

COST-EFFECTIVE CHART REVIEW SAMPLING DESIGN TO ACCOUNT FOR PHENOTYPING ERROR IN ELECTRONIC HEALTH RECORDS (EHR) DATA

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Dec 06, 2022

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INTRO

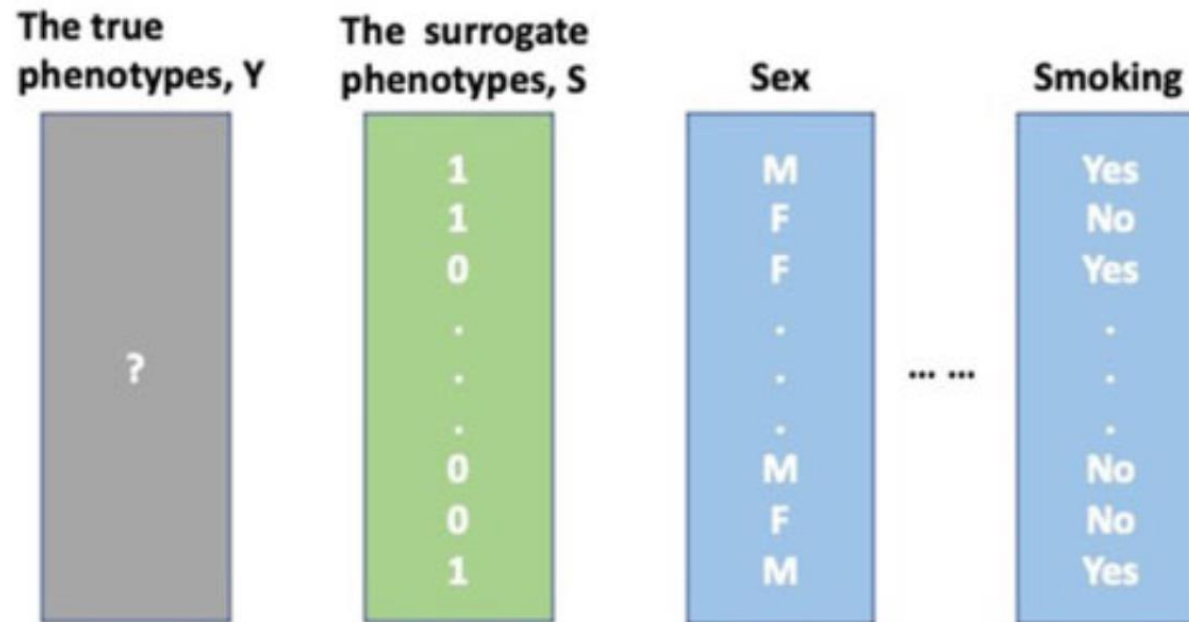
Introduction and Motivation

- Electronic health record (EHR) data
 - A digital data system that contains medical history of patients
 - Investigates the association between risk factors and EHR-derived phenotype (binary)
- Problems:
 - EHR-derived phenotypes are likely to be misclassified; a manual chart review is required
 - Chart review can only be conducted for a validation subcohort due to limited resources
 - For rare diseases, positive phenotype patients in full cohort is even fewer
 - chart review may loss efficiency
- Research goal:

investigate an efficient sampling procedure for association study when there is **a small number of positive phenotype subjects** in the full cohort

EHR Data

The original full cohort with unknown Y



Relationship between Y and X: $\text{logit}\{P(Y_i = 1|\mathbf{X}_i)\} = \mathbf{X}_i\beta$

