

A COMPARATIVE STUDY OF PARAMETRIC AND AI-ENHANCED NONPARAMETRIC RELIABILITY ESTIMATORS IN ONCOLOGY SURVIVAL MODELS

Authors:

Lujain Belal Al-Smadi
Yazan Nazih Ahmad Haj-Hussein
Qusai Saleh Mansour Shakhatreh
Dr. Mohammad Ahmed Orsan Obeidat
Dr. Abdul-Aziz Adnan Chrit



Outline

1.
Introduction

4.
Methods

7.
**Weibull
Results**

10.
Conclusion

2.
**Measurement
Requirement**

5.
**Simulation
Setup**

8.
**MSE
Summary**

11.
References

3.
**Motivation &
Gap**

6.
**Exponential
Results**

9.
Real Data



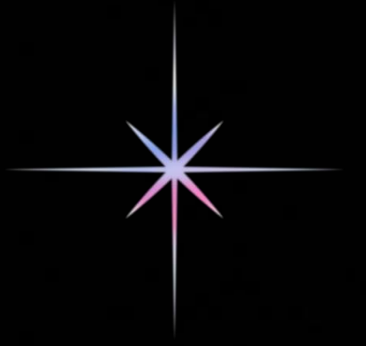
Introduction

Reliability in Medical Context

- $R = P(\text{Strength} > \text{Stress})$ $R = P(X > Y)$
- $X = \text{strength (patient survival time)}$ $Y = \text{stress (risk / treatment burden)}$
- Classical in Engineering, Evolving in Medicine
- AI-enhanced survival modeling = better decisions
- **Applications of R:**
 - Engineering: material strength, mechanical failure
 - Medicine: comparing survival time vs treatment burden



Measurement Requirement



X and Y must be in the same unit

Examples:

Time vs Time

Dose vs Dose

Pressure vs Pressure



Motivation & Research Gap



1. Classical methods assume specific distributions
2. Real survival data violates assumptions
3. AI methods add flexibility
4. Need a combined hybrid approach



Practical Limitation (Middle East Servers)

- DeepSurv and several survival AI models cannot run on most Middle Eastern servers
- They require U.S./European GPU-based cloud infrastructure
- Therefore, we used methods that can run locally : RSF (ranger)— GBM—Cox

Methods Overview (Merged)

✦ Empirical (Classical)

ECDF(X) + Kernel PDF(Y)

✦ RSF (AI, ranger)

Survival trees

Handles nonlinear patterns

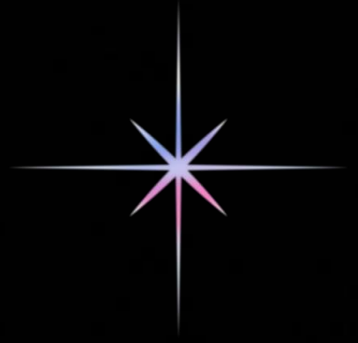
✦ GBM–Cox (AI)

