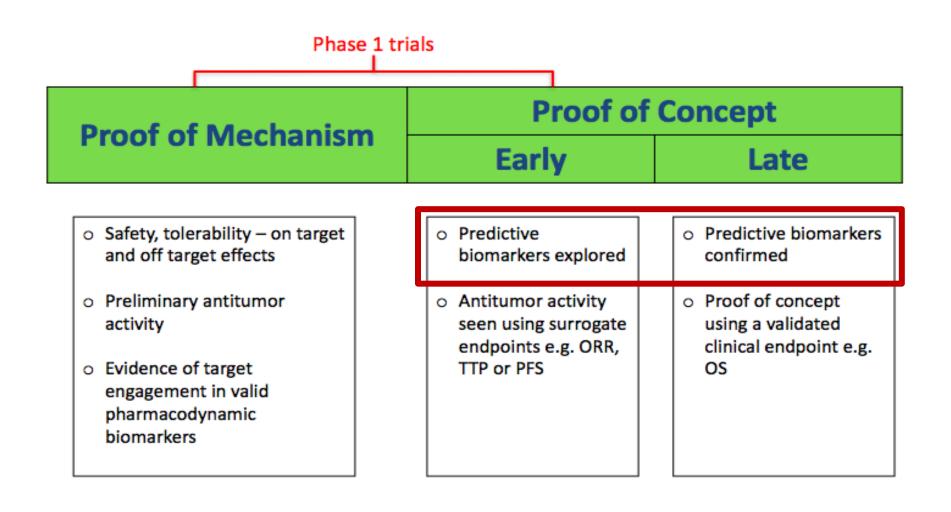
- Biology of early disease not fully explored
- Differences in analytical techniques
- Differences in statistical methods (study designs)
- Unintentional selective reporting
- Incomplete protocol reporting
- Lack of appropriate specimens and reagents
- Variations in interpretation
- Bias, chance and overfitting
- Lack of appropriate controls
- Need for additional knowledge in translation of laboratory tests into clinical tests
- Need for more collaboration

## **Phases of Biomarker Discovery and Validation**

Preclinical Exploratory	PHASE 1	Promising directions identified			
Clinical Assay and Validation	PHASE 2	Clinical assay detects established disease			
Retrospective Longitudinal	PHASE 3	Biomarker detects preclinical disease and a "screen positive" rule defined			
Prospective Screening	PHASE 4	Extent and characteristics of disease detected by the test and the false referral rate are identified			
Cancer Control	PHASE 5	Impact of screening on reducing burden of disease on population is quantified			

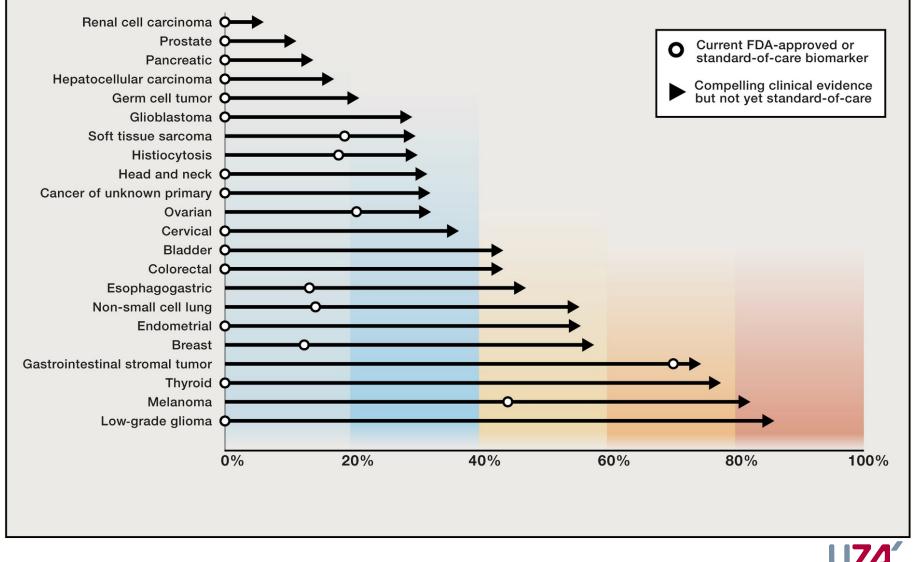


# The Current Drug Development Paradigm



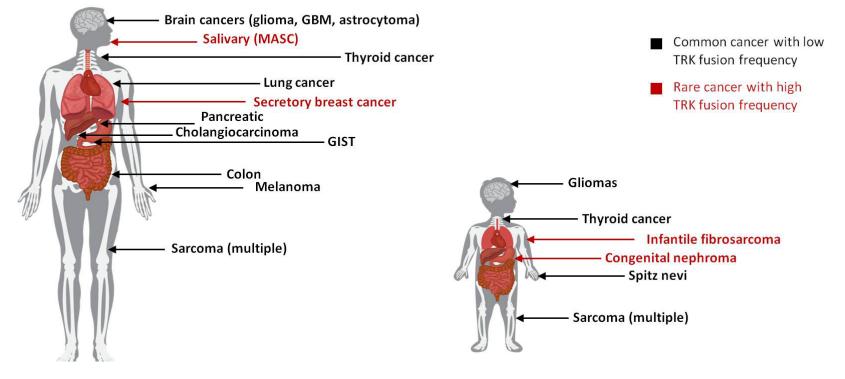


# Druggable Alterations in Oncology Today and in the Near Future



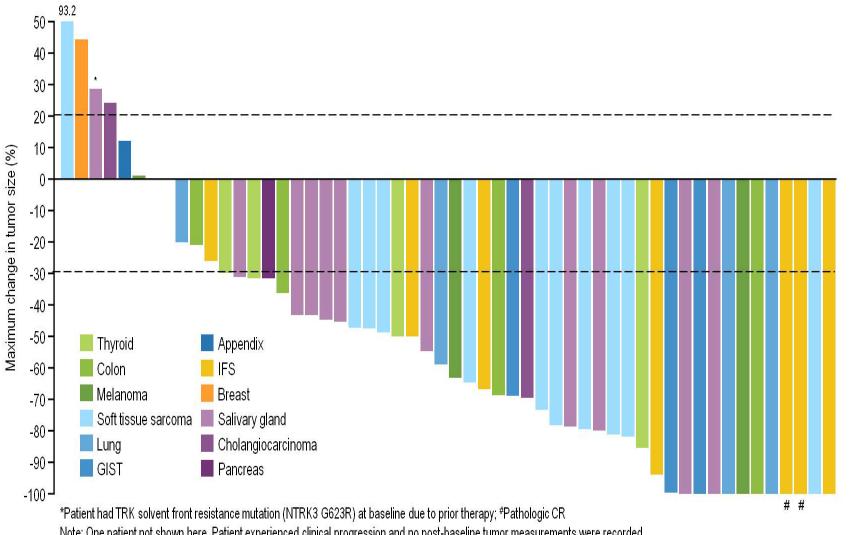
Hyman et al, Cell. 2017 Feb 9;168(4):584-599

