## Current Topics in Lung Cancer Research

# **Mutations Biomarkers**

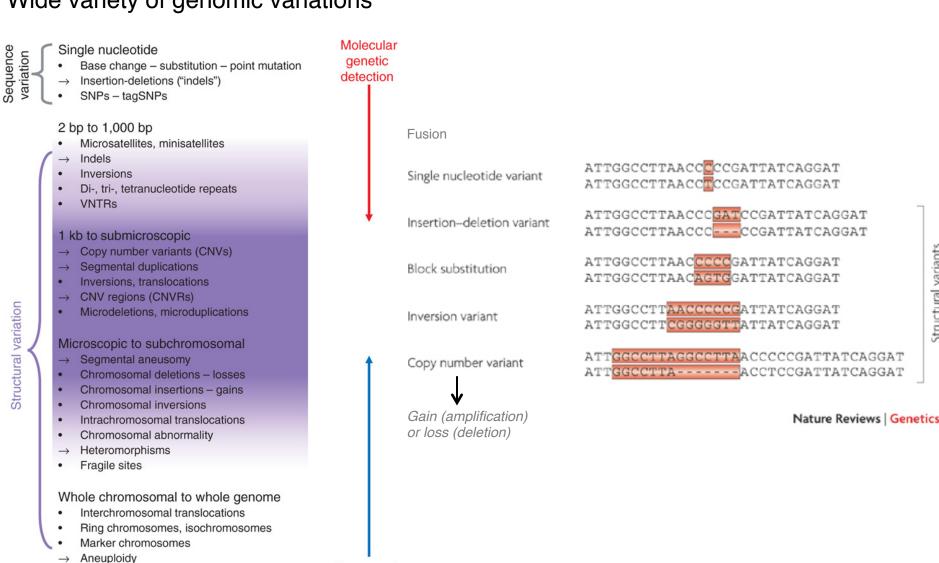
RPN532: Oncology for Scientists II

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Aneusomy

## "Mutations" in lung cancer

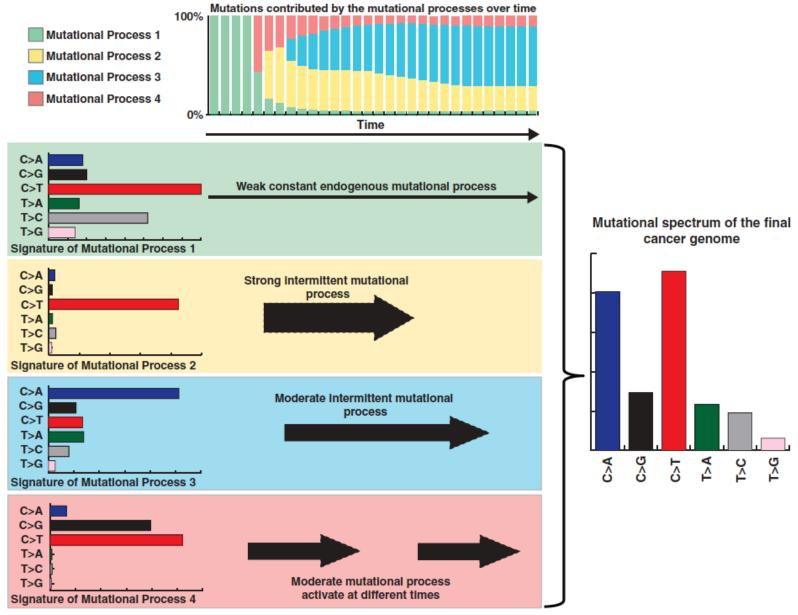
#### Wide variety of genomic variations



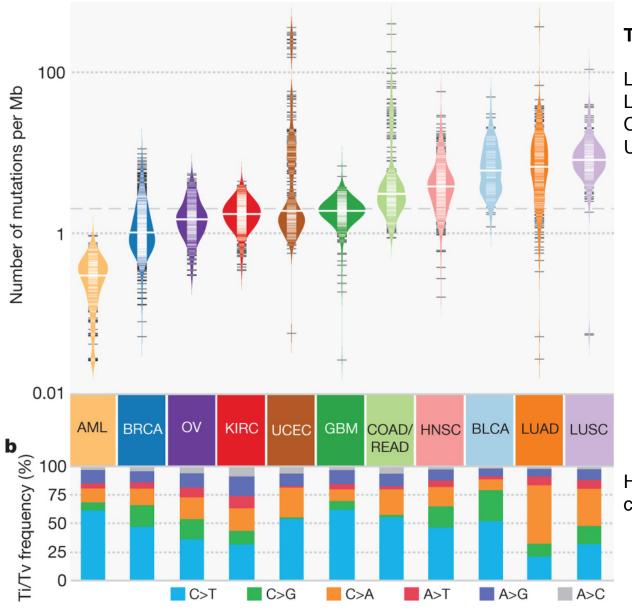
Cytogenetic

detection

## Multiple mutation processes in cancer



## **Mutations** Mutation landscape in 12 major cancers



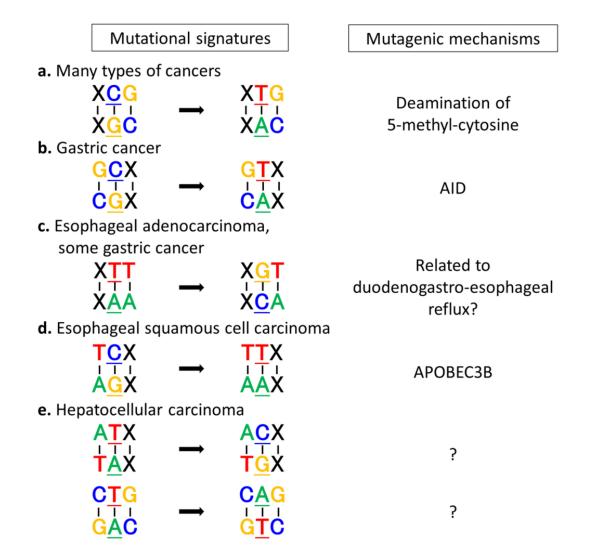
#### The Cancer Genome Atlas (TCGA)

Lung adenocarcinoma (LUAD) Lung squamous cell ca. (LUSC) Colorectal adenoca. (COAD/READ) Uterine corpus endometrial ca. (UCEC)