Boosted Cox

Strong predictive performance



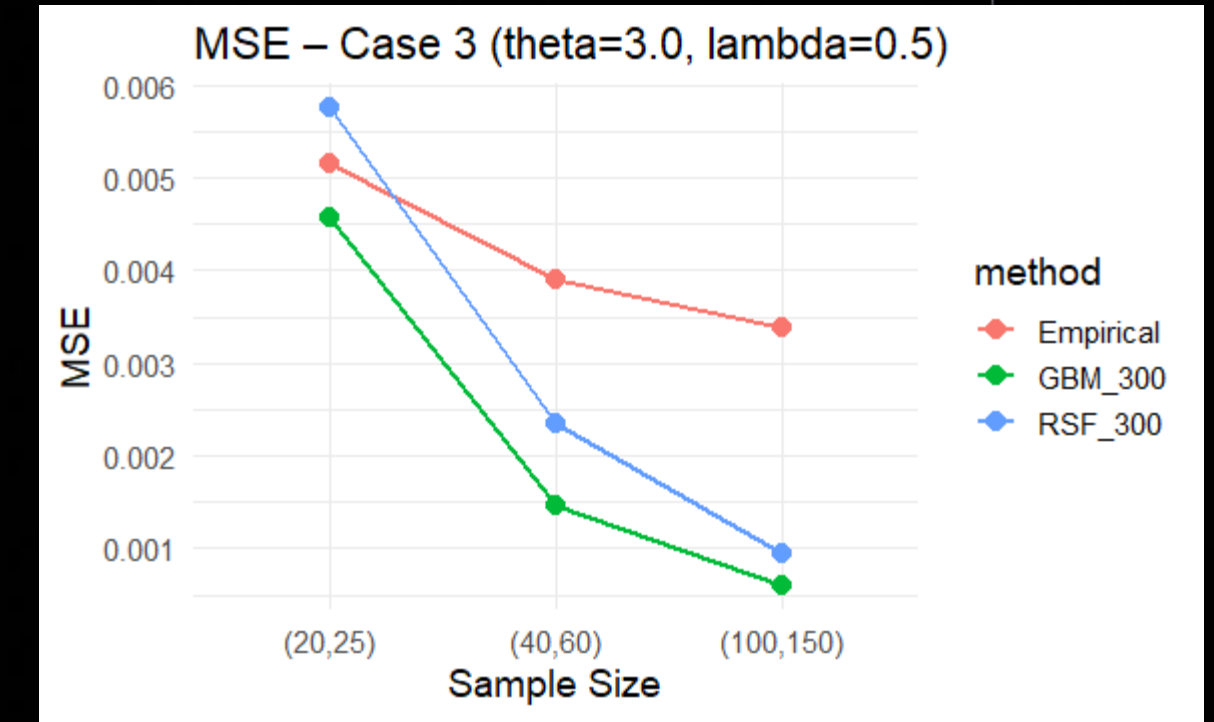
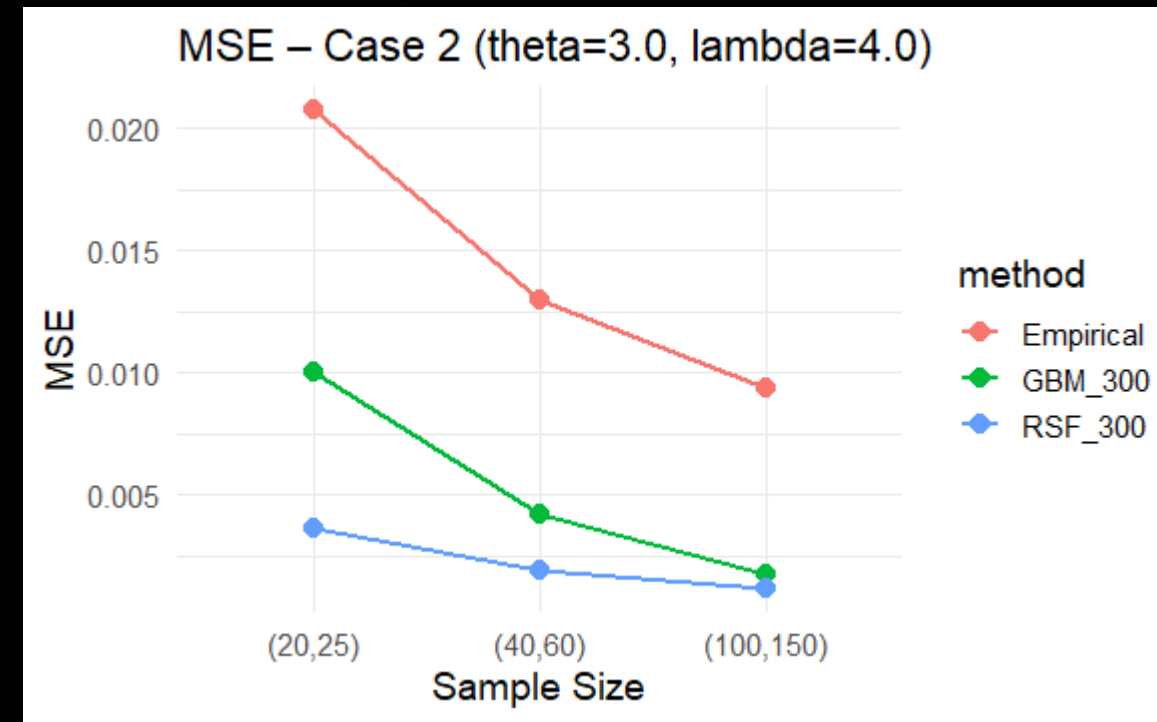
