

# Microsatellite Instability Classification by NGS

A decorative graphic consisting of several horizontal lines in shades of blue and white, extending from the left side of the slide towards the right, positioned below the title.

# Microsatellite Instability

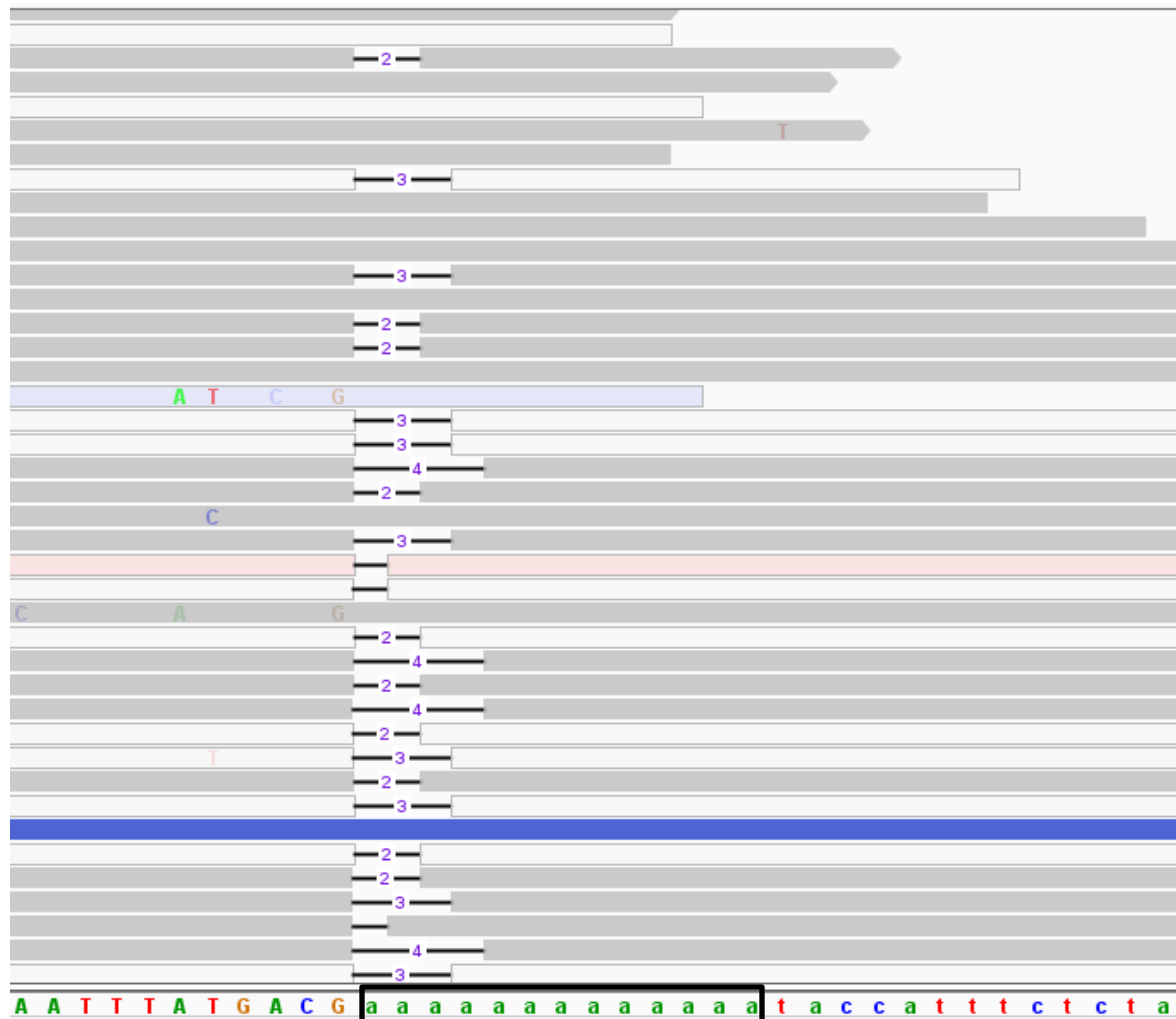
- Characterized by the “spontaneous loss or gain of nucleotides from repetitive DNA tracts”
  - Caused by defective mismatch repair
- Is a diagnostic phenotype for certain cancer types with clinical implications
  - Actionable marker for immune-checkpoint-blockade therapy

