

ITCR Monthly Meeting

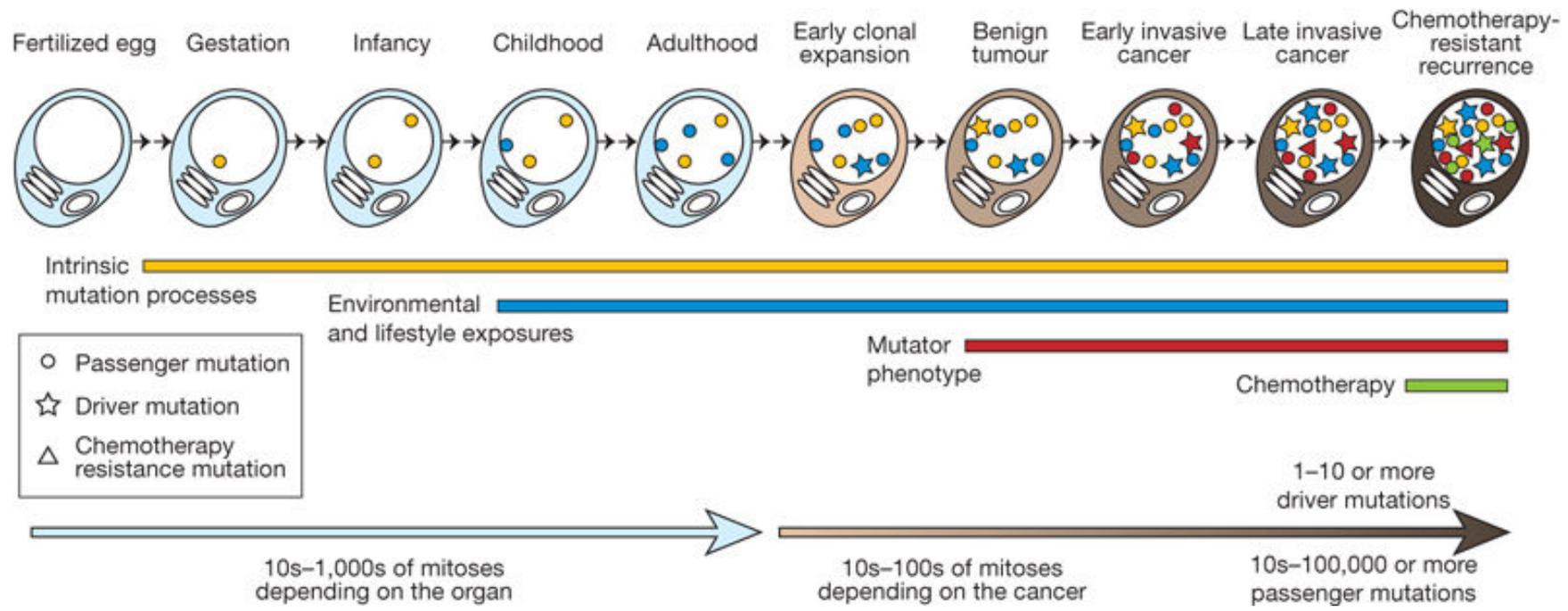
October 4, 2019

Eliezer Van Allen

Felix Dietlein

Dana-Farber Cancer Institute, Boston

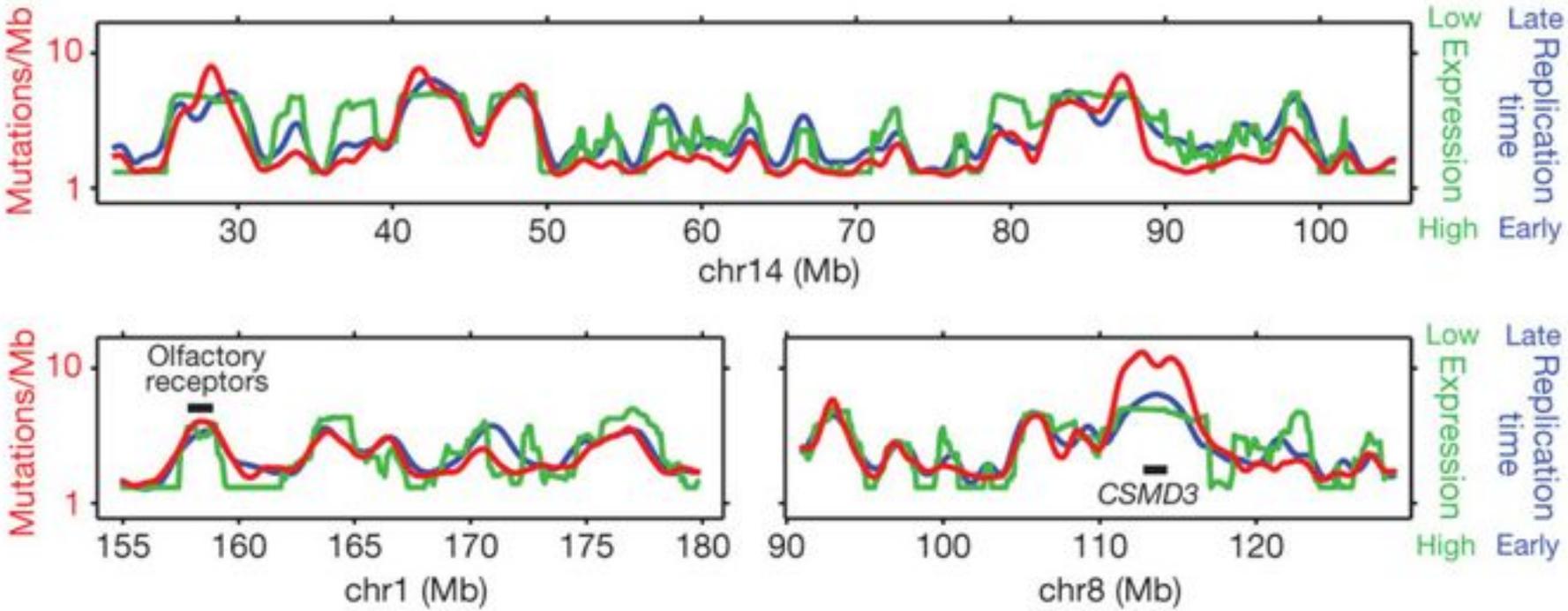
Passenger and driver mutations in cancer



Searching for a needle in the haystack

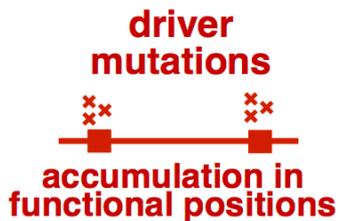