Higher C>A transversions in lung cancer

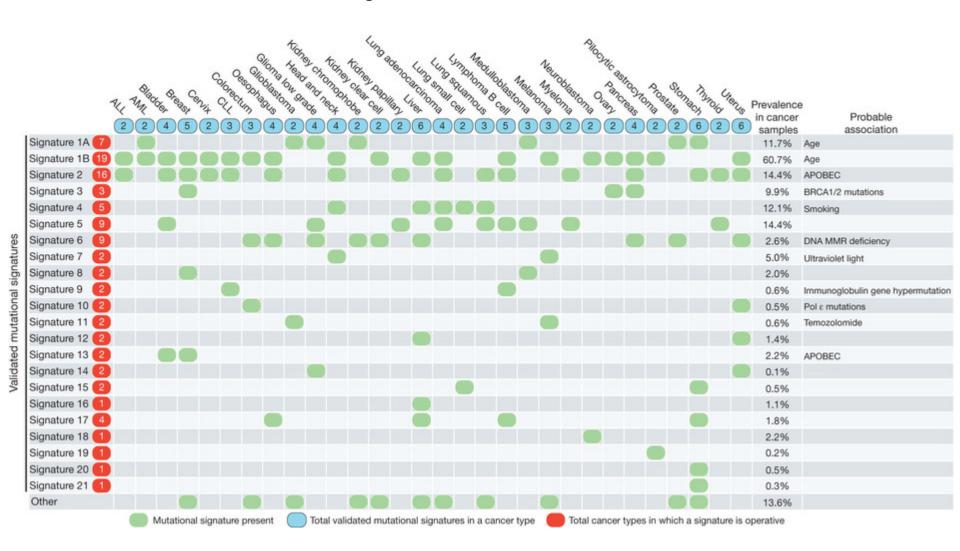
## **Mutation signatures**

#### Mutation signature = Sequence context of a mutation

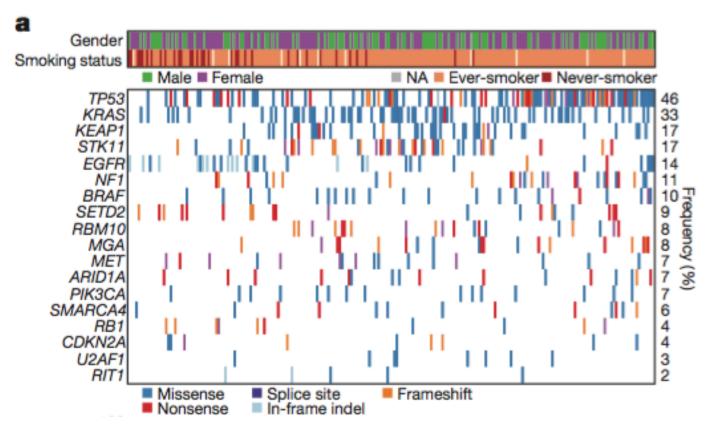


# Mutation signatures

#### Distinct mutation signatures are seen in different cancers

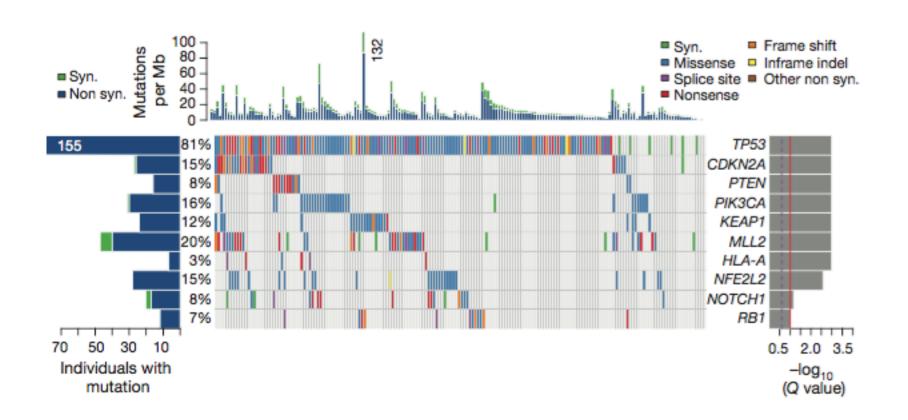


#### **Mutations Common mutations in adenocarcinoma**



TCGA project

### Common mutations in squamous cell ca.



TCGA project

# Common mutations in lung cancer

	Adenocarcinoma	SCC	SCLC
Mutations	TP53, KRAS, EGFR, NF1, BRAF, MET, RIT	TP53, CDKN2A, PIK3CA, NFE2L2, KEAP1, CUL3, PTEN, NF1, NOTCH1,2, and 3, DDR2, EGFR	TP53, RB1, EP300, CREBBP, PTEN, SLIT2
Fusions	ALK, ROS1, RET	FGFRs	
SCNAs	Gains: NKX2-1, TERT, EGFR, MET, KRAS, ERBB2, MDM2 Losses: LRP1B, PTPRD, and CDKN2A	Gains: Chr 3q 26 (SOX2, PIK3CA, TP63 etc) Losses: CDKN2A, PTEN	Gains: MYC, MYCN, MYCL1, SOX2, FGFR1, KIT Losses: Chr 3p (FHIT, FUS1, RASSF1A)