(Hause et al. 2016)

# Sample

- 602 BAM/BAI files from patients in TCGA
  - 241 Colon/Rectal/Colorectal Adenocarcinoma
  - 361 Uterine Endometrial Carcinoma
- Annotated with MSI-PCR result
  - MSS
  - MSI-L, MSI-H
- 27 representative loci gathered from literature

# Repeat Identification



# Repeat Identification



- Flanking regions
  - 7 bases long
    - Long enough to be unique, identifiable
    - Short enough to maximize usable reads
  - Immediately adjacent to the mononucleotide repeat

# Repeat Identification

- Scan and find flanking regions
  - Ordered search
  - Brute-force fuzzy matching
    - Mismatch allowance = 2
- Validation: filter
  - Baseline = 90%
  - Modified for some loci

# MSI Calling

- Binary classifier
  - Simplification of MSI-H and MSI-L
- Optimization
  - Accuracy
    - Sensitivity
    - Specificity
- Locus-based calls
  - More common in literature
- Overall MSI call based on all 27 loci

# Machine Learning Approach

- TensorFlow
  - Python API developed by Google
- Minimize error in linear function, apply sigmoid curve to calculate  $p(\text{MSI})$
- Simplification
  - 5 loci with high number of BAMs with reads available, good depth at the loci



# Datasets

## ■ All files

- 602 BAM files with MSS, MSI-L or MSI-H annotation
  - 241 COAD-READ, 361 UCEC
  - 212 MSI, 390 MSS

## ■ Mode training set

- Randomly generated set of 100 BAM files with MSS status to generate mode length
  - 43 COAD-READ, 57 UCEC
  - 0 MSI, 100 MSS

# Datasets (continued)

## ■ Training Set

- 300 BAM files of mixed status to train the model
  - 118 COAD-READ, 182 UCEC
  - 130 MSI, 170 MSS

## ■ Validation Set

- 100 BAM files of mixed status to validate the model
  - 44 COAD-READ, 56 UCEC
  - 40 MSI, 60 MSS

## ■ Test Set

- 102 remaining BAM files of mixed status to test the model
  - 36 COAD-READ, 66 UCEC
  - 42 MSI, 60 MSS

# Datasets (continued)

- Randomly generated sets
  - Representative of both cancer types
- Low coverage problem
  - Low coverage at certain loci excludes some files from some locus analyses
  - Actual size of datasets vary based on individual BAM coverage at any given locus

# MSI Calling - Features

- Number of lengths
  - $[11, 12, 11, 13, 14] = 4$
- Distance from mode
  - Average distance from MSS sample mode of all reads
- Standard deviation
- Average Length

# Locus-Based Calling

- Generate ML model for each locus individually
  - Weight for each feature, bias
  - Sigmoid activation function to generate a probability (0, 1)
- Choose loci with highest AUC
  - Exclude those that have no reads in many BAM files
  - Try different combinations of loci
- Determine MSI status based on number of loci with 'MSI' call

# Locus-Based Calling (continued)

- Loci examined
  - BAT-26
  - MSI-07
  - MSI-09
  - H-06
  - MSI-06
  - MSI-04
  - HSPH1-T17
- Threshold: 0.500000
- Min no. loci: 3
- Total files: 93
- Correct predictions: 78
  - True pos: 24
  - True neg: 54
  - False pos: 2
  - False neg: 13
- Accuracy: **0.838710**
- Sensitivity: **0.648649**
- Specificity: **0.964286**

# Probability-Based Calling

- Generate ML model for each locus individually
  - As before
- Choose loci with highest AUC
  - As before
- Determine MSI status based on average  $p(\text{MSI})$  across loci examined

