

# Annotation Free Analysis of Recurrent Somatic Mutations

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# Identifying Non-coding Driver Mutations

- ▶ Non-coding variants may serve as drivers in many cancer types:
  - ▶ TERT, PLEKHS1, WDR74 and SDHD promoters
  - ▶ miRNA binding sites
- ▶ Our goal is to identify regions in the noncoding regions that are highly mutated

## Previous Efforts

- ▶ Two papers

- ▶ Weinhold, N. *et al.* Genome-wide analysis of noncoding regulatory mutations in cancer. *Nature Genetics*
- ▶ Melton, C. *et al.* Recurrent somatic mutations in regulatory regions of human cancer genomes. *Nature Genetics*

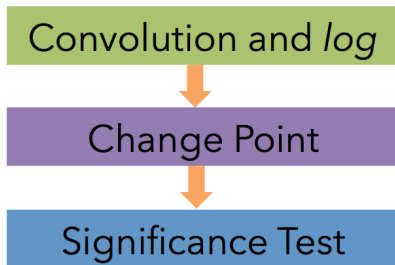
- ▶ Drawback

- ▶ Annotations - low genome coverage
- ▶ Small Fixed Regions - low mutation rate resolution
- ▶ Not dynamic, not true region



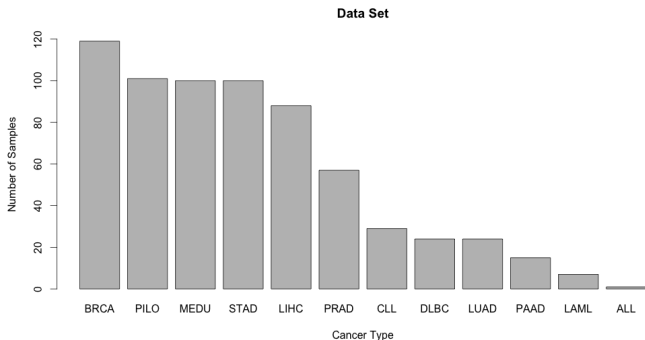
# Annotation Free Analysis

- ▶ Goal:
  - ▶ Auto-cluster genome into regions of enriched mutations
- ▶ 3 Steps:



# Dataset

- ▶ Somatic Mutations from:
  - ▶ 12 Cancer Types
  - ▶ 665 WGS total
- ▶ Includes *Alexandrov et al* data (WGS)

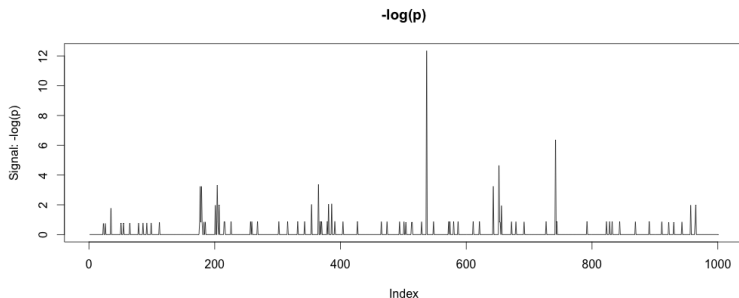


## Convolution and log

- ▶ Divide genome into 50bp bins
- ▶ Number of mutations in bin,  $k$
- ▶ For a single cancer type  $\sim$  Binomial
- ▶ Convolution Method: Combine discrete probabilities over all cancer types
  - ▶  $\Pr(K \geq k) = 1 - \Pr(K < k)$
  - ▶ linear combination of discrete probabilities
  - ▶ Result: single  $p$ -value for each 50bp bin

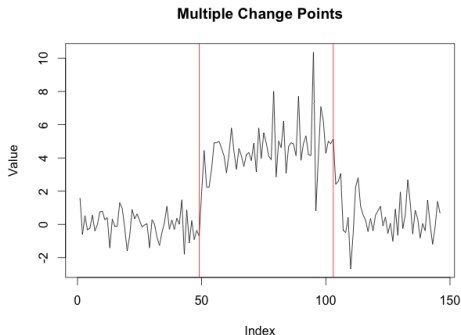
## Convolution and log

- ▶ For each  $p$ -value, take negative log ( $-\log$ )
- ▶ Creates signal for each 50bp bin, correlating to significance
- ▶ Pros
  - ▶ Amplify significant mutation count signal
  - ▶ Reduce signals that are less significant
  - ▶ Removes some noise found in mutation counts



# Change Point Detection

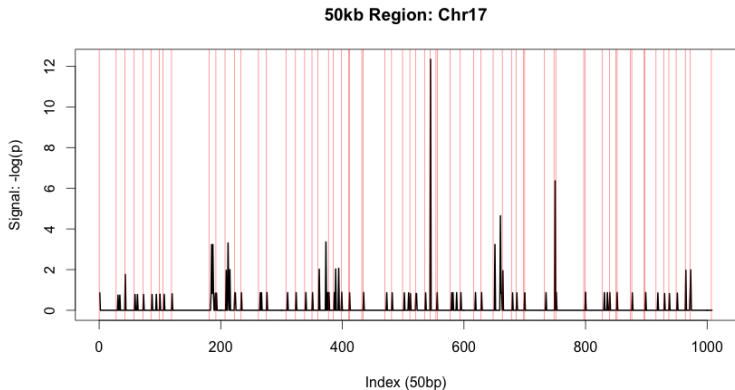
- ▶ Motivation:
  - ▶ Change points: determine start and end of region of interest
- ▶ Change in distribution before and after point
- ▶ Series of change points can be detected