Pathway alterations	RTK/RAS/RAF mTOR JAK-STAT DNA repair Cell cycle regulation Epigenetic deregulation	Squamous differentiation Oxidative stress response PIK3CA DNA repair Cell cycle regulation Epigenetic deregulation	Cell cycle regulation, epigenetic deregulation, Hedgehog, DNA repair, axonal guidance and neuroendocrine differentiation

SCNA=somatic copy number alteration.

SCC = lung squamous cell ca. SCLC = small cell lung cancer

## Why examine mutations?

Understand cancer biology

Accurately sub-classify cancer

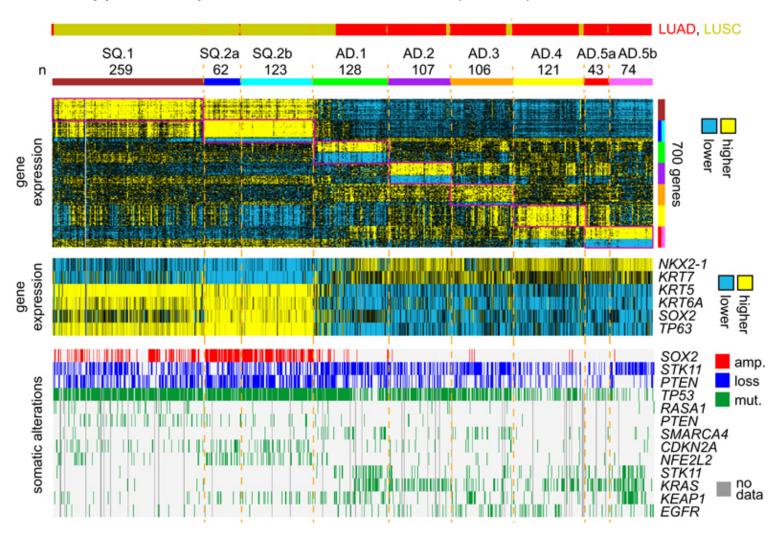
Target specific protein or pathway for drug treatment ("actionable mutation")

Design precision treatment

Utilize as biomarker

# Accurately sub-classify cancer

Mutation data combined with other molecular information identifies 6 sub-types of adenocarcinoma (LUAD) 3 sub-types of squamous cell carcinoma (LUSC)