# Probability-Based Calling (example)

- BAM file: TCGA-00-0000

- BAT-26
  - $p(\text{MSI}) = 0.998$
- MSI-07
  - $p(\text{MSI}) = 0.488$
- MSI-09
  - $p(\text{MSI}) = 0.478$
- H-06
  - $p(\text{MSI}) = 0.499$
- MSI-06
  - $p(\text{MSI}) = 0.898$
- MSI-04
  - $p(\text{MSI}) = 0.978$
- HSPH1-T17
  - $p(\text{MSI}) = 0.408$

- Locus-based call: MSS

- 3 MSI, 4 MSS

- Probability-based call: MSI

- Avg  $p(\text{MSI})$ : 0.678



# Probability-Based Calling (continued)

- Loci examined
  - BAT-26
  - MSI-07
  - MSI-09
  - H-06
  - MSI-06
  - MSI-04
  - HSPH1-T17
- Threshold: 0.500000
- Min no. loci: 3
- Total files: 93
- Correct predictions: 78
  - True pos: 23
  - True neg: 55
  - False pos: 1
  - False neg: 14
- **Accuracy: 0.838710**
- **Sensitivity: 0.621622**
- **Specificity: 0.982143**

# Probability-Based Calling (continued)

- Problem: some loci may be more indicative of MSI status than others
  - Need weights assigned to each locus individually

# A Bigger ML Problem

- Solution: instead of individually looking at one locus with  $m$  features, consider all  $n$  loci with all  $m$  features at once
  - Machine learning problem in  $n \times m$  dimensions
- Exclude loci with too few usable files

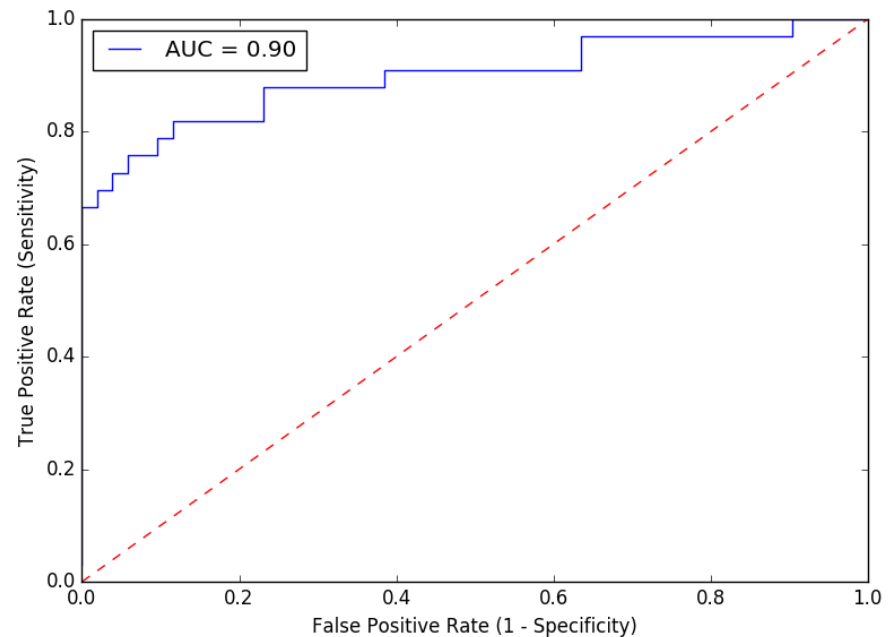
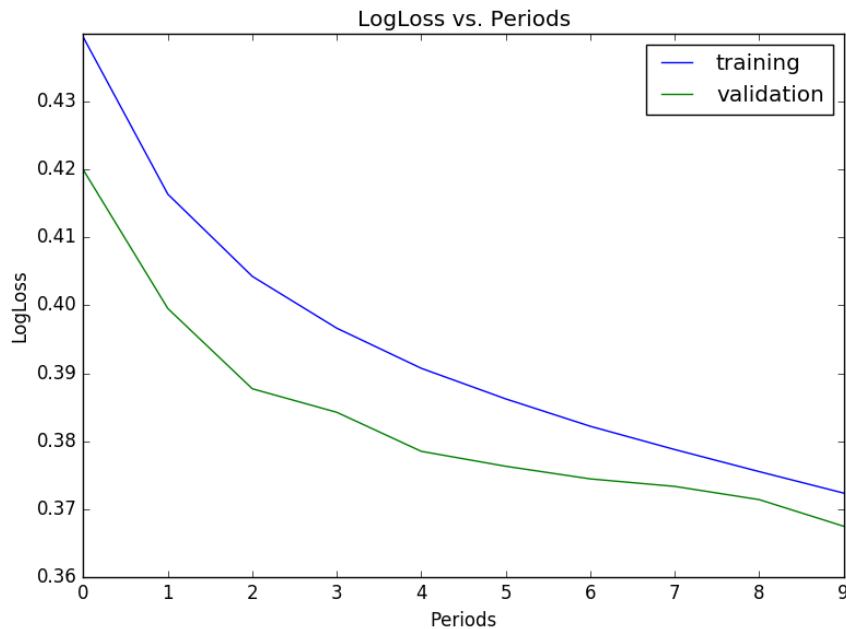
# Computational Cost