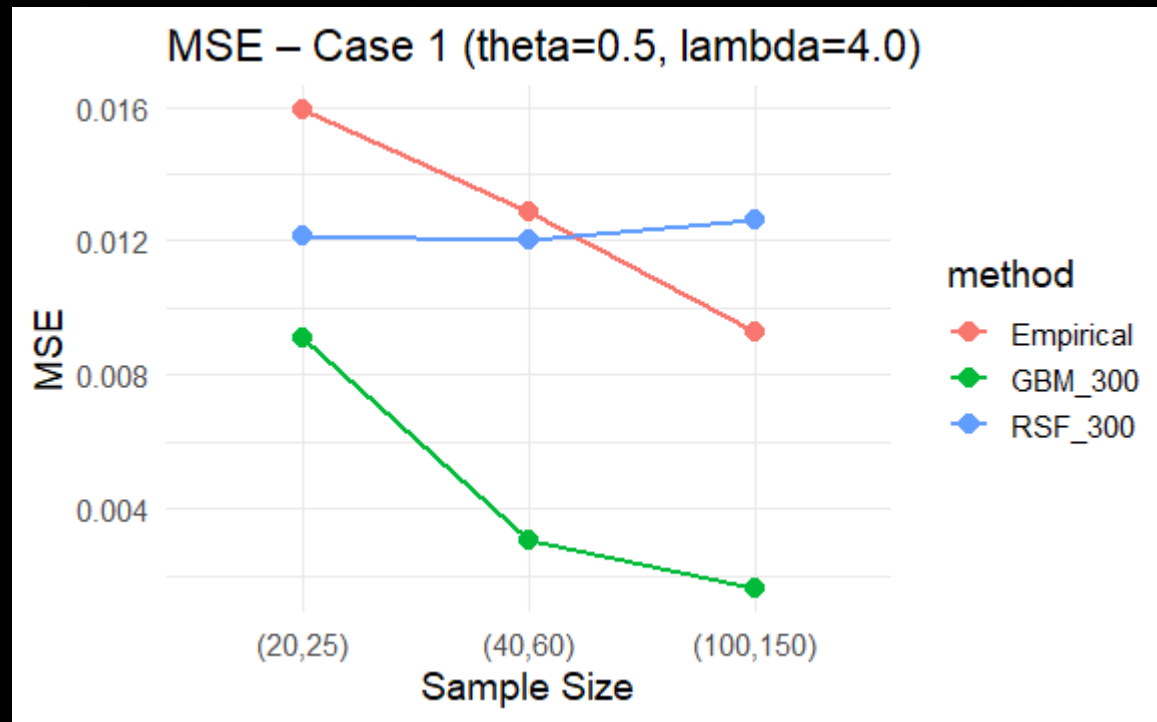
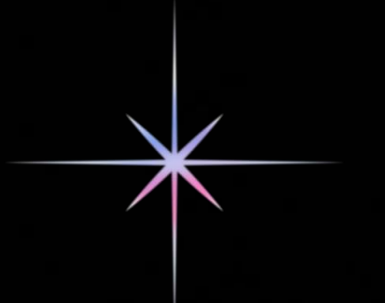
Simulation Setup