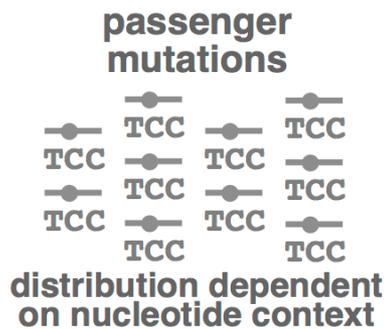


The background mutation rate varies across the human genome

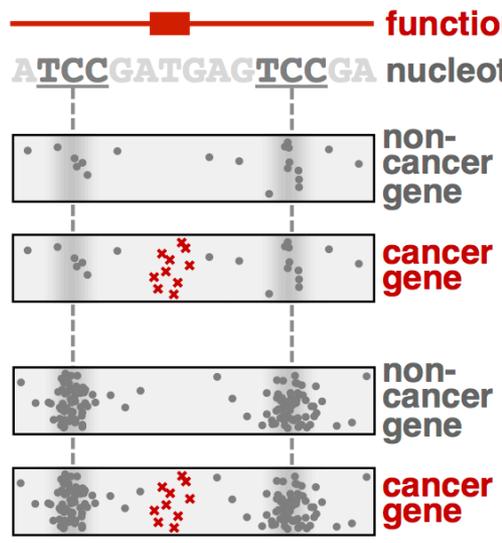
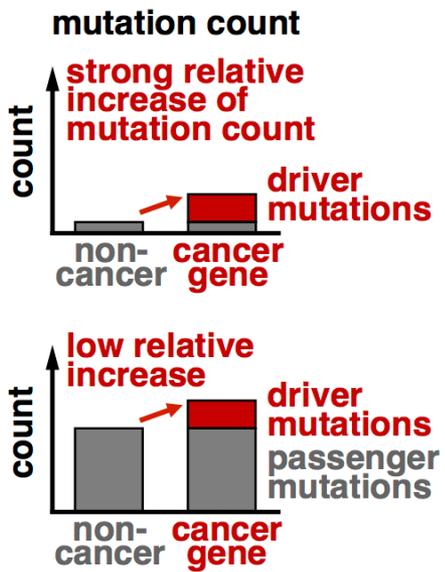


**synonymous mutations provide a way
to account for the variation of
the regional background mutation rate
in the coding space**

