Mitotic Count Simulation

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Clinical area: Digitization of Pathology, Division: DIDSR/OSEL/CDRH



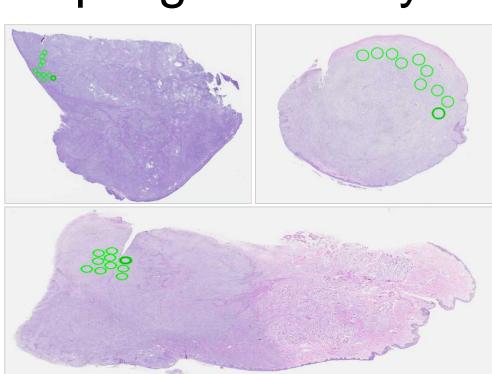
Regulatory Relevance

Shapes digital pathology's pathway to market by comparing pathologists' performance with digital images to the microscope.

Summary

We are generating a number of Monte Carlo (MC) trials. In each trial, there are a number of readers and a number of cases. Each case has a number of locations of interests. The readers provide counts to each case location on multiple modalities. We analyze the simulation results to understand the sampling variability of the agreement between survival time and counts and to evaluate novel variance of the sampling variability.







Simulation Configuration (Inputs)

Size: number of MC trials, readers, cases, locations per cases **Model Parameters:**

- > mean survival time, mean censoring time, mean counts (negative exponential distributions)
- > variances describing random effects for readers, cases, locations, and interactions (zero-mean normal distributions)

for all the patients

event = censored; if S > C

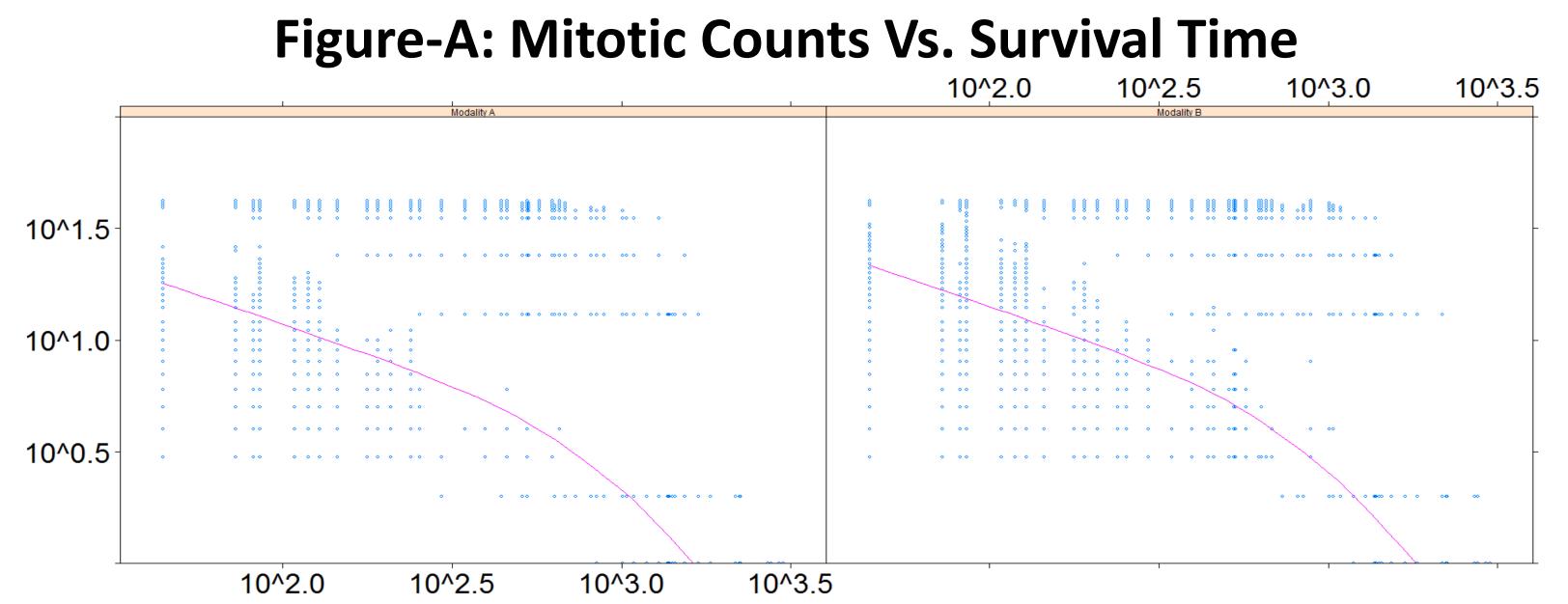
event = death;

Simulation Outputs for One Trial

- Survival time
- > Censoring time
- > Noise: the sources are readers, cases, modalities and interactions, also referred to as random effects
- > Mitotic counts: given by readers to each case-location
- > Reader-averaged agreement measure: between survival times and counts (correlation, concordance, agreement)
- > Variance of reader-averaged agreement measure (new methods to be evaluated)