- Distributions: Exponential & Weibull
- Three sample sizes
- Replications: N simulations per case
- AI models use 300 trees
- Evaluation metric: MSE only



Exponential Results (MSE)

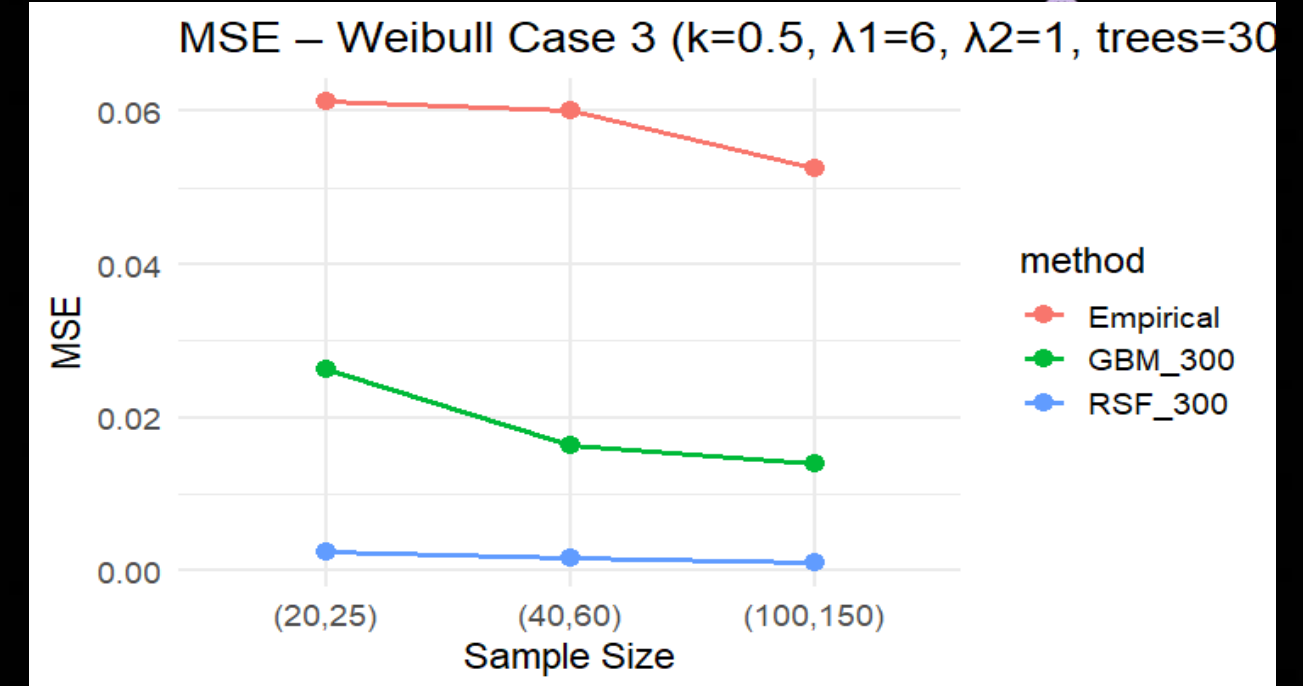
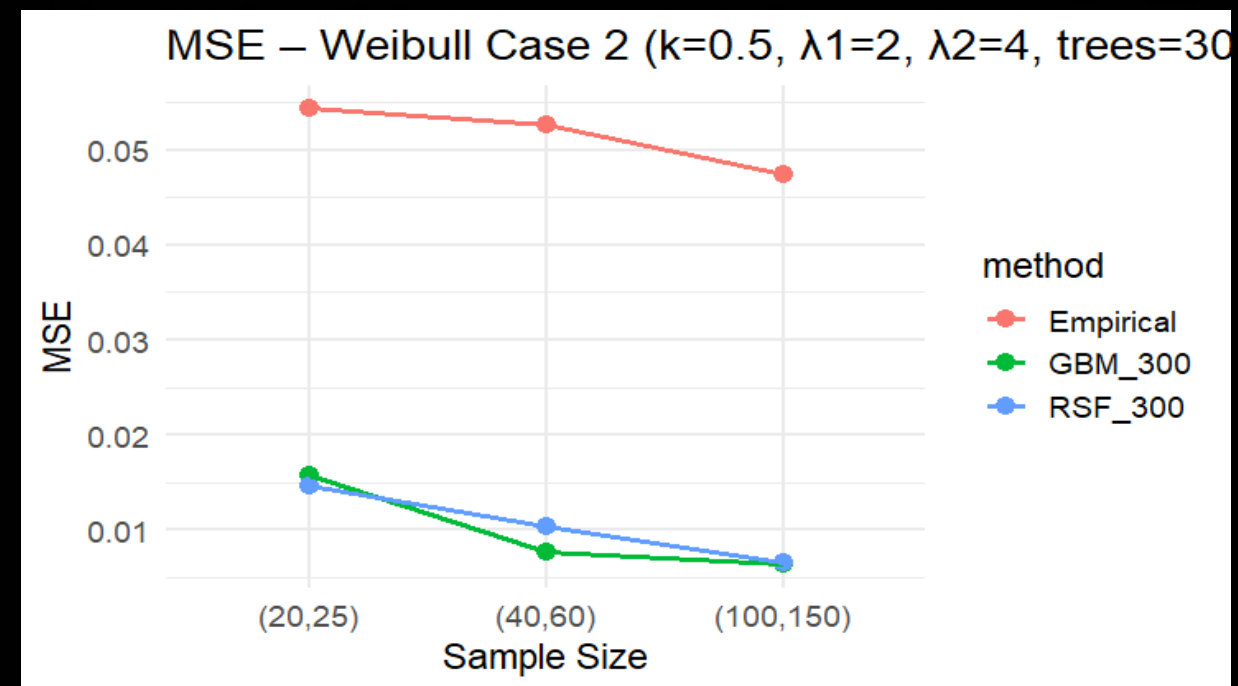
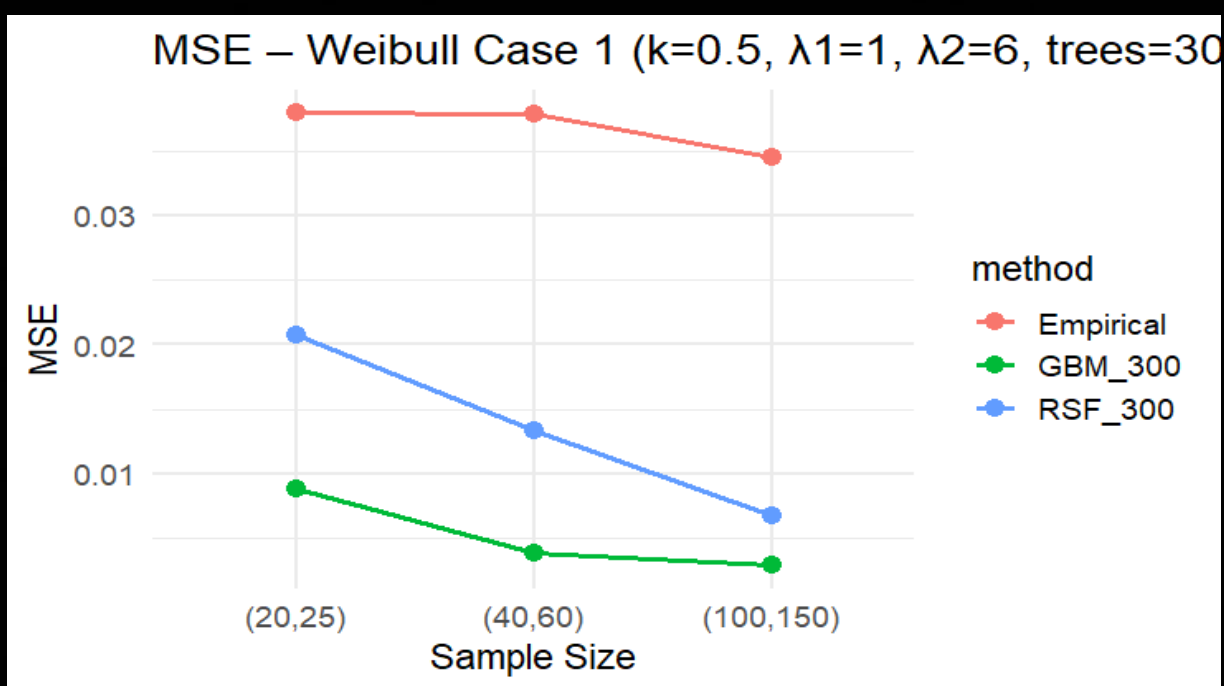