Nucleotide contexts provide an additional signal in favor of driver mutations



detection of cancer genes based on mutation counts and nucleotide context



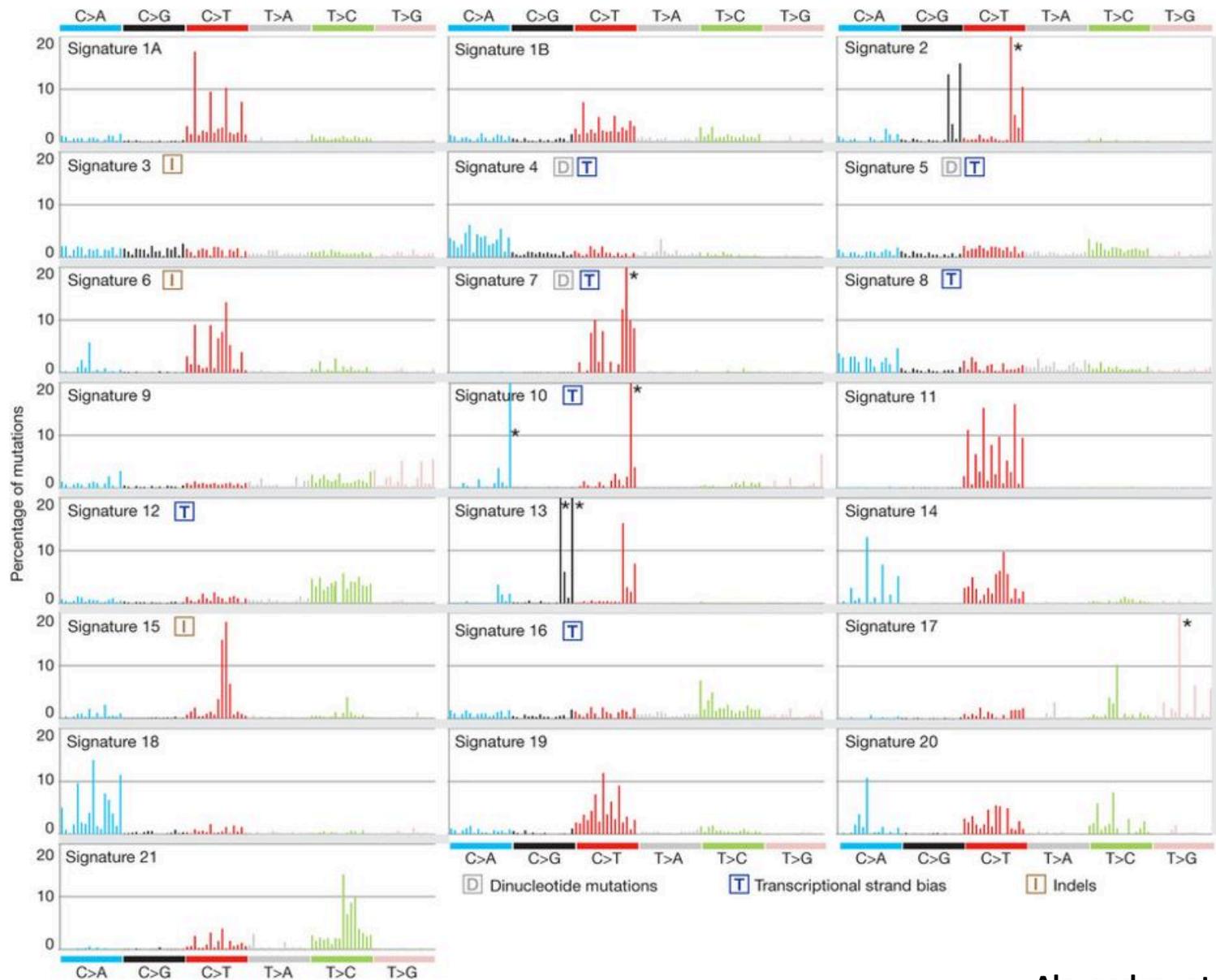
low background mutation rate

search for genes with increased no. mutations above the local bkgd. mutation rate

high background mutation rate

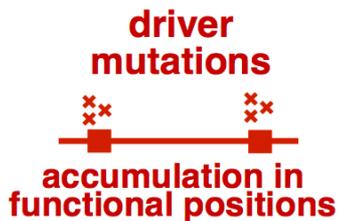
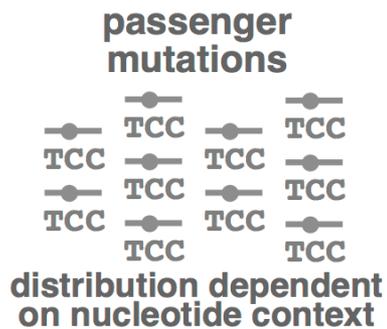
search for genes with increased no. mutations in an unusual context