Visualization of One Trial: Mitotic Counts Vs. Survival Time

readers = 50, cases = 50, locations = 10, mean survival time = 1000 days, mean censoring time = 500 days mean counts A = 5, mean counts B = 6, all variance components = 0.1



of Reader Averaged

Agreement Measure

Visualization

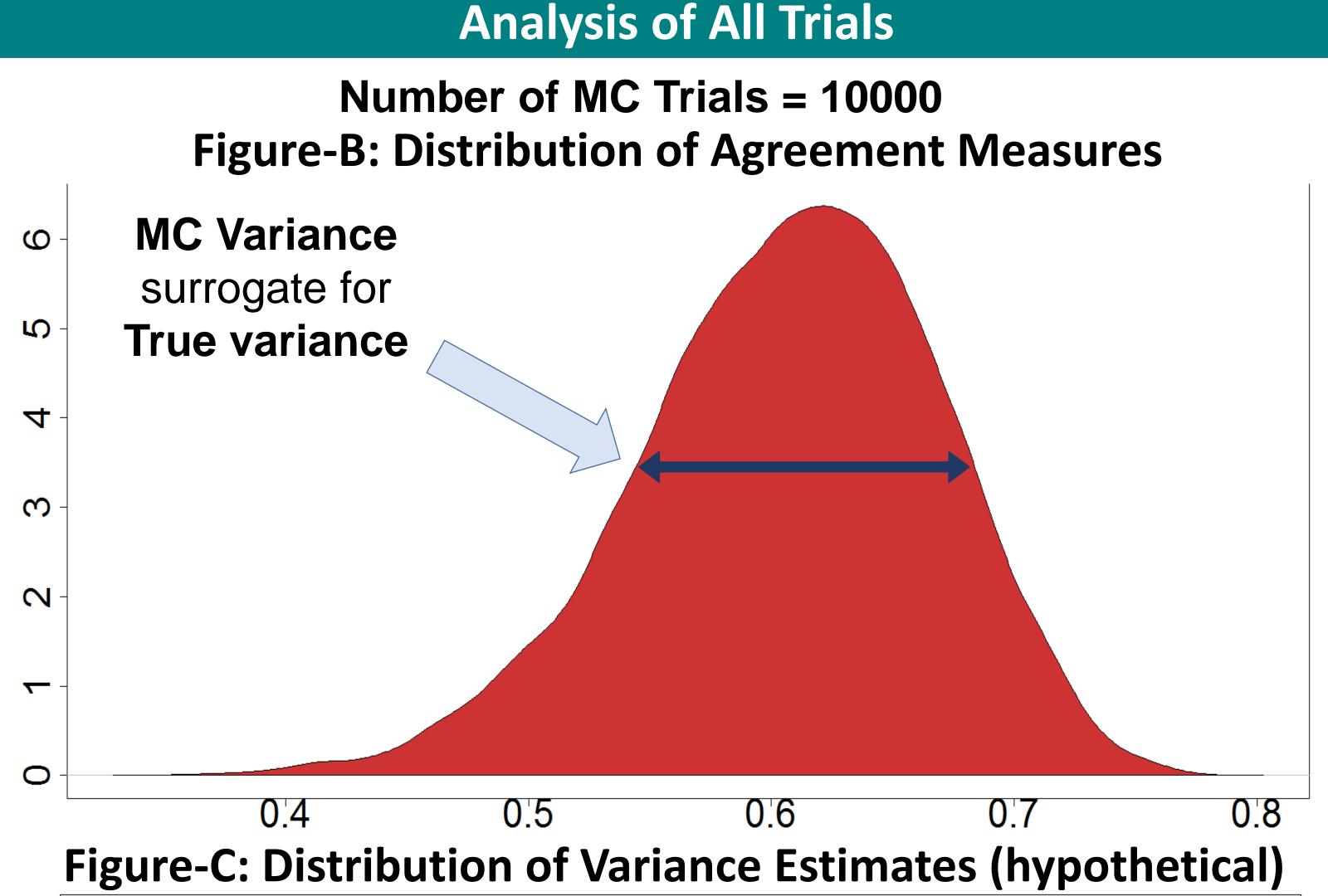
Simulation Flow for One Trial Simulate Cases Realize Mitotic Counts Simulate Effects Log Counts Survival Time (S) Censoring Time (C) **Modality B No-Modality** Modality A Modality A = Number of days the patient survived Days since the last visit Log(Mean Counts) Random Random Random + No-modality effects Effects Effects + ModalityA effects Reader Reader Reader **Histogram of Censoring Time Histogram of Survival Time** Case Log Counts Case Modality B = Case-Case-Log(Mean Counts) Case-Location Location Location + No-modality effects + ModalityA effects Reader-Reader-Reader-Case Reader-Reader-Reader-Calculate Reader Case-Case-Averaged Agreement Location Location Location Measure Generate event Estimate Variance censoring Fixed Effects