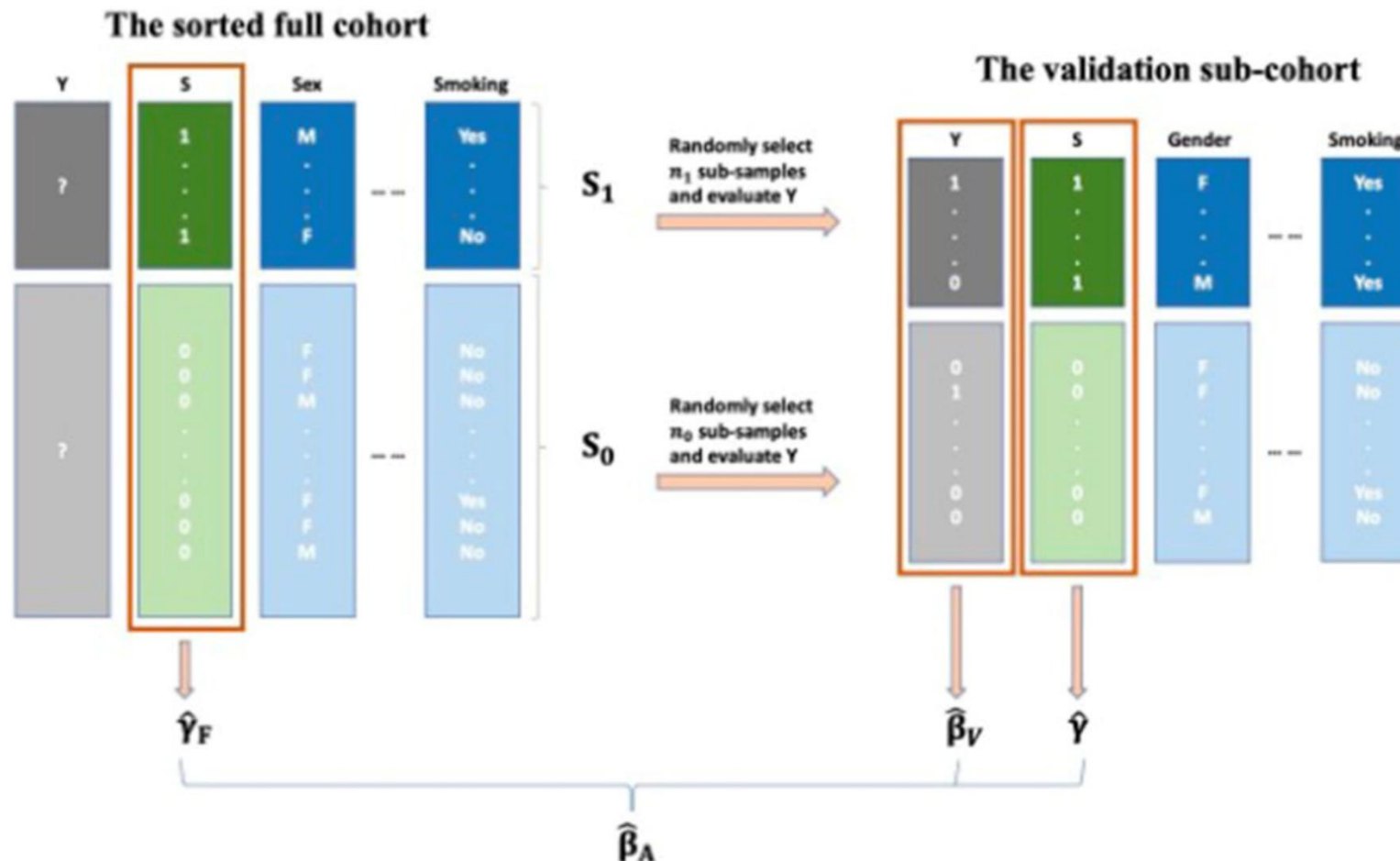
Previous Works

- 1. Original uniform (Ori-Unif) sampling
 - Cannot guarantee a balanced number of positive and negative phenotype patients in the subcohort
- 2. Original biased (Ori-Bias) sampling
 - Only explores the relationship between true phenotype and covariates but fails to take information in surrogate phenotype into account.
- 3. Augmented uniform (Aug-Unif) sampling
 - Introduces information in surrogate phenotype into the estimator
 - But uses a uniform sampling procedure

METHODS

Proposed Method

- Outcome-dependent sampling design for cost-effective chart review with augmented estimation procedure (OSCA)