- 21 loci x 4 features each
  - 84-dimension model
- L1 regularization
  - Drives weights of non-influential features to 0
  - Minimizes the number of features that contribute to the model

# Machine Learning v2 - Sets

- Training Set
  - 124 MSI, 142 MSS
- Validation Set
  - 33 MSI, 52 MSS

# Machine Learning v2 - Results



# Machine Learning v2 - Test Set

- Loci examined

- MSI-11
- MSI-14
- H-10
- HSPH1-T17
- BAT-26
- BAT-25
- MSI-04
- MSI-06
- MSI-07
- MSI-01
- MSI-03
- MSI-09
- H-09
- H-08
- H-01
- H-03
- H-02
- H-04
- H-07
- H-06
- H-05

- Threshold: 0.500000
- Total files: 85
- Correct predictions: 73
  - True pos: 27
  - True neg: 46
  - False pos: 6
  - False neg: 6
- **Accuracy: 0.858824**
- **Sensitivity: 0.818182**
- **Specificity: 0.884615**

# Machine Learning v2 - Test Set

- Loci examined
  - MSI-11
  - MSI-14
  - H-10
  - HSPH1-T17
  - BAT-26
  - BAT-25
  - MSI-04
  - MSI-06
  - MSI-07
  - MSI-01
  - MSI-03
  - MSI-09
  - H-09
  - H-08
  - H-01
  - H-03
  - H-02
  - H-04
  - H-07
  - H-06
  - H-05
- Upper threshold: 0.550000
- Lower threshold: 0.450000
- Total files: 85
- Indeterminate files: 9
- Predictions: 76
- Correct predictions: 67
  - True pos: 25
  - True neg: 42
  - False pos: 3
  - False neg: 6
- **Accuracy: 0.881579**
- **Sensitivity: 0.806452**
- **Specificity: 0.933333**
- **Indeterminate: 10%**



# Summary

- Maximum accuracy of 88% on test set
  - 10% labeled as 'Indeterminate'
- Machine learning using all loci and all features is most:
  - Accurate
  - Sensitive/Specific

# Further Investigations

- Add a label for 'MSI-L' and 'MSI-H'
- Research clinical relevance of MSI diagnosis
  - Fine-tune parameters to favor either false positives or false negatives
- Add features
  - Earth mover distance
- Regularization
  - L2 regularization
    - Penalizes any weight that is too large, drives weights asymptotically to 0
  - L1 regularization
    - Penalizes total weight of all features, drives non-informative weights to 0

# Further Investigations (continued)

- Get more BAM files
  - Training, validation and test sets are stale
  - Get BAM files with better coverage for more accurate results, better prediction
- Make a more complex model
  - Synthetic features
  - Feature crosses
  - Neural Nets
    - More complex models require additional samples to train, validate and test