Mutational signatures in cancer

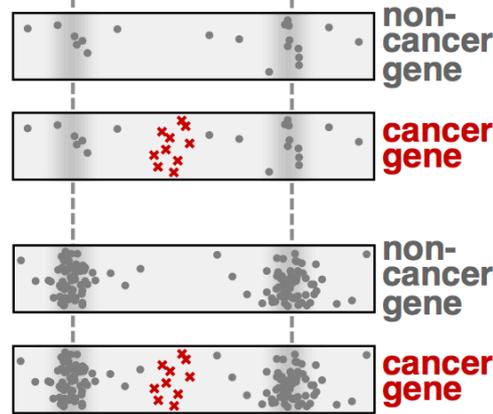
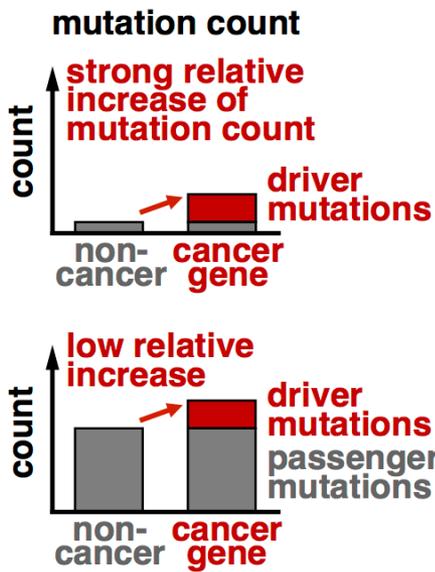