# TRK fusions found in diverse cancer histologies



Presented By David Hyman at 2017 ASCO Annual Meeting

# **NTRK Inhibitor** Efficacy regardless of tumor type



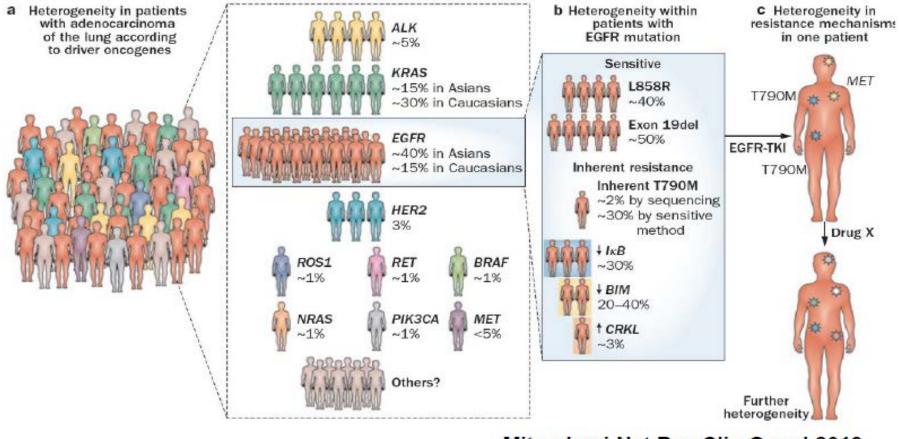
Note: One patient not shown here. Patient experienced clinical progression and no post-baseline tumor measurements were recorded.

Presented By David Hyman at 2017 ASCO Annual Meeting

Universiteit Antwerpen

# The efficacy of target therapy is affected by...

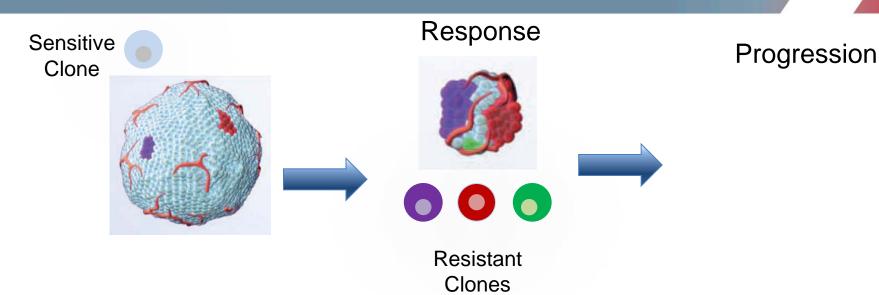
# **TUMOR HETEROGENEITY**



Mitsudomi Nat Rev Clin Oncol 2013

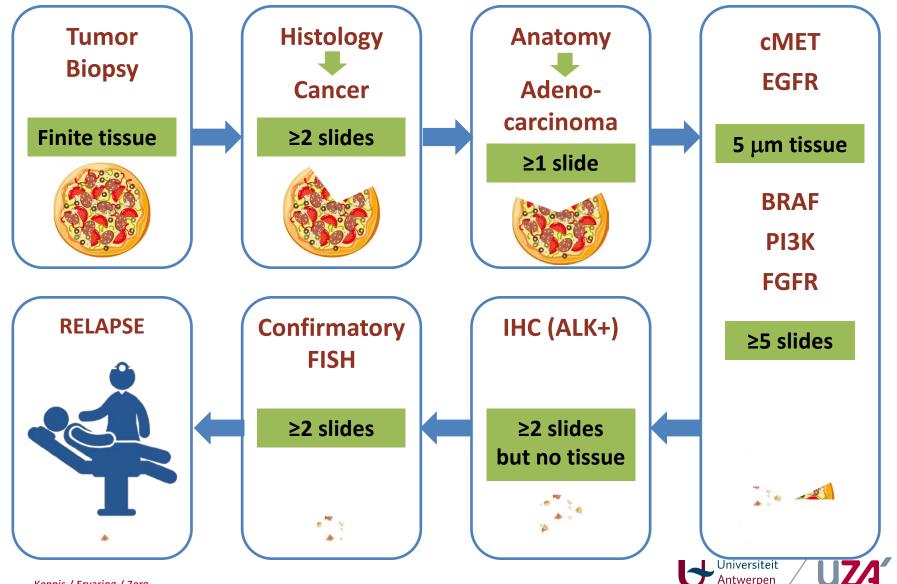


### Molecular Issues regarding T790M



- **T790M-positive and T790-wild-type clones may coexist** in some cancers with acquired resistance to initial EGFR TKIs
- Concept of cancer's "loss" of T790M suggests that the original lesion, although testing "positive" for T790M, may have contained both T790M-positive and T790—wild-type clones
- Spatial heterogeneity indicates inter-/intratumor differences at the genomic, epigenetic, and proteomic levels, whereas temporal heterogeneity reflects dynamic tumor evolution over time

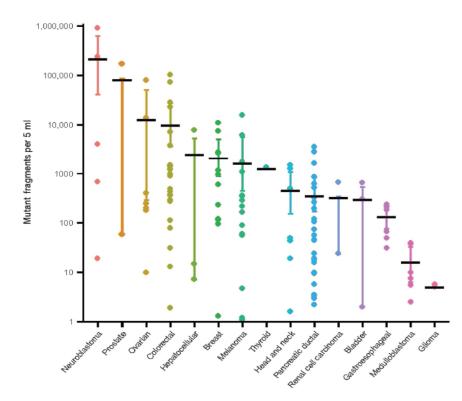
## **Multiple Tests Require Large Tissue Volume**

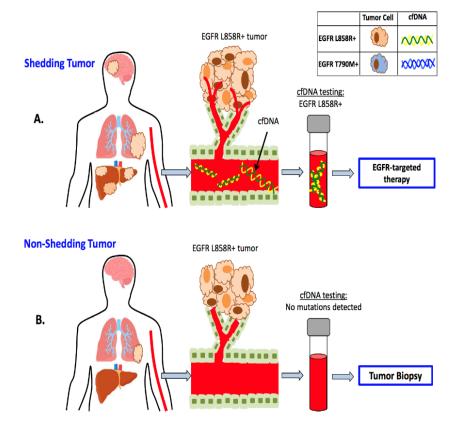


Kennis / Ervaring / Zorg

# Liquid biopsy: ctDNA

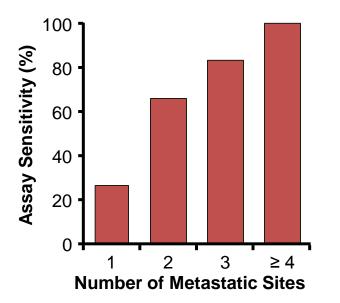
# Does ctDNA concentration is the same among patients with the same tumor?