OSCA

- Assumption: surrogate phenotype and the true phenotype are non-differentially associated
 - Fixed probabilities of making a correct chart review conditional on the observed phenotype

$$p_1 = P(S_i = 1|Y_i = 1) \text{ and } p_0 = P(S_i = 0|Y_i = 0).$$

- Improvement of OSCA
 - Fewer subjects are required for chart review to achieve the same statistical power
 - Smaller estimated standard error; higher statistical efficiency

RESULTS

RESULTS

- 3 Simulations
 - Own synthetic data sets
- 1 real data set
 - Real data: Colon cancer recurrence EHR data set in the KPW healthcare system
 - Proved closer estimates to the golden standard

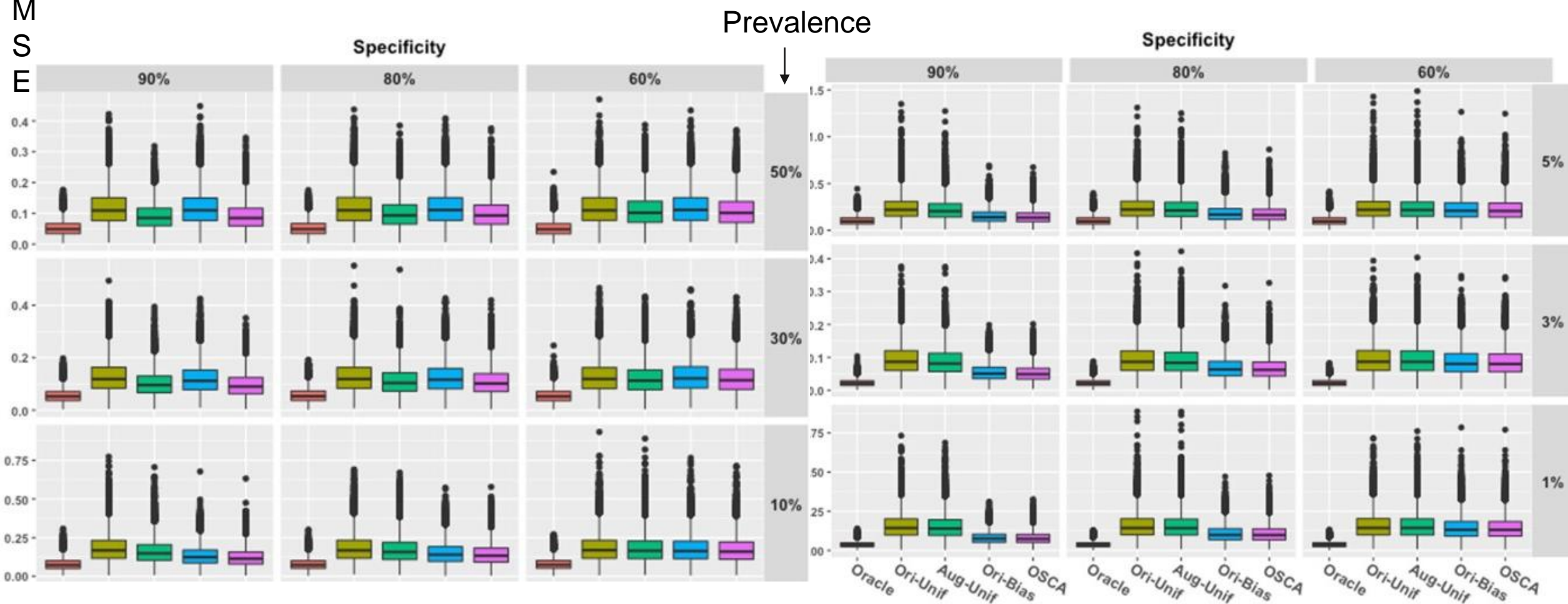
SIMULATION1: DATA GENERATION

EMPIRICAL COVERAGE PROBABILITIES AND CONFIDENCE INTERVALS

- $\text{logit}\{P(Y_i = 1|X_1, X_2, X_3)\} = b_0 + X_1 + X_2 + X_3$
- Prevalence: $\{\sim 1\%, \sim 3\%, \sim 5\%, \sim 10\%, \sim 30\%, \sim 50\%\}$
 - For $\sim 5\%, \sim 10\%, \sim 30\%, \sim 50\%$: $N=2000$, $n=600$
 - For $\sim 1\%, \sim 3\%$: $N=8000$, $n=2000$