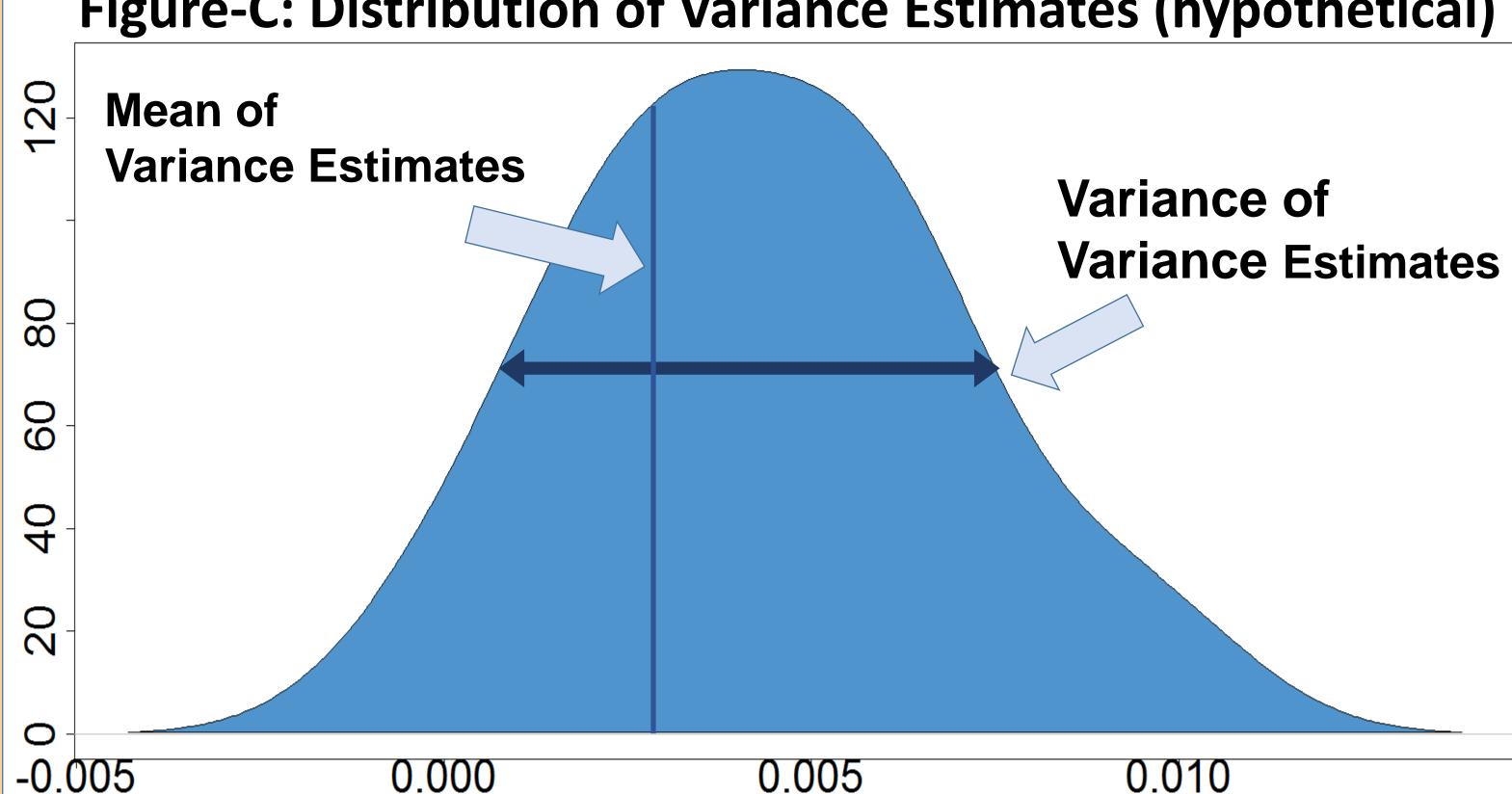
C = 0

Fixed Effects

Mean

Mean





Evaluate Variance Estimators:

- > Bias = Difference between true variance and mean of variance estimates
- > Precision = Variance of variance estimates

Results and Future Work

- > The simulation allows us to understand the sampling variability of the agreement between survival time and counts (Figure-A) and to evaluate novel variance estimates of the sampling variability (Figure A&B).
- > Code is written in R, runs on R Console, linux command line, cloud (NCIP hub access on request)
- > Source and documentation available at (private project until release) https://github.com/DIDSR/mitoticCountSimulation
- > In future, we want to parallelize the code and develop a GUI for a better user experience

Acknowledgement

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