60

- Empirical = highest MSE
- GBM best in Cases (1 & 3)
- RSF best in Case 2
- MSE ↓ with larger samples



Weibull Results (MSE)



- Empirical = worst
- GBM best in Cases 1 & 2
- RSF best in Case 3

MSE Summary (Exponential + Weibull)



Model	Case	Empirical MSE	GBM-300 MSE	RSF-300 MSE	Best
Exponential	Case 1 ($\theta=0.5$, $\lambda=4.0$)	0.0159 $\rightarrow$ 0.0093	0.0091 $\rightarrow$ 0.0016	0.0121 $\rightarrow$ 0.0126	★ GBM-300
Exponential	Case 2 ($\theta=3.0$, $\lambda=4.0$)	0.0207 $\rightarrow$ 0.0093	0.0100 $\rightarrow$ 0.0017	0.0036 $\rightarrow$ 0.0011	★ RSF-300
Exponential	Case 3 ($\theta=3.0$, $\lambda=0.5$)	0.00514 $\rightarrow$ 0.00337	0.00456 $\rightarrow$ 0.00059	0.00575 $\rightarrow$ 0.00095	★ GBM-300
Weibull (k=0.5)	Case 1 ($\lambda_1=1$, $\lambda_2=6$)	0.035–0.038	0.0029–0.0089	0.0067–0.0207	★ GBM-300
Weibull (k=0.5)	Case 2 ($\lambda_1=2$, $\lambda_2=4$)	0.047–0.054	0.0063–0.0157	0.0065–0.0147	★ GBM-300
Weibull (k=0.5)	Case 3 ($\lambda_1=6$, $\lambda_2=1$)	0.052–0.061	0.0139–0.0263	0.00098–0.00248	★ RSF-300



