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## Canonical model of drivers & passengers in cancer

#### **Drivers**

directly confer a selective growth advantage to the tumor cell.

A typical tumor contains 2-8 drivers.

identified through signals of positive selection.

Existing cohorts of ~100s give enough power to identify

#### **Passengers**

Conceptually, a passenger mutation has no direct or indirect effect on tumor progression.

There are 1000s of passengers in a typical cancer genome.

[Vogelstein Science 2013. 339:1546]



## Conceptual extension of the canonical model of drivers & passengers



### Mutational processes carry context-specific signatures



# PCAWG : most comprehensive resource for cancer whole genome analysis



Adapted from Campbell et. al., bioRxiv ('17)

Goal of PCAWG:

- To understand role of non-coding ٠ regions of cancer genomes in cancer progression.
- Union of TCGA-ICGC efforts •
- Jointly analyzing ~2800 whole genome • tumor/normal pairs
  - > > 580 researchers
  - 16 thematic working groups
  - ~30M total somatic SNVs



# A case study: pRCC

- Kidney cancer lifetime risk of 1.6% & the papillary type (pRCC) counts for ~10% of all cases
- TCGA project sequenced 161 pRCC exomes & classified them into subtypes
- Also, 35 WGS of TN pairs



### Introduction

- Background: driver-and-passenger model (w/ conceptual extension) & mutational spectra & signatures
- Data source: PCAWG comprehensive WGS on >2.5K + focus on 35 pRCC WGS

### Overall functional impact of variants

- FunSeq entropy-weights multiple features to evaluate the functional impact of SNVs
- Investigating how the fraction of highimpact (non-strong-driver) SNVs scales & how it relates to survival

### Differential burdening from various mutational processes

- Diff. burdening of TF sub-networks, results from spectra & signatures differentially affecting binding motifs
- High & low impact mutations assoc. w/ diff. signatures
- Number of mutations in DHSes assoc. w/ specific chromatin mod. mutation

### Functional impact & tumor evolution

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# Funseq: a flexible framework to determine functional impact & use this to prioritize variants



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- Entropy based method for weighting consistently many genomic features
- Practical web server
- Submission of variants & precomputed large data context from uniformly processing large-scale datasets

# Overall functional impact distribution of PCAWG mutations





 Funseq molecular functional impact of ~30M variants in >2500 PCAWG samples Division of PCAWG Lymph-CLL cohort based on average impact of non-driver variants (high v low)

# In many PCAWG cohorts, we observe the fraction of impactful "passengers" decreases with increase in total mutation burden.



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# Mutational burdening of TF-subnetworks due to SNVs breaking & creating binding sites



# Kidney cancer as an example: differential TF burdening correlates with mutational spectrum



# CpGs drive inter-patient variation in pRCC mutational spectra

- The loadings on PC1 are mostly [C>T]G
- Confirmed by higher C>T% in CpGs in the hypermethylated group (cluster1)







- Signatures burden the genome disproportionally
- We found 1 pRCC has ApoBEC signature, but nothing in a larger ccRCC cohort

#### **Total mutation counts**

DHS mutation %



Key mutation affects mutational landscape which, in turn, affects overall burden in pRCC

 Chromatin remodeling defect ("mut") leads to more mutations in open chromatin (raw number & fraction) in those pRCC cases w/ the mutation

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Sub-clonal architecture of mutations in PCAWG

As expected, drivers are enriched in earlier subclones. Overall, no such enrichment among passengers.

High impact passengers are slightly enriched among early subclones.

Particularly, passengers in tumor suppressor (in contrast to oncogenes, which require specific mutations).



# Continuous correlation of functional impact & VAF

Among mutations in driver genes: higher impact mutation => spreads to more cells.

Still true after removing all known driver variants from driver genes. (Latent drivers?)

Outside driver genes: higher impact mutation => spreads to fewer cells (Deleterious passengers?)

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Functional impact in PCAWG (PanCancer.info)

S **Kumar,** J **Warrell**, W Meyerson, P McGillivary, L Salichos, S Li, A Fundichely, E Khurana, C Chan, M Nielsen,C Herrman, A Harmanci, L Lochovsky,Y Zhang, X Li,

**PCAWG** Drivers & Functional Interpretation Group, ICGC/TCGA Pan-Cancer Analysis of Whole Genomes Network

FunSeq.gersteinlab.org Y Fu, E Khurana, Z Liu, S Lou, J Bedford, XJ Mu,

S Lou, J Bedford, XJ Mu KY Yip,

pRCC S Li, B Shuch See gersteinlab.org/jobs ! Hiring Postdocs.

> Acknowledgements

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