## Design precision treatment

50 y woman with 7 lung tumors (colorectal cancer metastases)

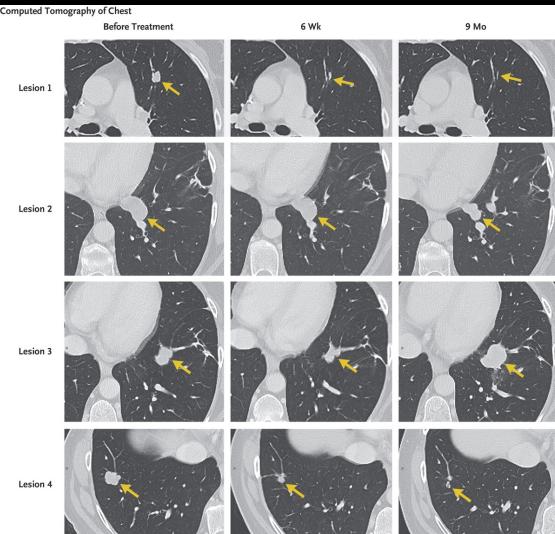
Exome DNA sequencing of 3 tumors => ~70 mutated genes

KRAS G12D mutation (27% freq.)

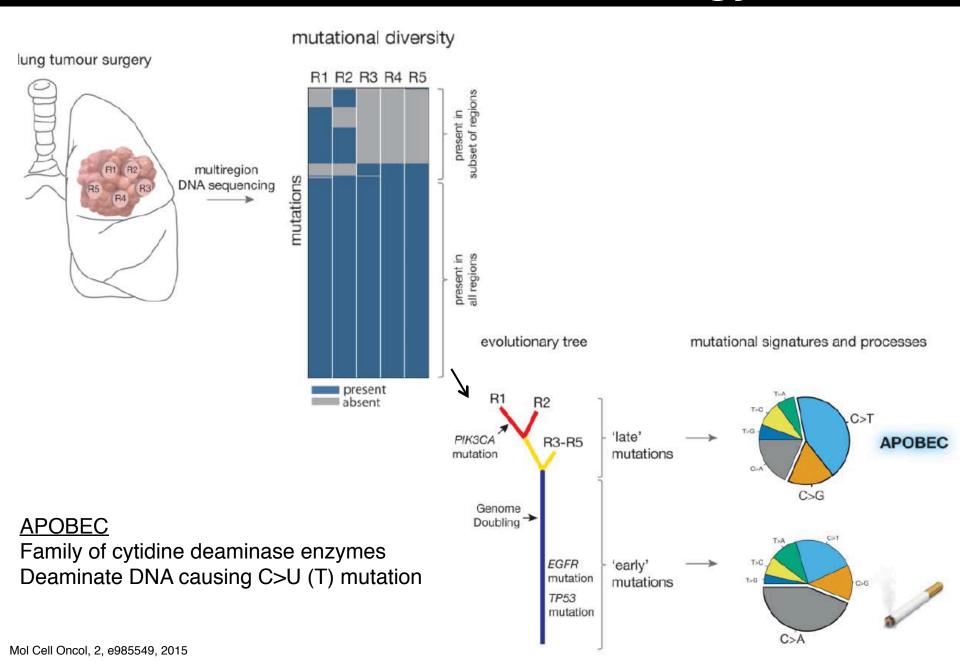
Biopsy => isolate CD8+ T cells

Expand KRAS G12D-reactive population => inject

All 7 lung tumors regress



## Understand cancer biology

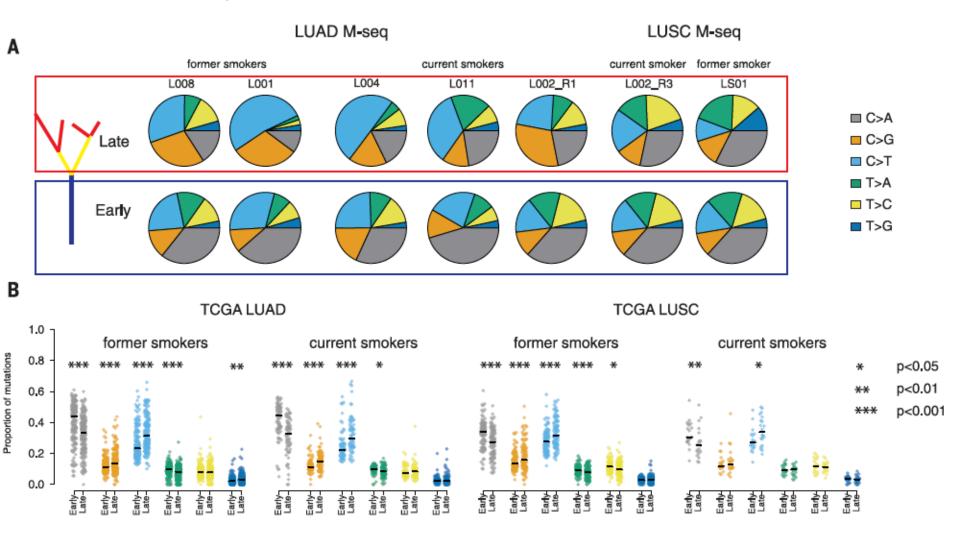


## **Understand cancer biology**

#### **APOBEC (C>T) mutation process**

Occurs later in time (more "Late" mutations)
Persists after smoking cessation

# Smoking-associated (C>A) mutation process Starts early (more "Early" mutations) Slows after smoking cessation



#### **Biomarkers**

Biomarkers "mark" a specific "biological" state

#### **Aspects of biomarkers**

Purpose Diagnostic, prognostic, etc.

Biological state – specificity, sensitivity