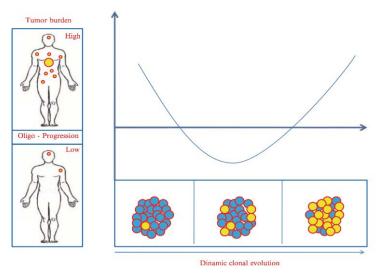


Sacher, Komatsubara, Oxnard J Thorac Oncol. 2017 Sep;12(9):1344-1356

Kennis / Ervaring / Zorg Bettegowda et al., Sci Trans Med, 2014 Sensitivity of Plasma ddPCR Higher in Pts With Metastases



Correlation between tumor burden (*y*-axis) and dynamic clonal evolution of the tumor



Increasing number of metastatic sites (P = .001) and presence of bone (P = .007), hepatic (P = .001) metastases significantly associated with assay sensitivity

UZA'

Kennis / Ervaring / Zorg

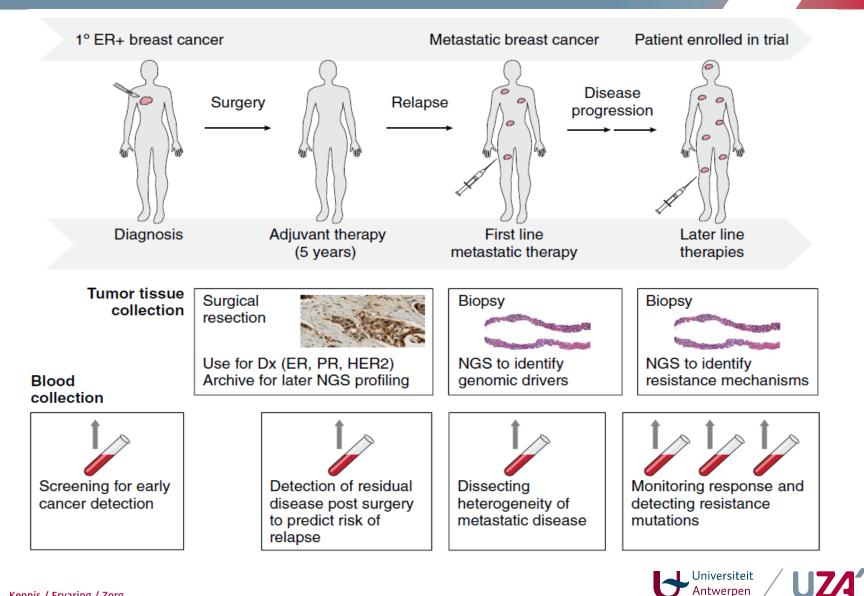
Sacher AG, et al. JAMA Oncol. 2016

## Special considerations...





## The Role of Next-Generation Sequencing in Enabling **Personalized Oncology Therapy**



Kennis / Ervaring / Zorg

Cummings et al, Clin Transl Sci (2016) 9, 283-292

#### Point Mutations - Complete\* or Critical Exon Coverage in 73 Genes

AKT1	ALK	APC	AR	ARAF	ARID1A	ATM	BRAF	BRCA1	BRCA2
CCND1	CCND2	CCNE1	CDH1	CDK4	CDK6	CDKN2A	CDKN2B	CTNNB1	EGFR
ERBB2	ESR1	EZH2	FBXW7	FGFR1	FGFR2	FGFR3	GATA3	GNA11	GNAQ
GNAS	HNF1A	HRAS	IDH1	IDH2	JAK2	JAK3	ΚΙΤ	KRAS	MAP2K1
MAP2K2	MET	MLH1	MPL	МҮС	NF1	NFE2L2	NOTCH1	NPM1	NRAS
NTRK1	PDGFRA	PIK3CA	PTEN	PTPN11	RAF1	RB1	RET	RHEB	RHOA
RIT1	ROS1	SMAD4	SMO	SRC	STK11	TERT	TP53	TSC1	VHL

#### AMPLIFICATIONS

AR	BRAF	CCND1	CCND2	CCNE1	CDK4	CDK6	EGFR	ERBB2
FGFR1	FGFR2	KIT	KRAS	MET	MYC	PDGFRA	PIK3CA	RAF1

#### **FUSIONS**

ALK FGFR2	FGFR3	RET	ROS1	NTRK1
-----------	-------	-----	------	-------

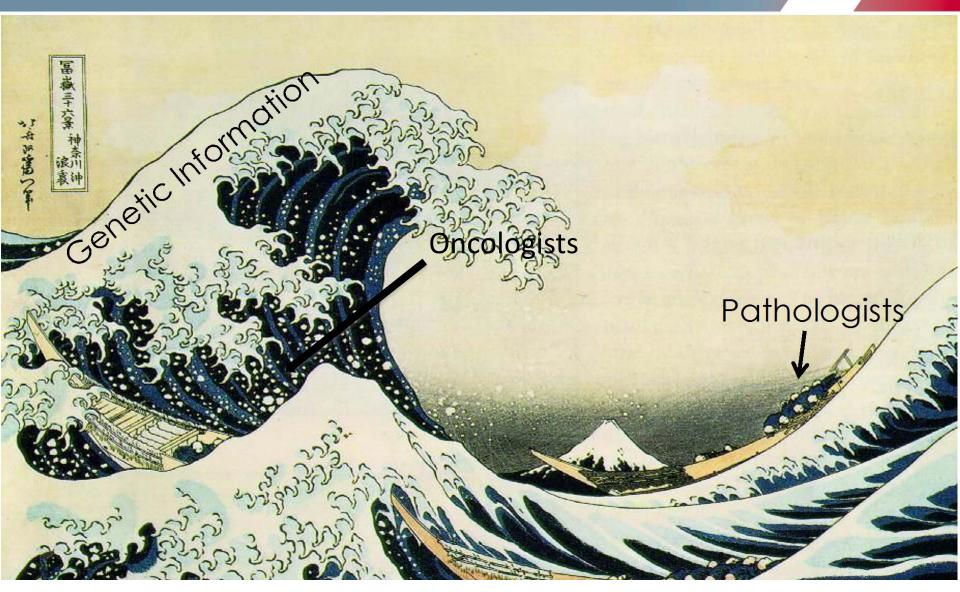
#### INDELS

EGFR exons 19/20 ERBB2 exons 19/20 MET exon 14 skipping





## Data Tsunami



- Mountain: number of mutations in a gene is very high. Any reasonable statistic will indicate that the gene is a driver
- Hill: few mutations.