**nucleotide contexts may provide a way
to identify driver mutations
in the non-coding space**

Nucleotide contexts provide an additional signal in favor of driver mutations



detection of cancer genes based on mutation counts and nucleotide context



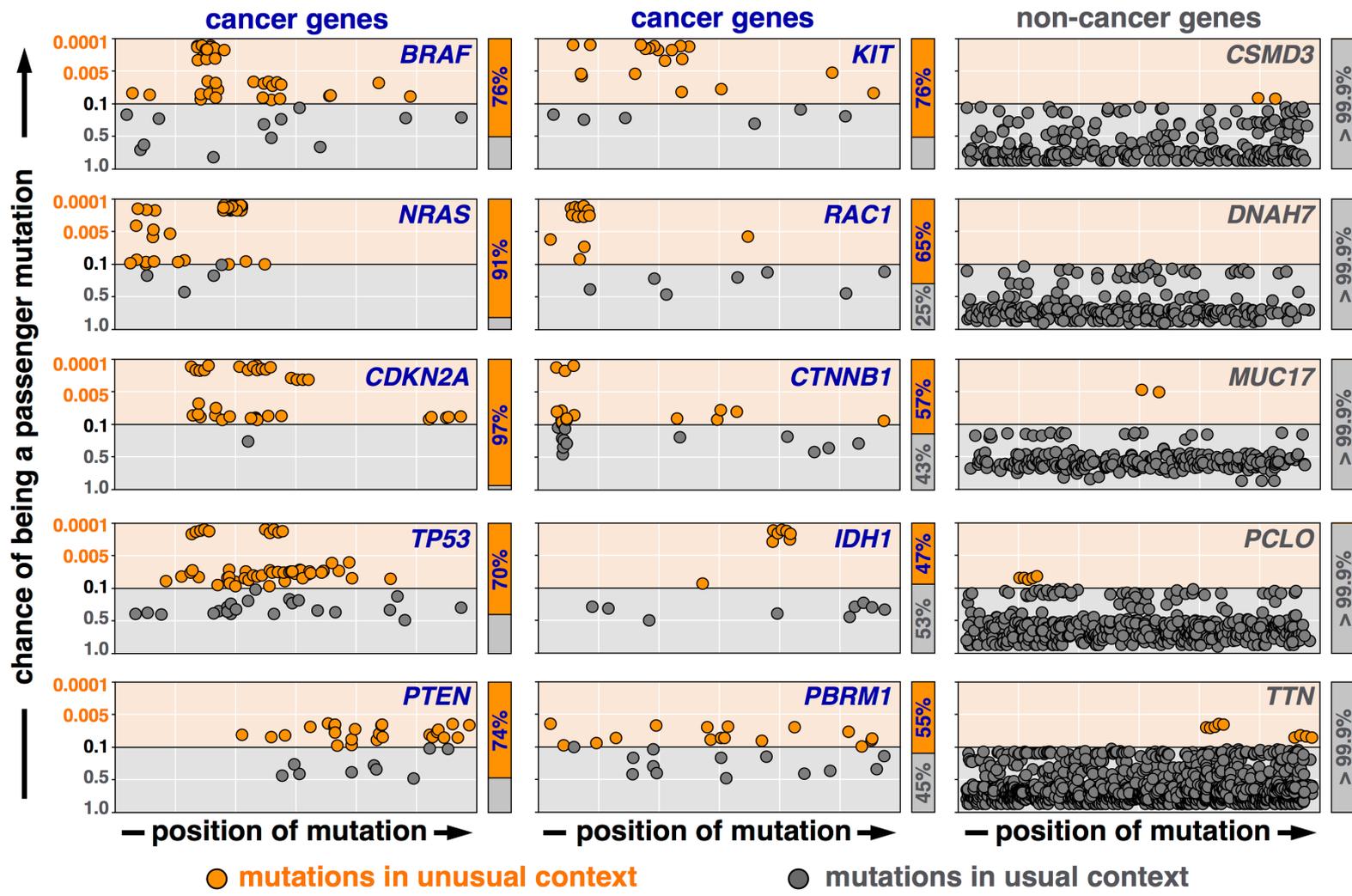
low background mutation rate

search for genes with increased no. mutations above the local bkgd. mutation rate

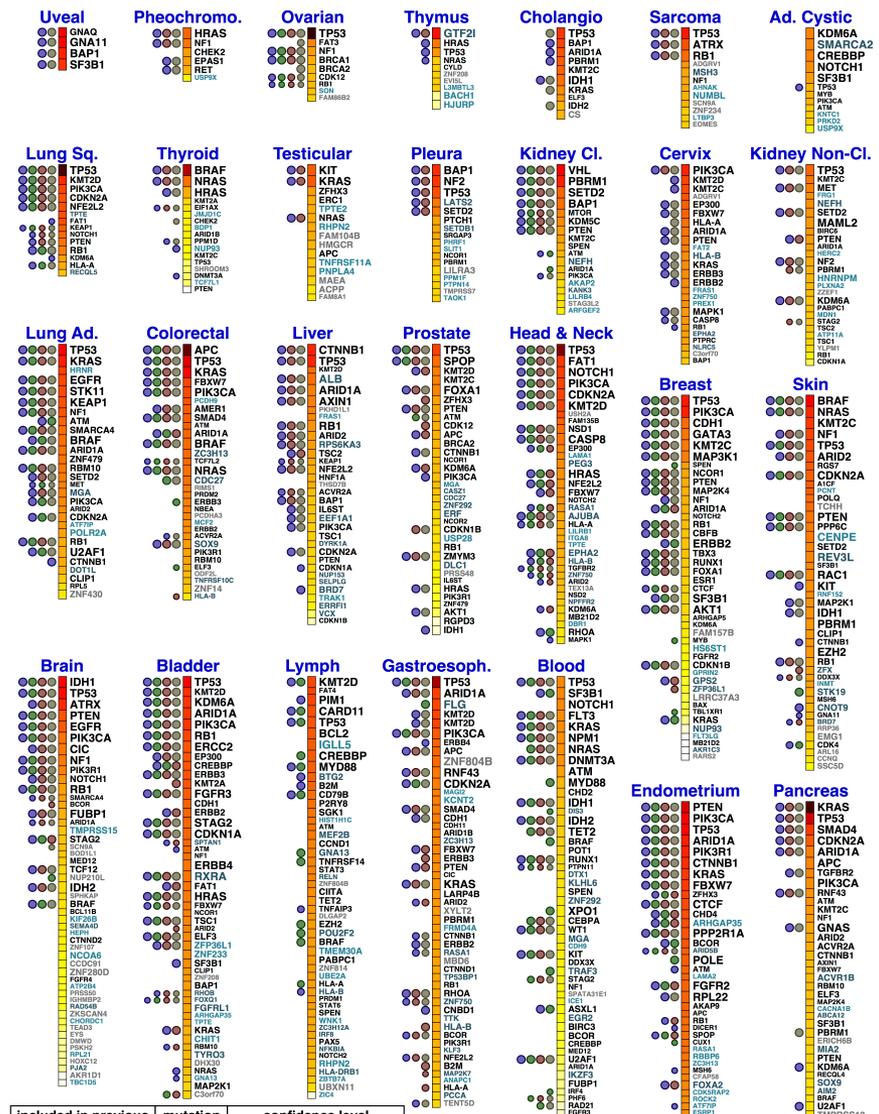
high background mutation rate

search for genes with increased no. mutations in an unusual context

Nucleotide contexts provide an additional signal in favor of driver mutations



A catalog of driver genes in human cancer



<p>included in previous catalog</p> <p> Bailey et al., Cell 2018 Lawrence et al., Nature 2014 Martignoni et al., Cell 2017 TCGA papers </p>	<p>mutation freq. [%]</p> <p> 25 6 1 <1 </p>	<p>confidence level based on literature</p> <p> level A Cancer Gene Census level B literature support in same tumor type level C literature support in different tumor type level D no literature support </p>	<p>significance</p> <p> highly significant gene less significant gene </p>
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Aim 1 To define the landscape of noncoding cancer driver mutations using nucleotide context