Analyte Type – gene, RNA, protein, cell number, small molecule, etc.

Number – single analyte, or multiple

Measurement Sensitivity of method

Specificity of method

Sample Diseased material (e.g., biopsy of tumor)

Indirectly affected material (e.g., blood lymphocytes in lung cancer)

Sample acquisition Invasive, non-invasive (blood, urine, etc.)

### MicroRNA-based biomarkers

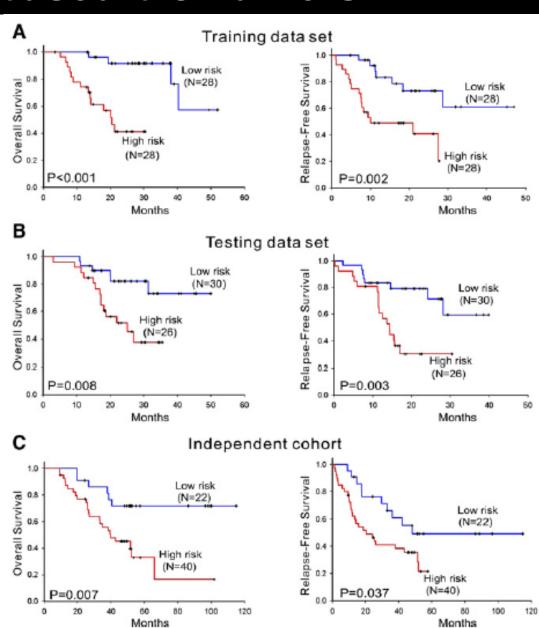
Prediction of recurrence and survival in non-small cell lung cancer

Sample: Tumor

Analyte: Multiple microRNAs

Measurement method: RTPCR

Kaplan-Meier survival curves:



### MicroRNA-based biomarkers

# Diagnosing lung cancer in smokers

Sample: Serum

Analyte: 34 microRNAs

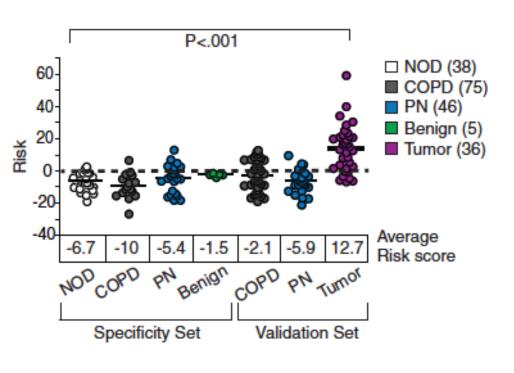
Measurement method: RTPCR

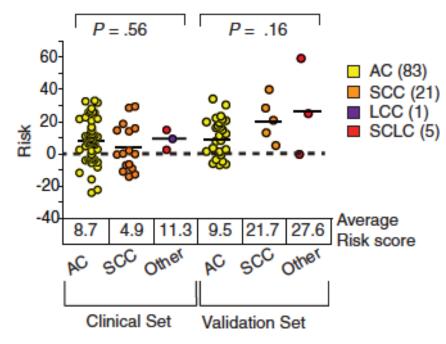
PN = pneumonia

COPD = chronic obstructive pulmonary disease NOD = non-cancer lung nodule detected by CT