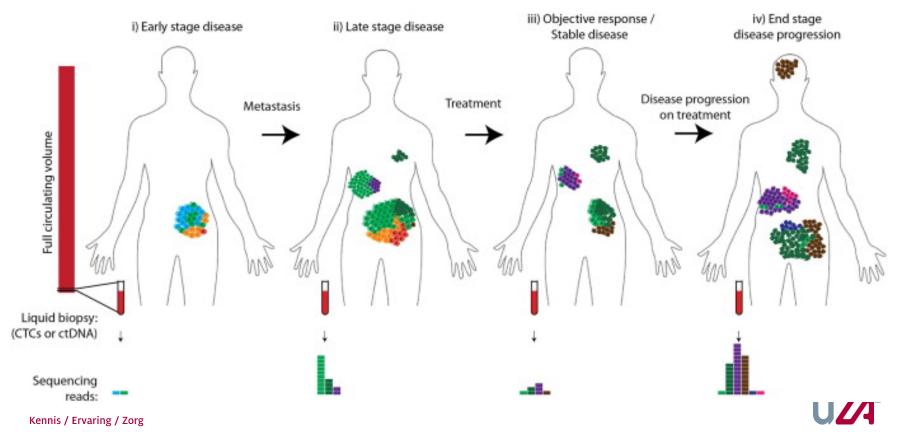




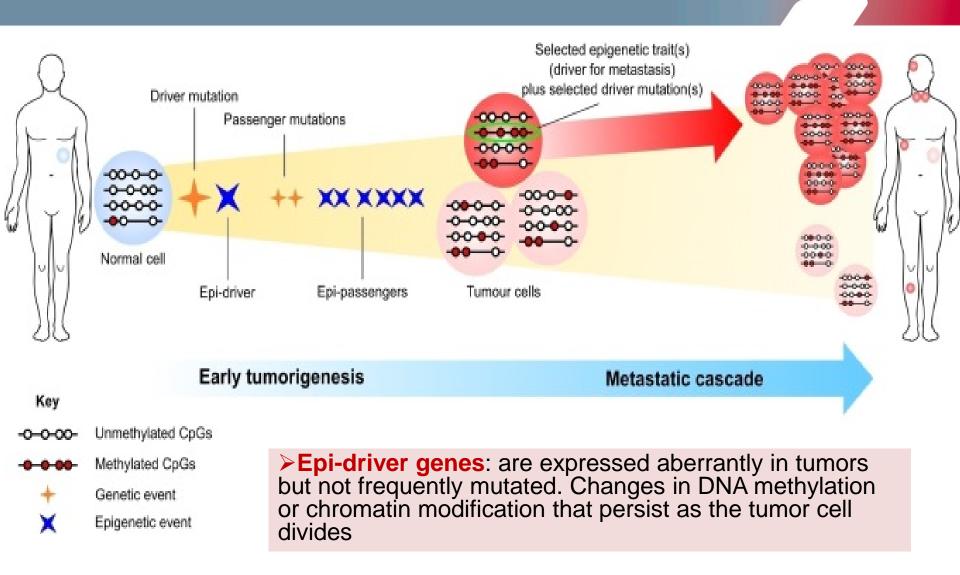


## **DRIVER MUTATIONS**

- Passenger mutations can transform into driver mutations ("latent drivers" or "mini-drivers")
- In the context of resistant and/or recurrent disease.



R. Burrell, C. Swanton Mollecular Oncology. 2014



Chatterjee, E.J. Rodger, M. Eccles. Seminars in Cancer Biology, 2017

UZ/



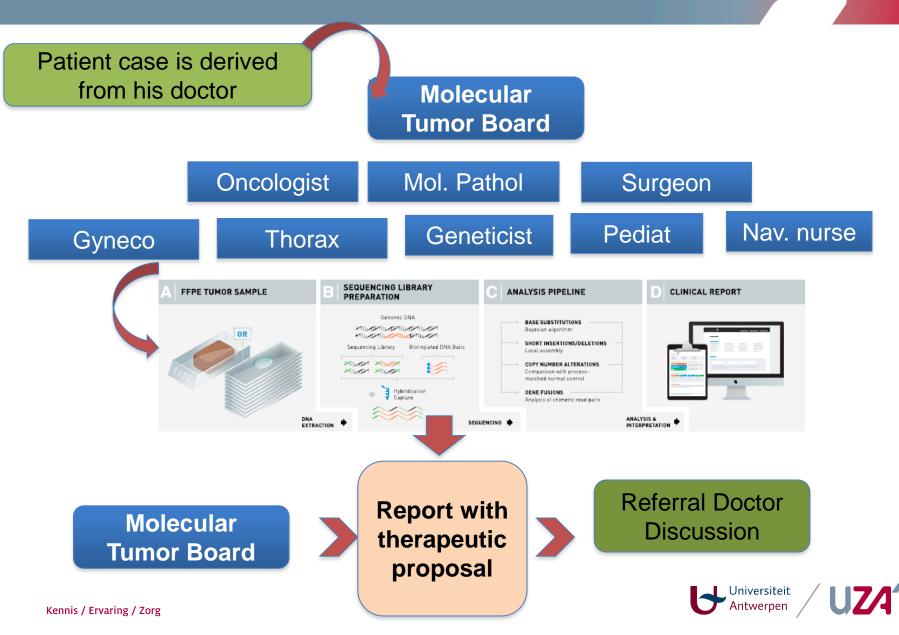


### Multidisciplinary Molecular Tumour Board: a tool to improve Clinical Practice and selection accrual for Clinical Trials in Cancer Patients

Christian Rolfo, Paolo Manca, Andreia Coelho, Jose Ferri, Peter Van Dam, Amelie Dendooven, Christine Weyn, Marika Rasschaert, Lucas Van Houtven, Xuan Bich Trinh, Jan Van Meerbeeck, Roberto Salgado, Marc PeetersPatrick Pauwels

On behalf of Molecular Tumour Board of Antwerp University Hospital, Edegem, Belgium.

# **Molecular Tumor Board**



# **MSK Levels of Evidence**

Level **R1** 

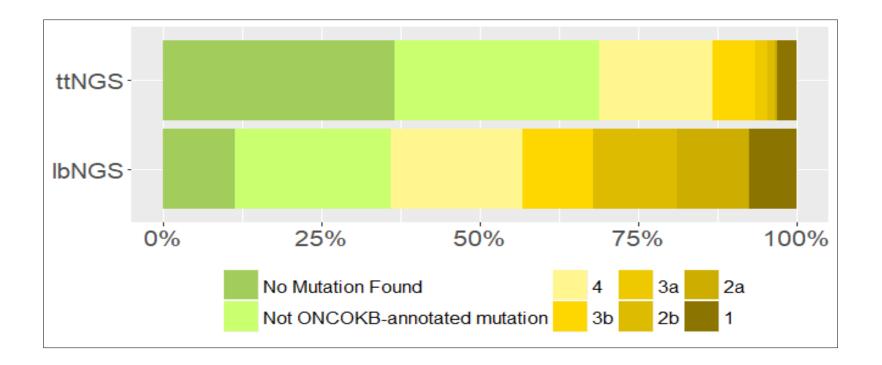
varing / Zorg



Level		_	
	FDA-recognized biomarker predictive of response to an FDA- approved drug in this indication		Standard Therapeutic Implications *Includes biomarkers
Level			that are recommended
2A	Standard of care biomarker predictive of response to an FDA- approved drug in this indication*		as standard of care by the NCCN or other expert panels
Level 2B	<b>Standard of care</b> biomarker predictive of response to an <b>FDA</b> - <b>approved</b> drug <b>in another indication</b> , <i>but not standard of care</i> <i>for this indication</i>		but not necessarily FDA-recognized for a particular indication
Level			Investigational
3A	<b>Compelling clinical evidence</b> supports the biomarker as being predictive of response to a drug in this indication, but neither biomarker nor drug are standard of care		Investigational Therapeutic Implications possibly directed
Level 3B	Compalling clinical avidance supports the biomarker as being		to clinical trials
	<b>Compelling clinical evidence</b> supports the biomarker as being predictive of response to a <i>drug</i> <b>in another indication</b> , <i>but neither biomarker nor drug are standard of care</i>		Hypothetical
Level			Therapeutic Implications
4	<b>Compelling biological evidence</b> supports the biomarker as being predictive of response to a drug, but neither biomarker nor drug are standard of care		based on preliminary, non- clincial data

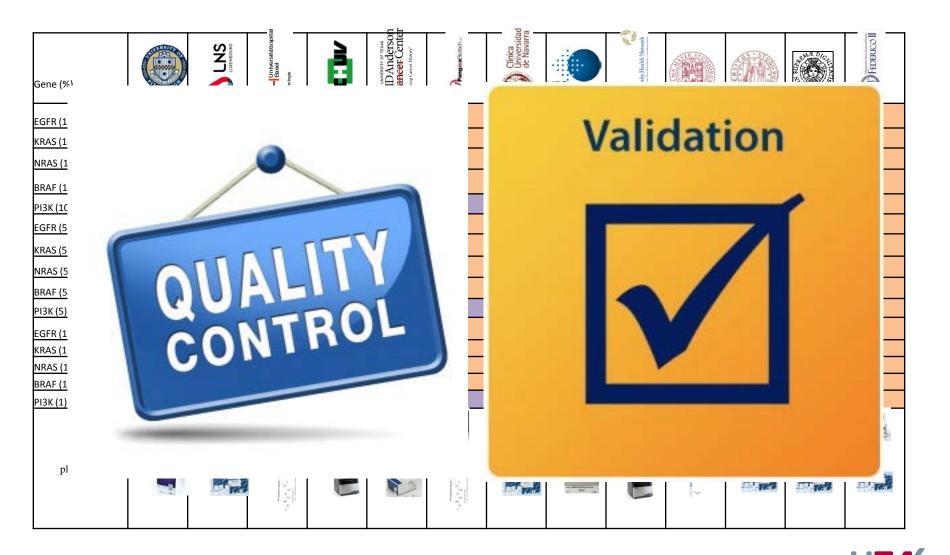
Standard of care biomarker predictive of resistance to an FDA-approved drug in this indication