SIMULATION 1: RESULTS

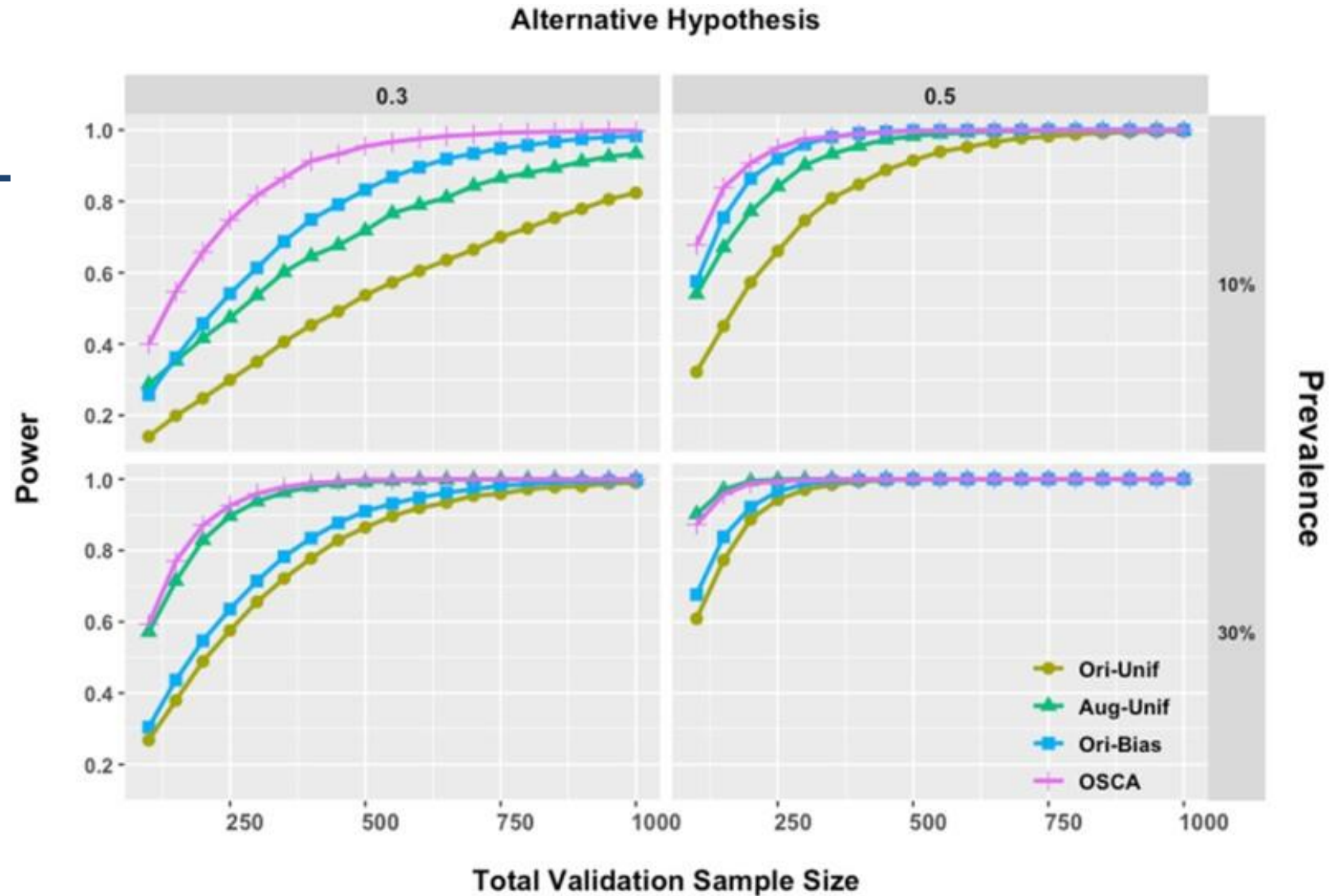