As sample size increases, RSF and GBM-Cox outperform the Empirical method

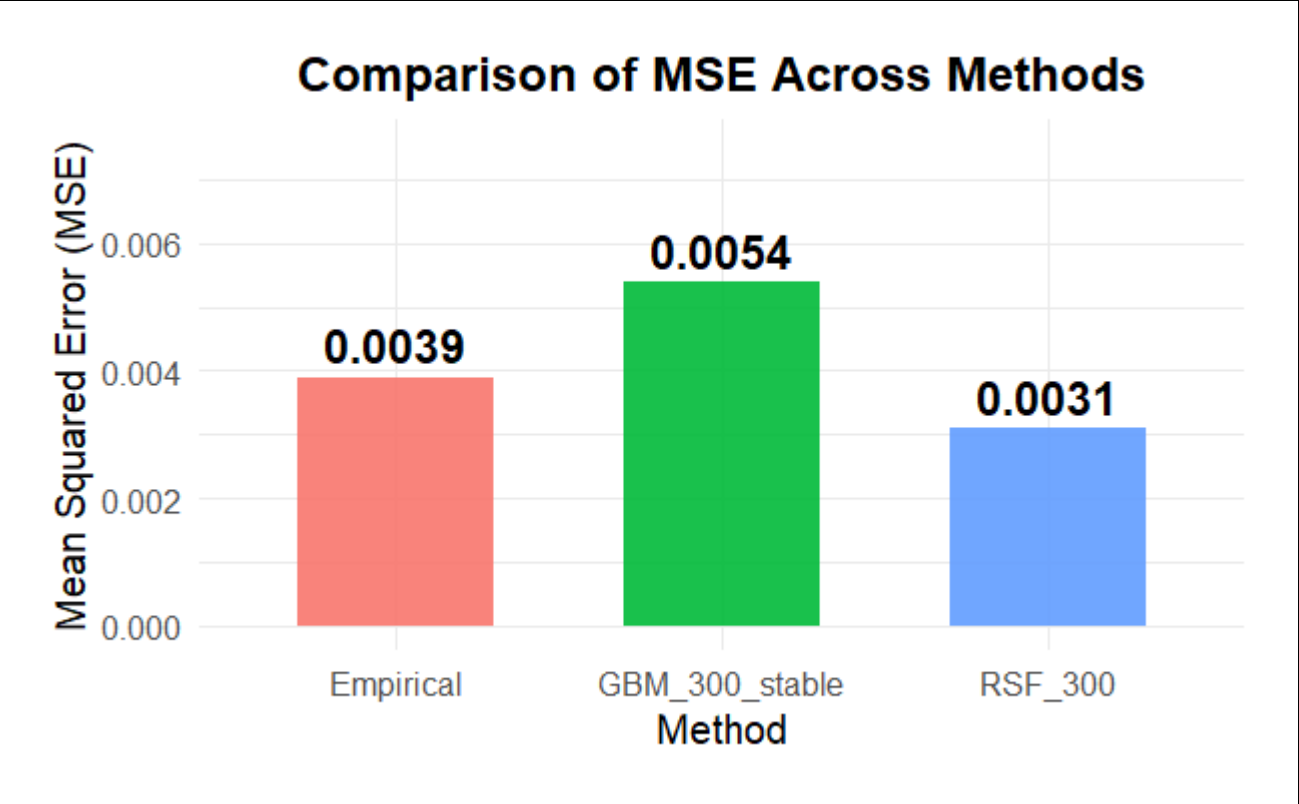
Real Data & Conclusion

- Collected manually from King Abdullah University Hospital (KAUH)
- Data extraction completed by:
 - Lujain Al-Smadi
 - Yazan Haj-Hussein
 - Qusai Shakhatreh
- Supervised by:
 - Dr. Hani Taani
 - Dr. Abdul-Aziz Chrit
- X = patient survival time
- Y = treatment burden / disease severity



Phase 1: True R (Full Dataset)

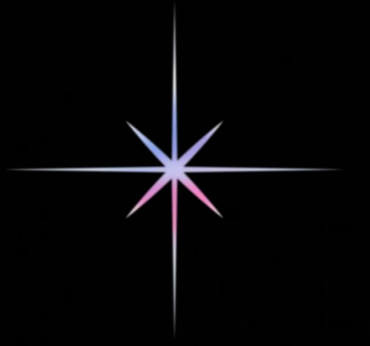
Method	True R
Empirical	0.3556
GBM_300_stable	0.3788
RSF_300	0.3754



Phase 2: Sampling-Based Evaluation (Bias, MSE, Variance)

Method	R_mean	Bias	MSE	Variance
Empirical	0.3575	0.0019	0.00385	0.00385
GBM_300_stable	0.4022	0.0234	0.00536	0.00482
RSF_300	0.3908	0.0154	0.0031	0.00287

Conclusion



- AI methods outperform the classical Empirical estimator
- GBM–Cox shows the strongest overall performance
- RSF performs best when X and Y distributions differ
- MSE decreases consistently with larger samples
- AI-enhanced survival modeling improves medical decision-making

References

- Meeker, W. Q., & Escobar, L. A. (1998).
Statistical methods for reliability data. Wiley.
- de la Cruz, R., Salinas, H. S., & Meza, C. (2022).
Reliability estimation for stress–strength model based on unit half normal distribution. *Symmetry*, 14(4), 837.
- Liao, Y., Pan, X., Zhang, Z., Wang, Y., & Zhang, Y. (2024).
Random survival forest algorithm for risk stratification and survival prediction in gastric neuroendocrine neoplasms. *Scientific Reports*, 14, 77988.

THANK YOU!

Any Question ?

LUJAINSMADI6@GMAIL.COM