Standard Therapeutic Universiteit Antwerpen





### Everybody can do it?



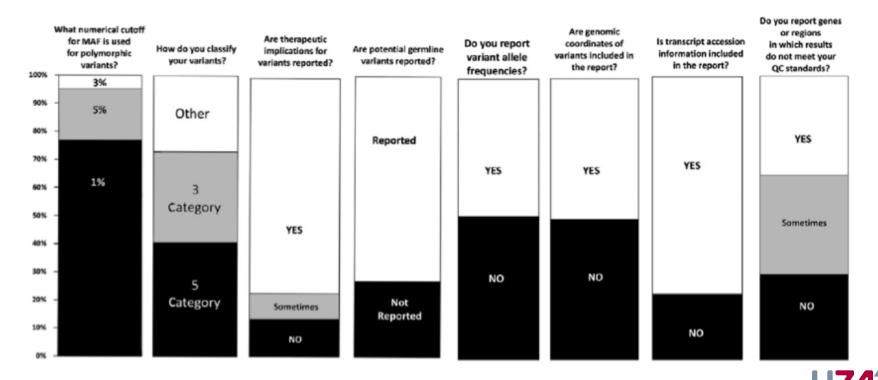
Malapelle et al. Cancer Cytopathology 2017

#### Standards and Guidelines for the Interpretation and Reporting of Sequence Variants in Cancer

#### A Joint Consensus Recommendation of the Association for Molecular Pathology, American Society of Clinical Oncology, and College of American Pathologists

CrossMaric

Marilyn M. Li,\*<sup>1</sup> Michael Datto,\*<sup>1</sup> Eric J. Duncavage,\*<sup>1</sup> Shashikant Kulkami,\*<sup>9</sup> Neal I. Lindeman,\*<sup>1</sup> Somak Roy,\*\*\*\* Apostolia M. Tsimberidou,\*<sup>11</sup> Cindy L. Vnencak-Jones,\*<sup>11</sup> Daynna J. Wolff,\*<sup>10</sup> Anas Younes,\*<sup>11</sup> and Marina N. Nikiforova\*\*\*\*

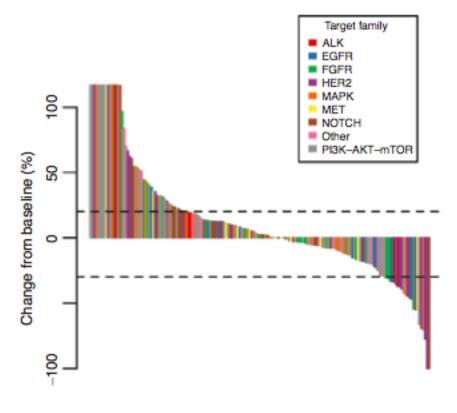


3

2



# High-Throughput Genomics and Clinical Outcome in Hard-to-Treat Advanced Cancers:



high-throughput genomics could improve outcomes in a subset of patients with hard-to-treat cancers. Although these results are encouraging, only 7% of the successfully screened patients benefited from this approach

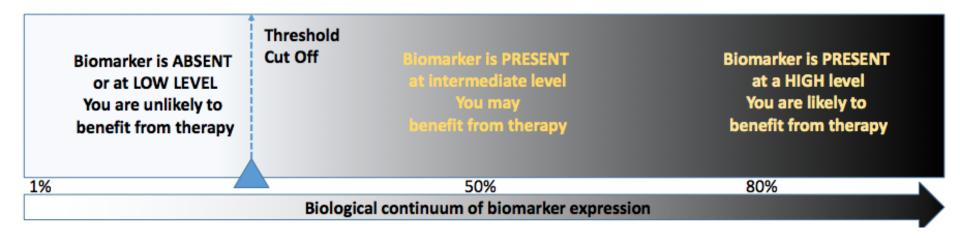


## Immunotherapy in Cancer





Biomarker is ABSENT You are unlikely to benefit from therapy Biomarker is PRESENT You are likely to benefit from therapy





# PD-L1 & the Meta-analysis



Systematic Review and Meta-Analysis



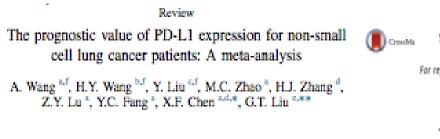
# treatm

Cuihua Wang, PD-L1 expression as predictive biomarker in patients with NSCLC: a pooled analysis



Francesco Passiglia<sup>1,\*</sup>, Giuseppe Bronte<sup>1,\*</sup>, Viviana Bazan<sup>1,\*</sup>, Clara Natoli<sup>2</sup>, Sergio Rizzo<sup>1</sup>, Antonio Galvano<sup>1</sup>, Angela Listì<sup>1</sup>, Giuseppe Cicero<sup>1</sup>, Christian Rolfo<sup>3</sup>, Daniele Santini<sup>4</sup>, Antonio Russo<sup>1</sup>





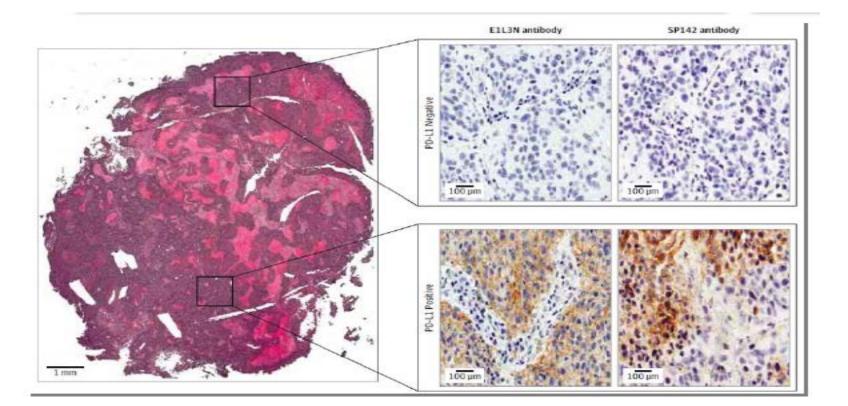
Systematic Review

For reprint orders, please contact: reprints@futuremedicine.com

The role of PD-L1 expression as a predictive biomarker in advanced non-small-cell lung cancer: a network meta-analysis