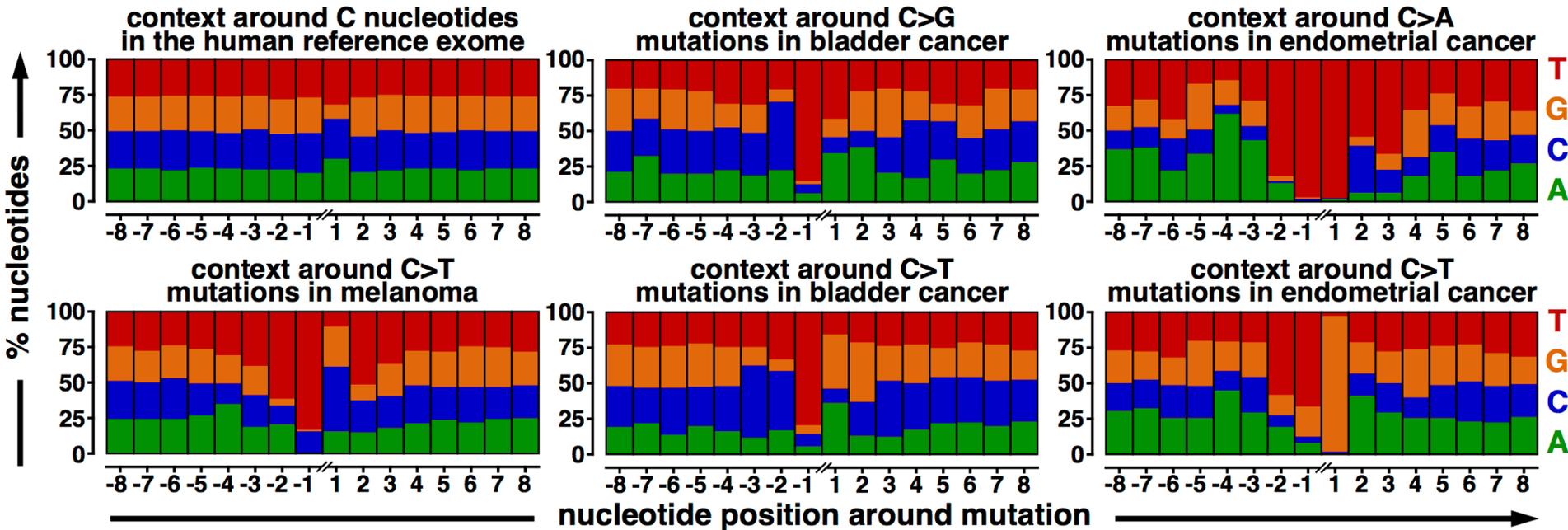
Rationale: The role of noncoding mutations in tumor formation largely remains unknown. The emerging availability of large whole-genome sequencing datasets creates an ample opportunity to develop new mutational significance algorithms

Hypothesis: Current mutational significance algorithms have been developed for coding mutations. Applying them to noncoding regions does not exploit the full potential of whole-genome sequencing datasets