Tumor = lung cancer

LCC = large cell ca. of lung





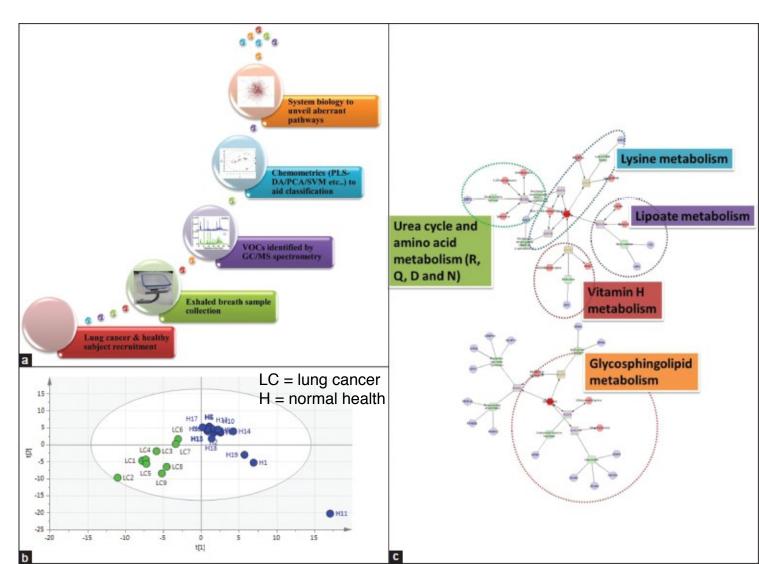
### **Biomarkers in breath**

Diagnosing lung cancer by breath volatile organic compounds

Sample: Breath

Analyte: Volatile organic compounds

Measurement method: Gas chromatography + mass spectrometry



J Carcinog, 12, 3, 2013

## **Gene mutation biomarkers**

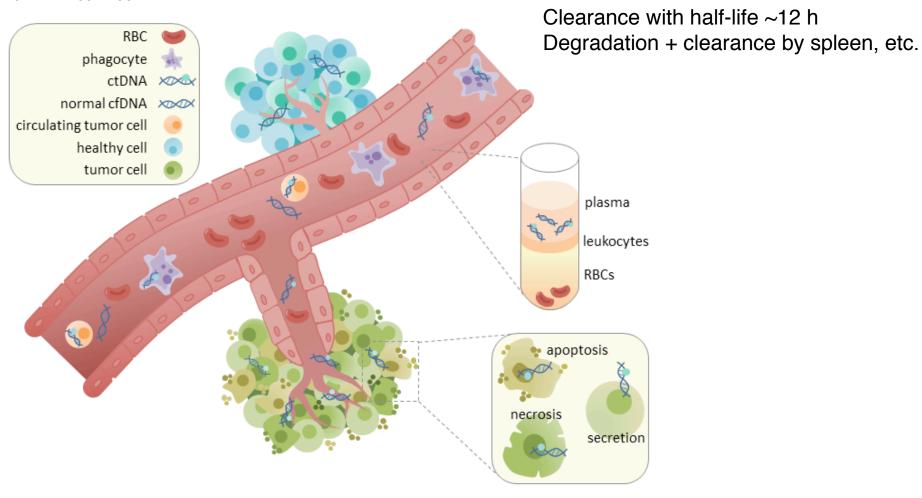
#### Examples that are clinically used

Biomarker	Treatment	Genomic aberration	Prevalence in NSCLC patients
EGFR	1. Tyrosine kinase inhibitors (e.g., gefitinib, erlotinib, and afatinib) 2. Monoclonal antibodies (e.g., cetuximab and necitumumab)	Activating mutation within intracellular catalytic domain of EGFR     Over-expression of extracellular part of EGFR	EGFR mutations (non-squamous histology)  1. ~15% in Caucasians  2. ~40% in Asians  3. ~75–80% in never-smoker Asians  EGFR mutations (squamous histology)  1. ~5%  EGFR over-expression  1. 39% in adenocarcinoma  2. 58% in squamous cell carcinoma  3. 38% in large-cell carcinoma
ALK	Tyrosine kinase inhibitors (e.g., crizotinib and ceritinib)	Chromosomal translocation and fusion of <i>ALK</i> gene	1. 3–5% in unselected NSCLC 2. ~10% in non-never-smokers 3. <1% in squamous carcinoma
MET	Tyrosine kinase inhibitors (e.g., tivantinib, cabozantinib, and crizotinib)     Monoclonal antibodies     (onartuzumab, AMG 102, ficlatuzumab)	Increased MET copy number     Over-expression of extracellular part     of MET receptor	2–4% MET amplification (untreated)     5–20% MET amplification in     EGFR-TKI-resistant tumors     3. 25–75% over-expression of extracellular part of MET receptor
ROS-1	Tyrosine kinase inhibitor (crizotinib)	Chromosomal translocation and fusion of <i>ROS-1</i> gene	1–2% in unselected population
KRAS	Downstream pathway inhibitors (e.g., MEK inhibitors selumetinib and trametinib)	Activating mutation within catalytic RAS domain	<ol> <li>KRAS are rare in never-smokers</li> <li>~25–30% in adenocarcinoma</li> <li>~5% in squamous cell carcinoma</li> </ol>