Oncotarget, Vol. 7, No. 15

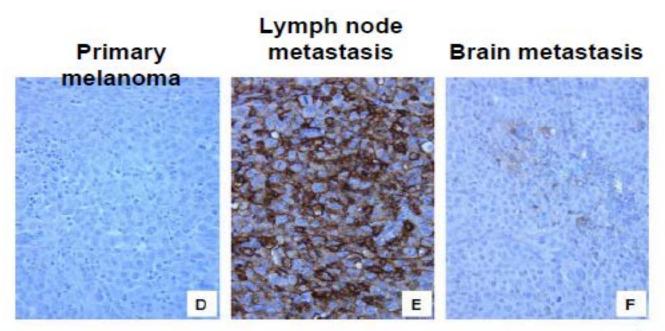




# Using the PD-L1 IHC 28-8 pharmDx, 6% discordance was observed in 30 cases of multisamples per case<sup>5</sup>



### **PDL-1** status



Adapted from Madore et al.<sup>2</sup>

# PD-L1 discordance observed within same patient



# PD-L1 as a predictive immune biomarker: assays, sample collection and analysis in NSCLC studies

	Pembrolizumab Merck	Nivolumab Bristol-Myers Squibb	Atezolizumab Roche/Genentech	Durvalumab AstraZeneca	Avelumab Pfizer/Merck Serono
kesse TD-CI	<ul> <li>Prototype or clinical trial IHC assay (22C3 Ab)<sup>1,2</sup></li> </ul>	<ul> <li>Dako automated IHC assay</li> <li>(28-8 Ab)<sup>3,4</sup></li> </ul>	Central laboratory IHC assay     Ventana PD-L1 (SP142)	<ul> <li>Ventana automated IHC (BenchMark ULTRA using Ventana PD-L1 (SP263) clone)<sup>7,8</sup></li> </ul>	Dako assay     Clone not known
	<ul> <li>Surface expression of PD-L1 on tumour specimen<sup>1,2</sup></li> </ul>	<ul> <li>Surface expression of PD-L1 on tumour cells<sup>3,4</sup></li> </ul>	<ul> <li>Surface expression of PD-L1 on TILs or tumour cells</li> </ul>	<ul> <li>Surface expression of PD-L1 on tumour cells<sup>7,8</sup></li> </ul>	<ul> <li>Surface expression of PD-L1 on tumour cells</li> </ul>
Sample source and collection	Ph I: Fresh or archival tissue <sup>1,2</sup>	<ul> <li>Archival or fresh tissue<sup>3,4</sup></li> </ul>	Archival or fresh tissue	recent or archival samples	Unknown
Definition of positivity <sup>1</sup> Samp	<ul> <li>HC staining:</li> <li>Strong vs weak expression<sup>1,2</sup></li> <li>PD-L1 expression required for NSCLC for enrolment<sup>1</sup></li> <li>Note that one arm of KEYNOTE 001 trial requires PD-L1: tumours<sup>1</sup></li> </ul>	<ul> <li>IHC staining:</li> <li>Strong vs weak expression<sup>3,4</sup></li> <li>Patients not restricted by PD-L1 status in 2nd- &amp; 3rd-line</li> <li>Ph III 1st-line trial in PD-L1+<sup>5</sup></li> </ul>	<ul> <li>IHC staining intensity         <ul> <li>(TC: 0, 1, 2, 3):</li> <li>IHC 3 (≥50% PD-L1<sup>+</sup>)</li> <li>IHC 2,3 (≥5% PD-L1<sup>+</sup>)</li> <li>IHC 1,2,3 (≥1% PD-L1<sup>-</sup>)</li> <li>IHC 0,1,2,3 (all patients with evaluable status)<sup>5,6</sup></li> <li>PD-L1 expression required for NSCLC for enrolment in Ph II trials</li> </ul> </li> </ul>	<ul> <li>IHC staining intensity:</li> <li>proportion of cell staining regardless of intensity</li> </ul>	IHC staining intensity: • Not presented to date
	Tumour PD-L1 expression: <sup>1,2</sup> <ul> <li>≥50% PD-L1* cut-off: 32% (41/129)</li> <li>1-49% PD-L1* cut-off: 37% (48/129)</li> </ul>	Tumour PD-L1 expression: • 1% PD-L1 + cut off • 5% PD-L1 <sup>+</sup> cut-off: 59% (10/17) <sup>3</sup> • 5% PD-L1 <sup>+</sup> cut-off: 49% (33/68) <sup>4</sup> • 10% PD-L1 + cut off	IC: TIL PD-L1 expression: • IHC 3 (≥10% PD-L1*): 11% (6/53) • PD-L1 low (IHC 1, 0): 62% (33/53)	Tumour PD-L1 expression:7 • PD-L1 + cut off 25% • PD-L1*: 34% (20/58) • PD-L1*: 50% (29/58)	Tumour PD-L1 expression (all doses): • PD-L1 + cut off 1% • PD-L1*: 34% (20/58) • PD-L1*: 50% (29/58)

\*Definition of PD-L1 positivity differs between assay methodologies

1. Garon EB et al. ESMO 2014. Abs LBA43; 2. Rizvi NA et al. ASCO 2014. Abs 8007; 3. Gettinger S et al. ASCO 2014. Abs 8024; 4. Brahmer JR et al. ASCO 2014. Abs 8112;

5. Rizvi NA et al. ASCO 2014. Abs TPS8123; 6. Soria J-C et al. ESMO 2014. Abs 1322P; 7. Brahmer JR et al. ASCO 2014. Abs 8021; 8. Segal NH et al. ASCO 2014. Abs 3002



### The IASLC Blue Print Study

- 39 NSCLC tumor stained with four PD-L1 assays
- Independent review by three expert pathologists
- Similar PD-L1 expression for three assays
- 1. Blueprint phase 2A involving real-life clinical lung cancer samples and 25 pathologists largely affirms the results of Blueprint phase 1
- 2. 22C3, 28-8 and SP263 are comparable, SP142 detects less, while 73-10 stains more PD-L1 positive tumor cells
- 3. PD-L1 scoring on digital images and glass slides show comparable reliability

28-8	36/38 (94.7%)	38/38 (100%)	31/38 (81.6%)	33/38 (86.8%)
SP142	24/38 (63.2%)	24/38 (63.2%)	38/38 (100%)	25/38 (65.8%)
SP263	34/38 (89.5%)	34/38 (89.5%)	33/38 (86.8%)	38/38 (100%)

\* Tumor cell (TC) and immune cell (IC) scoring ranges are described in chapter 6. TC0 is defined as less than 1% of tumor cells expressing PD-L1, TC1 is 1% to 5% expression, TC2 is 5% to 50% expression, and TC3 is greater than 50% expression. Table adapted from Hirsch FR et al, J Thorac Oncol. 2017;12(2):208-222.