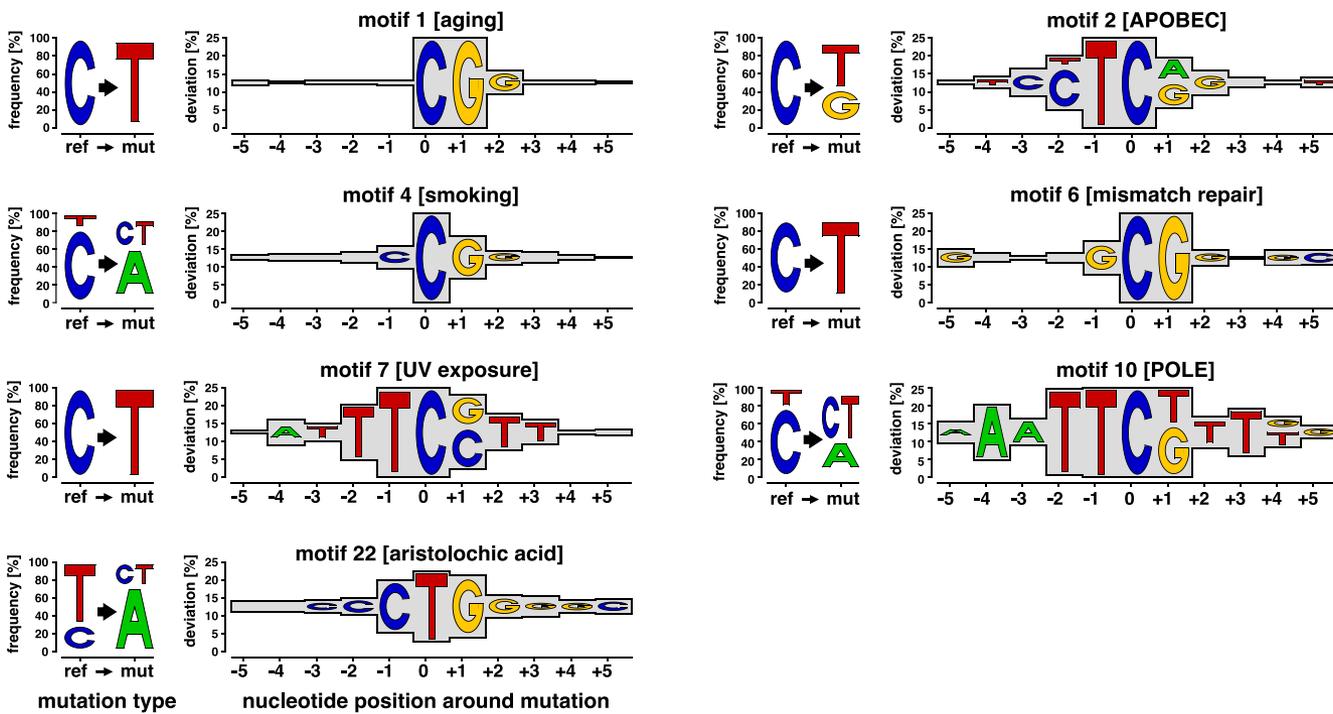
Approach: Develop a new statistical model that accounts for:

- Context-dependent distribution of passenger mutations
- Modeling of the background mutation rate
- Partitioning of the background mutation rate:
- Modeling the sequence composition of the reference genome
- Coverage fluctuation