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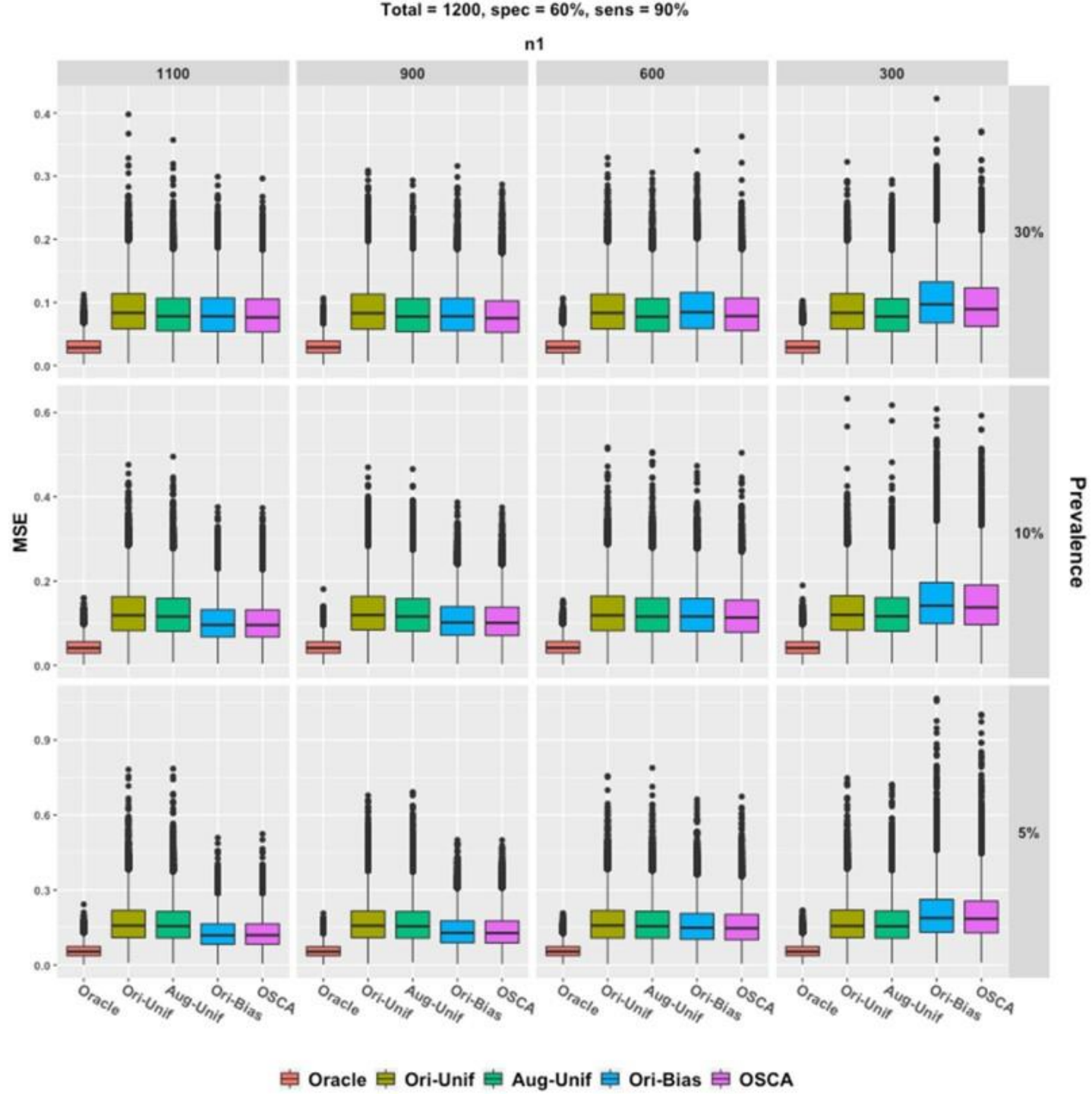
SIMULATION 2:

POWER ANALYSIS

- $\text{logit}\{P(Y_i = 1|X)\} = b_0 + \beta_1 X$
- $H_0: \beta_1 = 0$
- $H_1: \beta_1 = 0.3 \text{ or } 0.5$



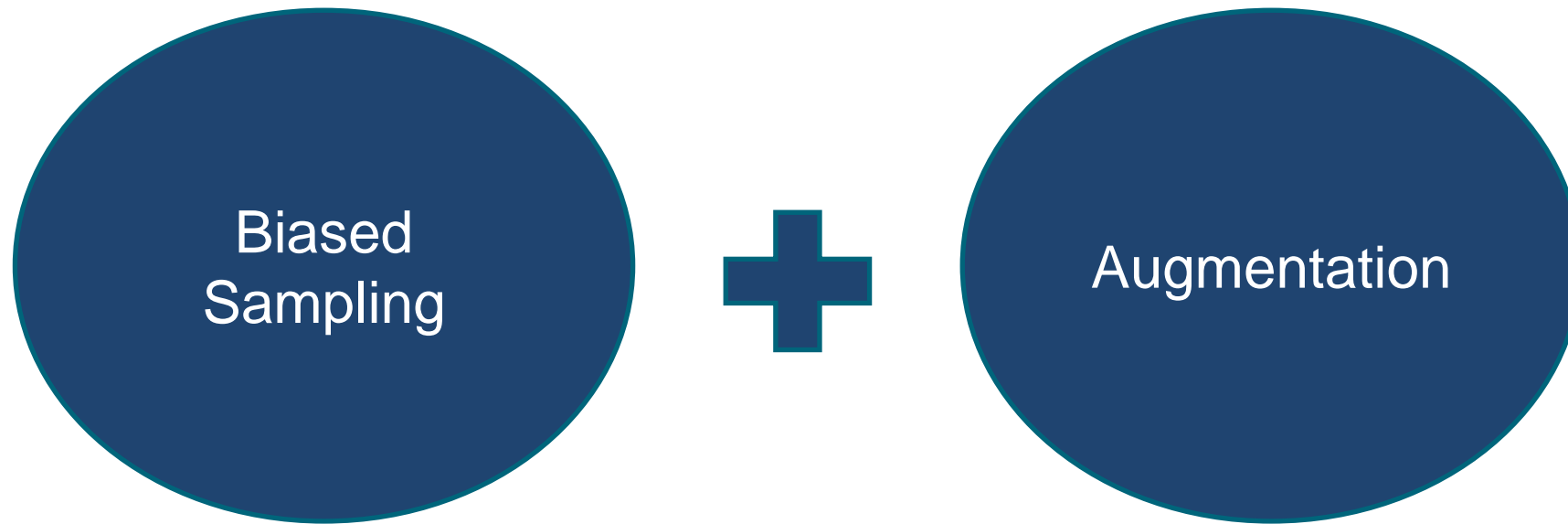
SIMULATION 3: EFFECT OF IMBALANCED SAMPLING



CONCLUSION

PROBLEMS?

- EHR data: an electronic medical records of patients
- phenotyping error in EHR-derived outcomes
 - Systematic bias
- a manual chart review is required
 - Time-consuming



CONCLUSION

- Simulation 1: concentrated MSE box especially when disease is rare
- Simulation 2: Need smaller sub-cohort to reach the same level power
- Simulation 3: validated effectiveness of biased sampling
- In conclusion
 - reduces estimation bias while maintaining low variance
 - Is cost-effective in chart review.

Thank You!