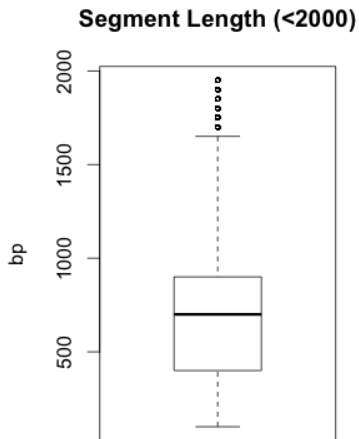
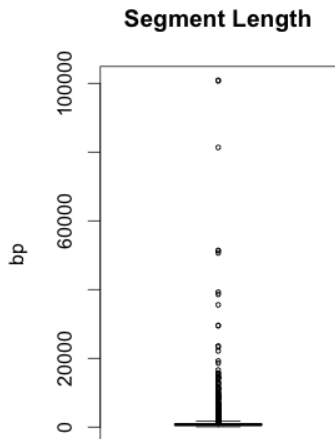


# Change Point Detection: Usage

- ▶  $p$ -values  $\sim$  Uniform
- ▶  $-\log(p) \sim$  Exponential
- ▶ Apply change point algorithm to dataset of  $-\log(p)$  for whole genome
- ▶ Example Result:

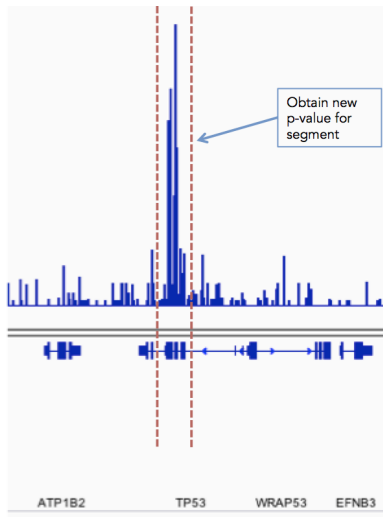


## Change Point Segment Lengths



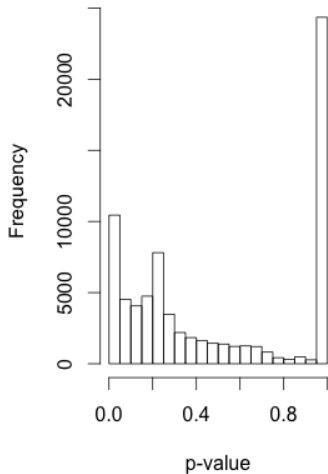
# Statistical Testing

- ▶ Statistical testing on each segment
- ▶ Assess significance of segments determined by change point
- ▶ New  $p$ -value for segment (Convolution Method)

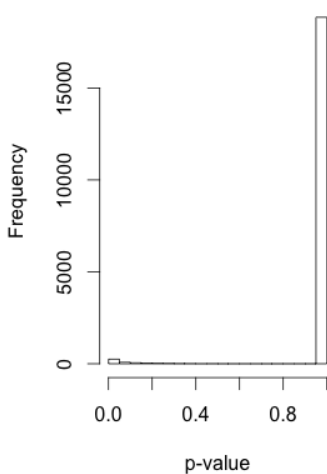


## Segment P-values

**Segment < 1000bp**

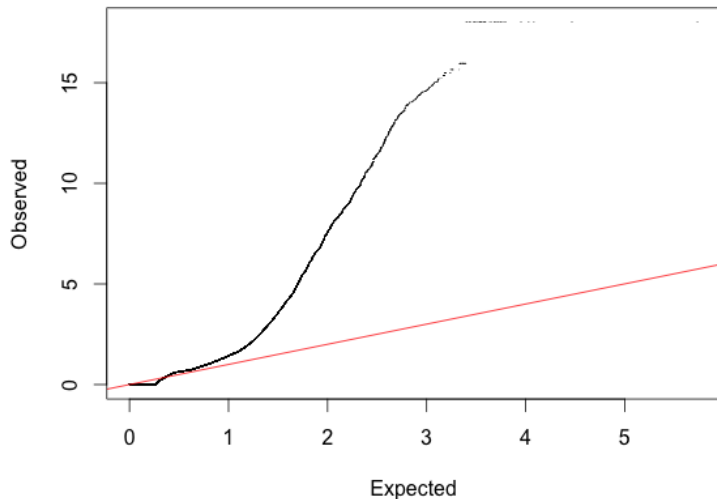


**Segment > 1000bp**



## Preliminary Results

chr17: p-value QQ Plot



## Further Analysis

- ▶ Perform FDR or other p-value correction
  - ▶ Filter for significant segments
- ▶ Intersect significant segments with annotations
  - ▶ Expectation:
    - ▶ Intersections with known regulatory elements
    - ▶ Regions not contained in annotations, but also significant