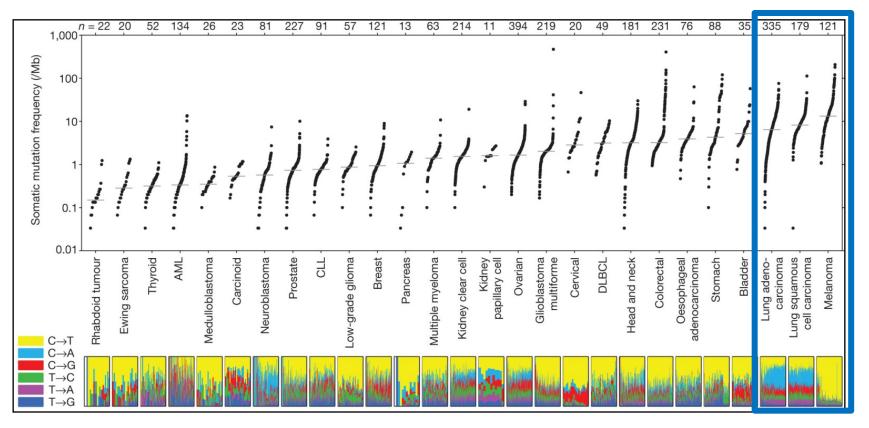
**Other biomarkers to better** 

# select our patients?





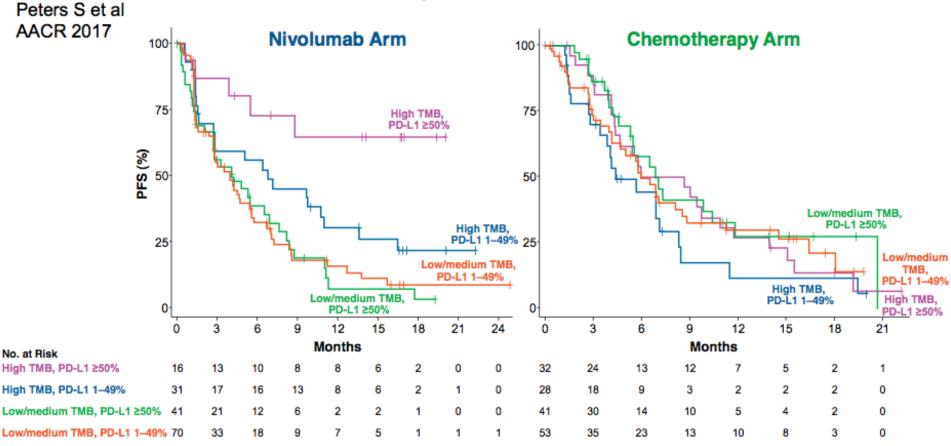
Somatic mutation frequencies observed in exomes from 3,083 tumour–normal pairs.





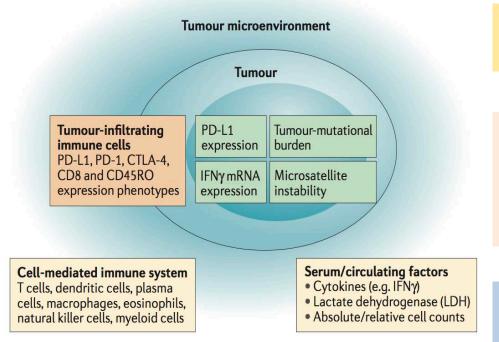
# **Mutational Tumor Burden**

#### PFS by TMB Subgroup and PD-L1 Expression CheckMate 026 TMB Analysis: Nivolumab in First-line NSCLC



Hontwerpen UNIX

# Liquid Biopsies in Immunotherapy



#### **Unmeet Medical Need:**

#### **Validated Biomarkers in Blood!**

#### Potential Utility of Liquid Biopsy in Immunotherapy

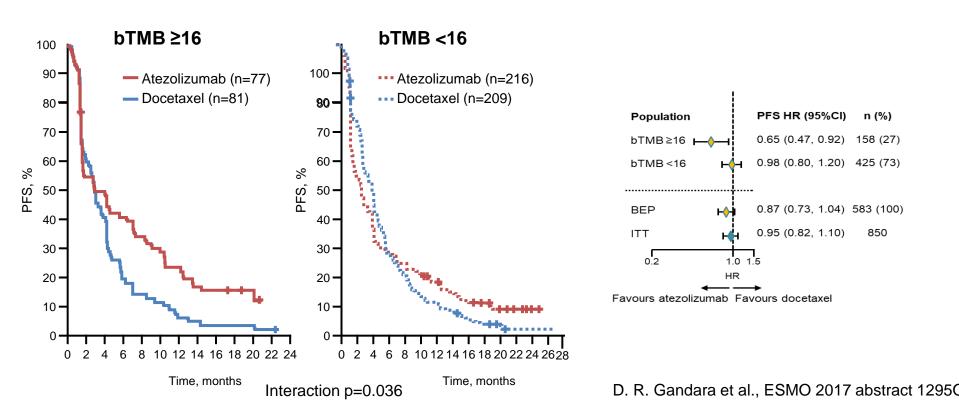
- Diagnostic
- Prognostic
- Predictive of Response
- Monitoring
- •Mechanisms if Resistance

#### **Current tools:**

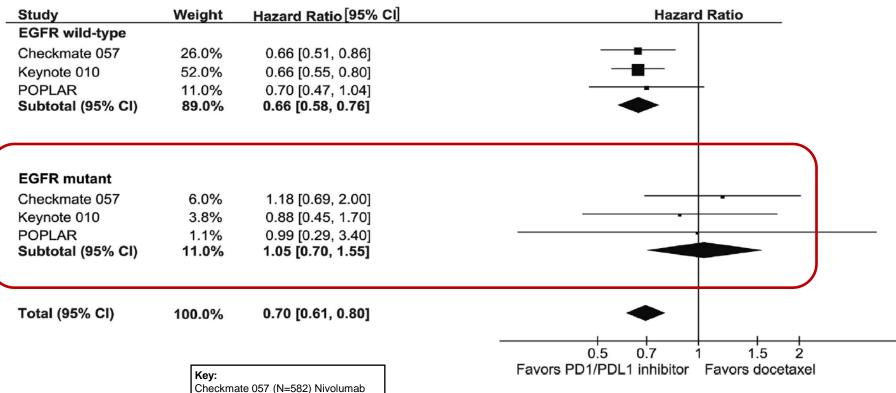
- Calculation of circulating TMB
- Detection of bPDL1
- Alellic Fraction Variation Dynamic



**bMTB** 

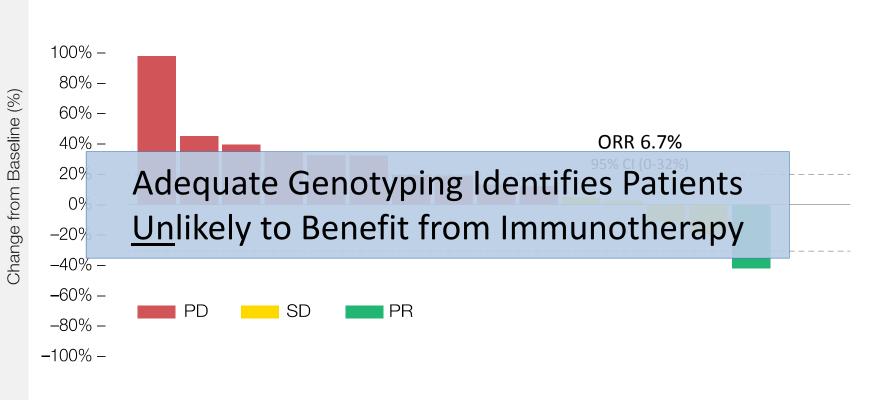






Checkmate 057 (N=582) Nivolumab Keynote 010 (N=1034) Pembrozulimab POPLAR (N=287) Atezolizumab

Universiteit UZA



Note: PD defined as  $\geq$  20% growth or appearance of new lesions



Sabari et al, J Clin Oncol 35, 2017 (suppl; abstr 8512)

**BRAF & IO** 

 $\bigcirc$ 



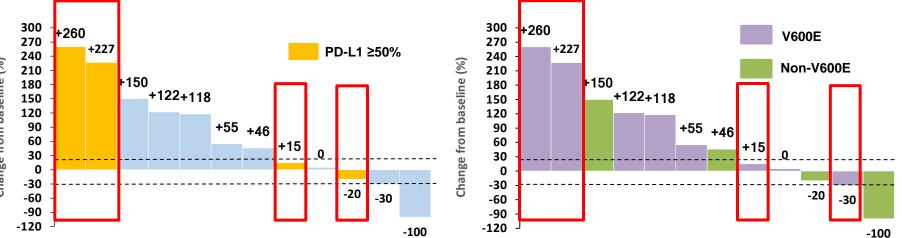
#### **Objective response with ICPi**

# **BRAF MUT NSCLC**

- n-15 (V600E, n-8; non-V600E, n-7)
- Nivolumab, n-10; pembrolizumab, n-5 ٠
- ICPi 1<sup>st</sup>-line, n-4 (V600E, n-1; non-V600E, n-3); 2<sup>nd</sup>-line, n-9 (V600E, n-5; non-V600E, n-4); 3<sup>rd</sup>-line, n-2 (V600E, n-2)



Abbreviations: ICPi - immune check-point inhibitors s, ORR - objective response rate.