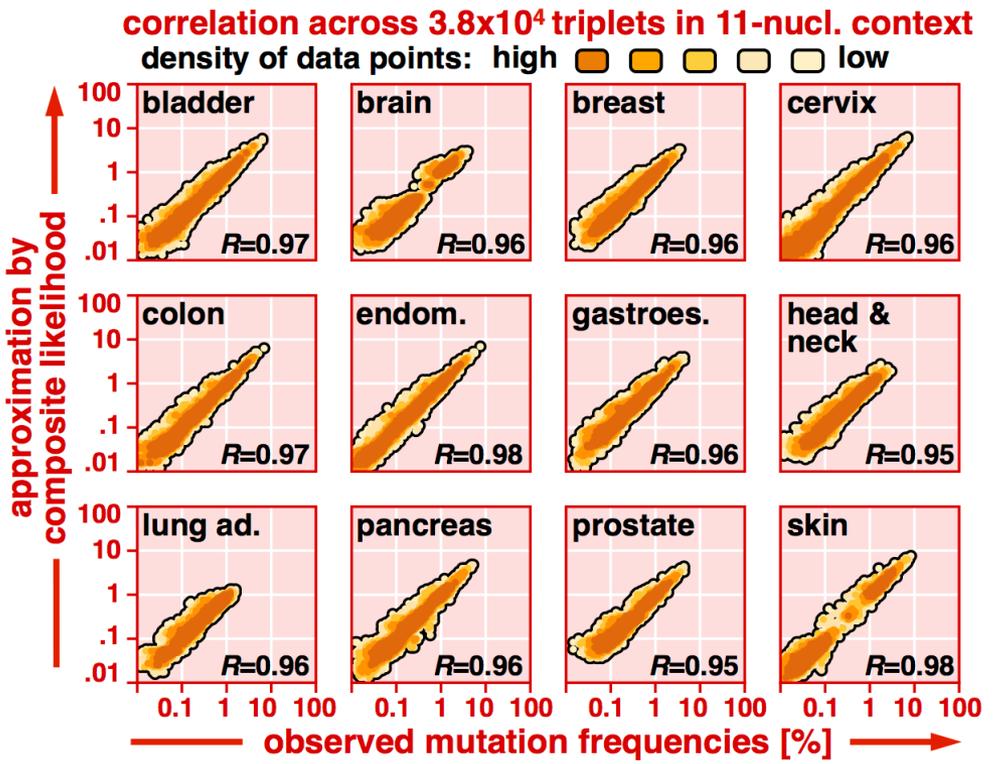
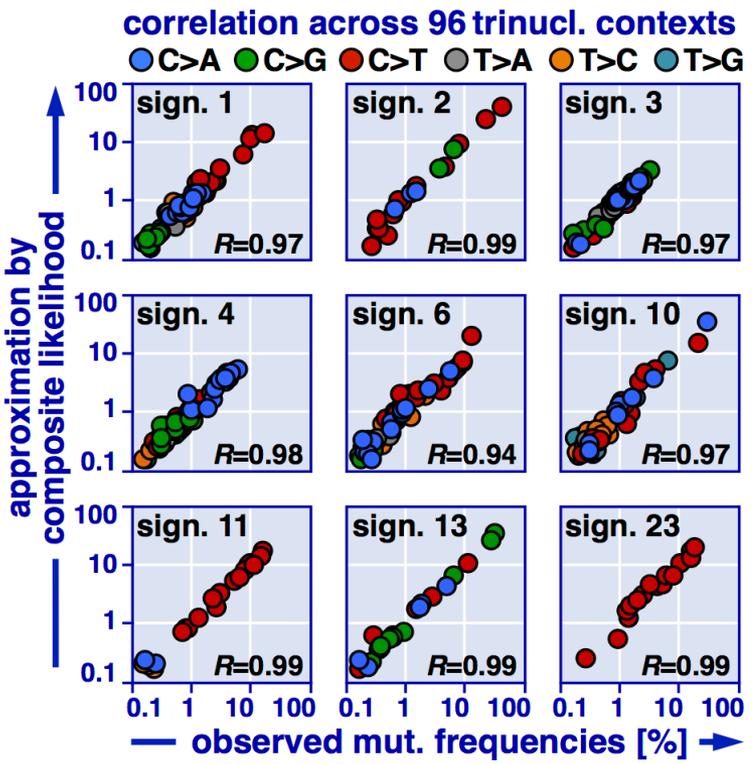
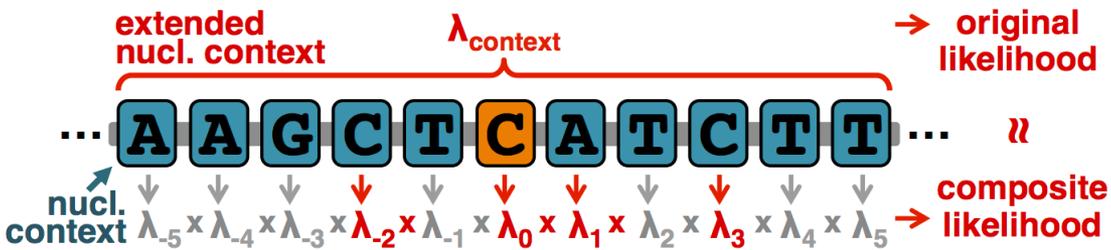
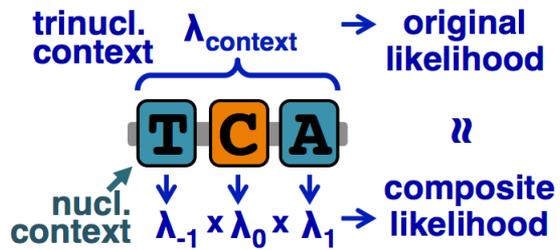
Extended nucleotide contexts around passengers contain a signal



Extended nucleotide contexts around passengers contain a signal



A composite likelihood model to account for extended nucleotide contexts



Aim 2 To evaluate the clinical significance of passenger mutation distribution patterns for immunotherapies

Rationale: Despite the clinical relevance of immunotherapies, robust markers for patient stratification are currently lacking. The most common parameter to predict clinical response is the mutational burden.

Hypothesis: Recent studies have suggested that not only the total number of passenger mutations but also their underlying biology mediates clinical response to immunotherapy. We and others have shown that the underlying biology of passenger mutations is not only reflected by their surrounding trinucleotide context but that a relevant signal is in the extended nucleotide context.

Aim 2 To evaluate the clinical significance of passenger mutation distribution patterns for immunotherapies

Approach: Use extended nucleotide contexts for “improved” mutational signatures

- Apply our composite likelihood model to a sequencing dataset of 249 tumors with clinically annotated outcomes to immune checkpoint therapy
- Evaluate the immunogenicity of each individual passenger mutation by using established algorithms , such as POLYSOLVER algorithm and NetMHCpan-3.0
- Determine which passenger mutations from the whole-exome sequencing data can be rediscovered in the corresponding RNA sequencing data of the same tumor.
- Develop an integrative model to combine these aspects and predict the immunotherapy response of a tumor, based on the immunogenicity of its passenger mutations