# **Circulating DNA biomarkers**

ctDNA = circulating tumor DNA cfDNA = cell-free DNA

~50 ng circulating cfDNA per ml of plasma = ~10,000 genome DNA



# Assigned ctDNA biomarker paper

# nature

# Accelerated Article Preview

#### ARTICLE

doi:10.1038/nature22364

#### Phylogenetic ctDNA analysis depicts early stage lung cancer evolution

Christopher Abbosh, Nicolai J. Birkbak, Gareth A. Wilson, Mariam Jamal-Hanjani, Tudor Constantin, Raheleh Salari, John Le Quesne, David A Moore, Selvaraju Veeriah, Rachel Rosenthal, Teresa Marafioti, Eser Kirkizlar, Thomas B K Watkins, Nicholas McGranahan, Sophia Ward, Luke Martinson, Joan Riley, Francesco Fraioli, Maise Al Bakir, Eva GrÖnroos, Francisco Zambrana, Raymondo Endozo, Wenya Linda Bi, Fiona M. Fennessy, Nicole Sponer, Diana Johnson, Joanne Laycock, Seema Shafi, Justyna Czyzewska-Khan, Andrew Rowan, Tim Chambers, Nik Matthews, Samra Turajlic, Crispin Hiley, Siow Ming Lee, Martin D. Forster, Tanya Ahmad, Mary Falzon, Elaine Borg, David Lawrence, Martin Hayward, Shyam Kolvekar, Nikolaos Panagiotopoulos, Sam M Janes, Ricky Thakrar, Asia Ahmed, Fiona Blackhall, Yvonne Summers, Dina Hafez, Ashwini Naik, Apratim Ganguly, Stephanie Kareht, Rajesh Shah, Leena Joseph, Anne Marie Quinn, Phil Crosbie, Babu Naidu, Gary Middleton, Gerald Langman, Simon Trotter, Marianne Nicolson, Hardy Remmen, Keith Kerr, Mahendran Chetty, Lesley Gomersall, Dean A. Fennell, Apostolos Nakas, Sridhar Rathinam, Girija Anand, Sajid Khan, Peter Russell, Veni Ezhil, Babikir Ismail, Melanie Irvin-sellers, Vineet Prakash, Jason F. Lester, Malgorzata Kornaszewska, Richard Attanoos, Haydn Adams, Helen Davies, Dahmane Oukrif, Ayse U Akarca, John A Hartley, Helen L Lowe, Sara Lock, Natasha Iles, Harriet Bell, Yenting Ngai, Greg Elgar, Zoltan Szallasi, Roland F Schwarz, Javier Herrero, Aengus Stewart, Sergio A Quezada, Peter Van Loo, Caroline Dive, C. Jimmy Lin, Matthew Rabinowitz, Hugo JWL Aerts, Allan Hackshaw, Jacqui A Shaw, Bernhard G. Zimmermann, the TRACERx consortium, the PEACE consortium & Charles Swanton

# Study design

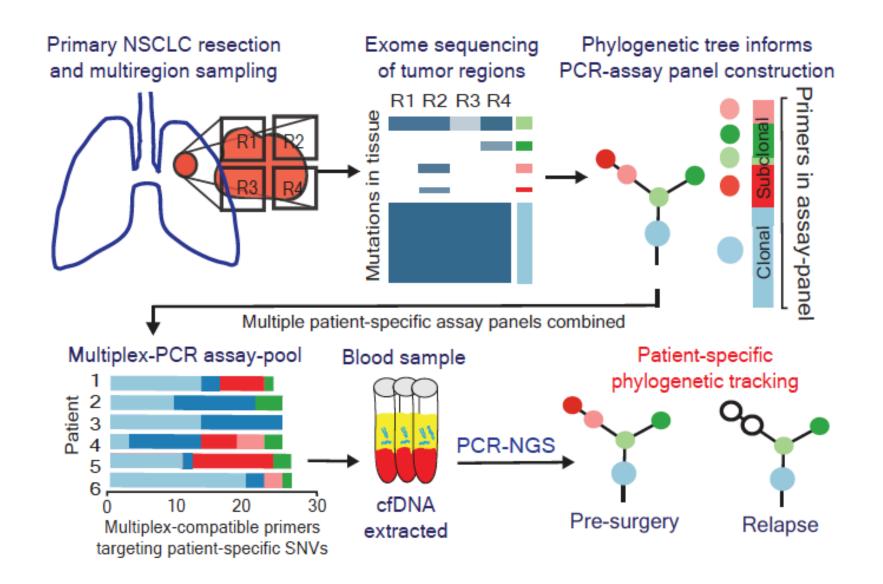
#### **TRACER**x

Sequencing
NSCLC blood and samples
~900 patients
Diagnosis to 5 y later
Cancer Research UK
2013–