# How to integrate biomarkers in

# clinical trials design?





- Enhanced genomic screening efficiency
- Inclusion of wide array of molecular subtypes
- Use of common genomic platform or diagnostic tests
- Screening for variants of <u>multiple</u> genomic targets in each tumor sample in each tumor sample (requires sufficient tumor material)
- **↑** willingness of patients and HCPs to participate
- Deletion/insertion of new subprotocol by amendment instead of completely new protocol developmer
- More rapid clinical development





### **Basket Trials: Pros and Cons**

#### Prerequisites:

- **1.** Drug must sufficiently inhibit target
- 2. Tumor must depend on target



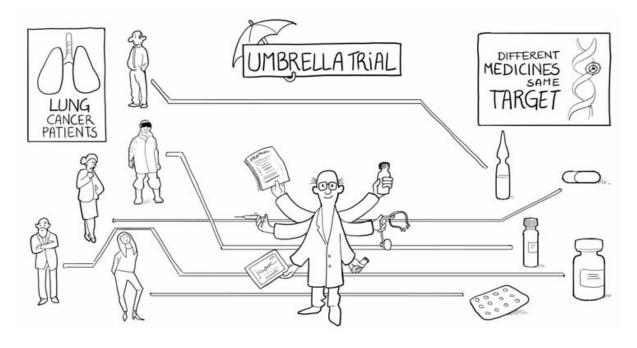
- <u>Benefits</u>:
  - Access to trial for patients with rare tumors (bust must have respective molecular marker)
  - Testing could be done locally
  - Small cohorts (usually single arm) may suffice to detect activity
  - Quick results

#### • <u>Challenges</u>:

- Molecular variant(s) may not be the only driver of tumor
- Contextual complexities in various histologies
- · Single biomarkers may be inferior to multi-gene signature
- Structural variants may need to be complemented with functional studies
- Different tumor types have different prognoses: single primary endpoint (eg ORR) may skew results



# <u>Hypothesis</u>: The response to targeted therapy is <u>primarily</u> determined by histologic context







Renfro Ann Oncol Oct 11 2016

# **Umbrella Trials: Pros and Cons**

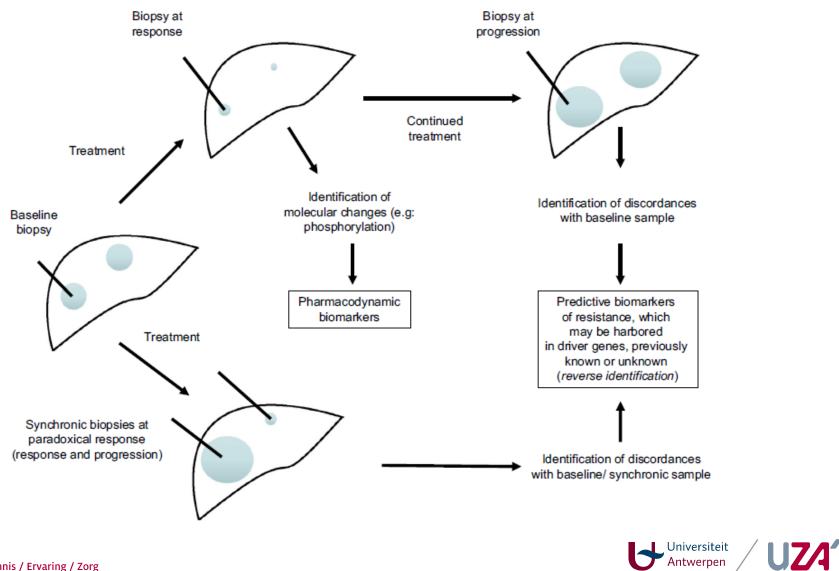
#### Benefits:

- Conclusions are specific for a given tumor type
  - Tumor heterogeneity limited to one tumor type
- For randomized substudies:
  - Potential to better understand the difference of targeted therapy vs SOC
  - Potential to differentiate between prognostic and predictive markers
  - Easier path to negotiate approval with regulatory agencies
- <u>Challenges</u>:
  - Requires:
    - Strong collaboration between academia and industry
    - Consistent marker profile , comparability of cohorts (bx, assay, Tx)

#### • Feasibility:

- Subclassification into rare populations (particularly with rare cancers to start out with)
- $\circ \rightarrow \downarrow$  speed of accrual
- Randomization requiring a larger sample size may be challenging
- Appearance of new SOCs during trial conduct changes the environment

# Design of studies exploring responses following progression or paradoxical responses



Perez Gracia et al, Cancer Treatment Reviews 53 (2017) 79-97

# Why is Discovery of Clinically Useful Biomarkers Difficult?

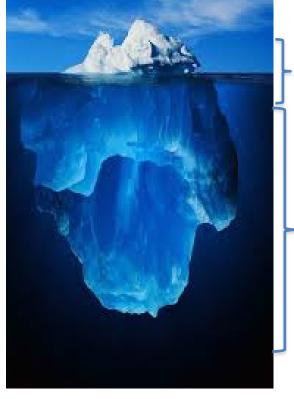
- Biology
- Need for Infrastructural Support
- Need for Collaborations Among Stakeholders
  - Basic scientists
  - Clinicians
  - Public Health Professionals
  - Informaticians and Bioinformaticians
  - Advocates
  - Funding organizations
  - Regulatory authorities

#### **Iceberg of Cancer**

Known Genetic Changes from Frankly Malignant Tumors

Unknown Genetic Changes in Preneoplastic (in situ lesion) and Neoplastic (benign or malignant conditions)





### **Project Team members**

#### Oncology – Phase I Early Clinical Trials Unit

Prof. Dr. Christian Rolfo -

Prof. Dr. Marc Peeters – head oncology and MOCA Dr. Marika Rasschaert – Dr. Katrine De Block Fellows: Dr. Helena Oliveres. Dr. Mariana Rocha Rolfo Lab:

Exosomes: Senior Dr. SimonaTaverna PhD students: Dr. Pablo Reclusa Asiain Dr. Marzia Pucci Dr. Mahafarin Maralani tFree DNA: Dr. Laura Sober – Karen Zwaenepoel Cell Lines & cMET: Dr. Nele Van Der Steen Logistics: Sam Van Gerwen, BsC Clinical Study –co: Amelie Lyessens, BsC Molecular Pathology Unit

Prof. Dr. Patrick Pauwels Dr. Amelie Dendooven

Dr. Karen Zwaenepoel

#### Tumor - Serum Bank

Dr. Annemieke De Wilde Dr. Sofie Goethals



#### **Next Generation Sequencing**

Dr. Christine Weyn – UZA Dr. Suzanne Lambin Dr. Ken Op De Beeck - UA **Database**: Dr. R. Mauceri Dr. Andreia Coelho

#### Proteomics

Prof. Inge Mertens Prof. Geert Baggerman Dr. Evelien Maes







# Dank u voor uw aandacht

Thank you for your attention