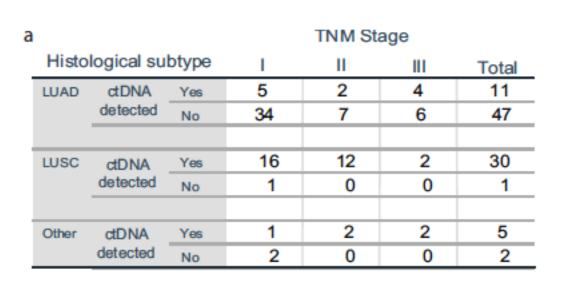
First 100 enrollees for this study

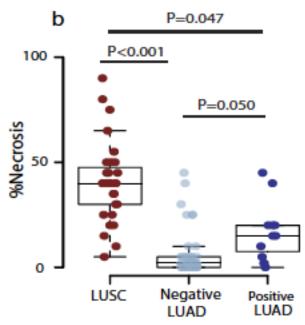
<u>Tissue microarray</u> with duplicates of primaries – Ki67 staining <u>Histopathology</u> – subtype, %necrosis, lymphovascular invasion <u>Radiology</u> (CT, PET) – SUV, tumor volume, TBR <u>Exome DNA</u> sequencing – blood, tumor (primary, metastasis): 200 ng DNA per sample <u>Circulating free DNA</u> – PCR for 18 (median) patient-specific SNVs (mutations) -> sequencing to find mutation frequency

# Study design



### **Tumor necrosis improves ctDNA detection**





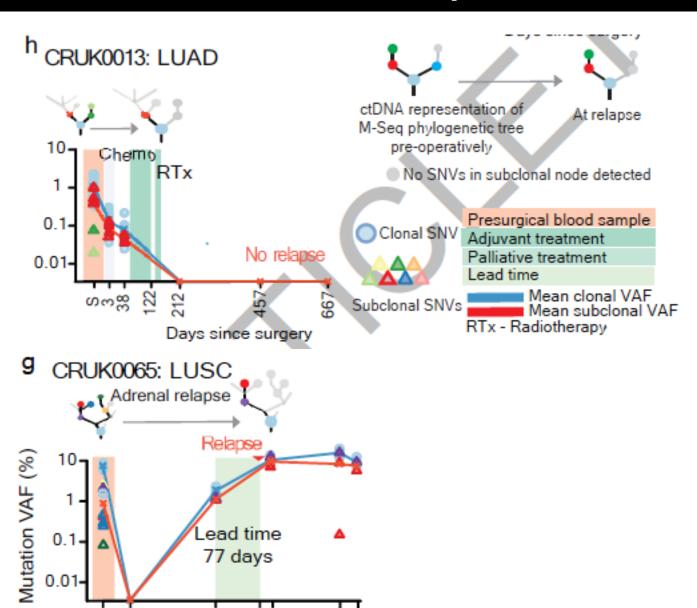
#### Predictors of ctDNA detection by multiplex-PCR NGS in early stage NSCLC

	Univariable analysis		Multivariable analysis	
	OR (95% CI)	P-value	OR (95% CI)	P-value
Clinicopathological variables				
Non-adenocarcinoma histology	49.85 (12.93 -192.19)	< 0.001	40.76 (4.55 - 365.14)	0.001
%Ki67+cells (10% increase)	1.72 (1.40 - 2.12)	< 0.001	1.40 (1.05 - 1.84)	0.022
Lympho-vascular invasion	2.53 (1.10 - 5.80)	0.028	5.84 (1.07 - 32,03)	0.042
Necrosis (10% increase)	2.16 (1.58 - 2.97)	< 0.001	1.04 (0.64 - 1.71)	0.862
Path tumor size (10mm increase)	1.45 (1.13 - 1.86)	0.004	1.32 (0.91-1.91)	0.134
Lymph-node involvement	3.60 (1.33 - 9.77)	0.012	3.82 (0.61 - 23.99)	0.153
Male gender	1.80 (0.78 - 4.16)	0.172	1.06 (0.21-5.39)	0.941
Age (years)	0.96 (0.92 - 1.01)	0.115	0.99 (0.92-1.07)	0.820

4

Days since surgery

#### ctDNA biomarkers predict cancer relapse



## **Questions?**

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