

Statistical considerations for classifying radiologists relative to targets for screening mammography interpretive performance

Rebecca Hubbard

Associate Professor

Department of Biostatistics & Epidemiology

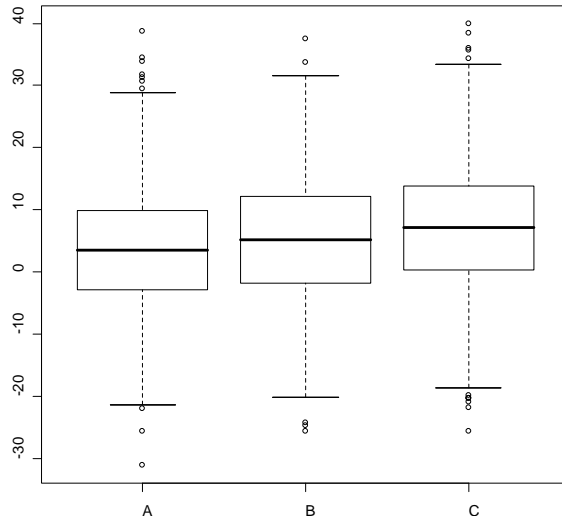
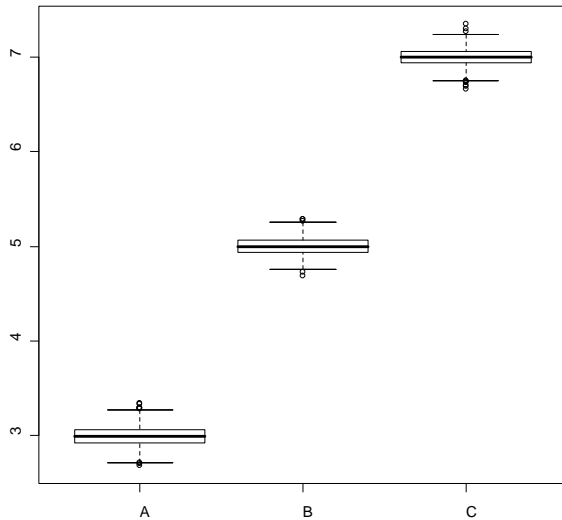
May 12, 2015



Are radiologists meeting targets for acceptable interpretive performance?

- ◆ Targets have been established for interpretive performance
- ◆ We would like to identify radiologists who are/are not meeting these targets
- ◆ Is this an achievable goal given the data that we have available?
- ◆ This is a **profiling** task with objective of classifying radiologists relative to a fixed benchmark

Challenges to estimating radiologist performance



- ◆ **Correct classification depends on our ability to distinguish between radiologists with differing performance**
 - Requires more variability between radiologists and less variability within radiologists
- ◆ **Within-provider (error) variation may be large due to small provider volume**
- ◆ **Reliability = between-provider variability/total variability**
 - Reliability >0.9 generally considered to be necessary for “high stakes” profiling (e.g., public reporting)

Objective

- ◆ Discuss relevant statistical considerations for identifying radiologists failing to meet targets for interpretive performance
- ◆ Use simulations to demonstrate performance of classification for recall and cancer detection rate
- ◆ Demonstrate how these results are modified by use of imperfect proxy for performance measures based on claims data

Conceptual framework

- ♦ **Binary event of interest observed for each patient**
 - Recall
 - Screen-detected cancer
- ♦ **Each provider has underlying, true performance**
 - Unobservable without complete data on entire patient population
 - Objective of profiling is to make inference on performance based on a finite sample

Profiling methods

♦ **Classification based on point estimate**

- For each radiologist compute rate or proportion
- Compare to guideline target and make classification
- Can be adjusted to account for differences in patient population (case-mix) using regression methods
- May be unstable for small patient volumes

♦ **Classification based on confidence intervals**

- Compute confidence interval around point estimate
- If confidence interval lies completely below/above target then classify as failing to meet target
- Desired precision can be tuned by varying confidence level
- Addresses instability in estimates for small volume providers
- May be overly conservative

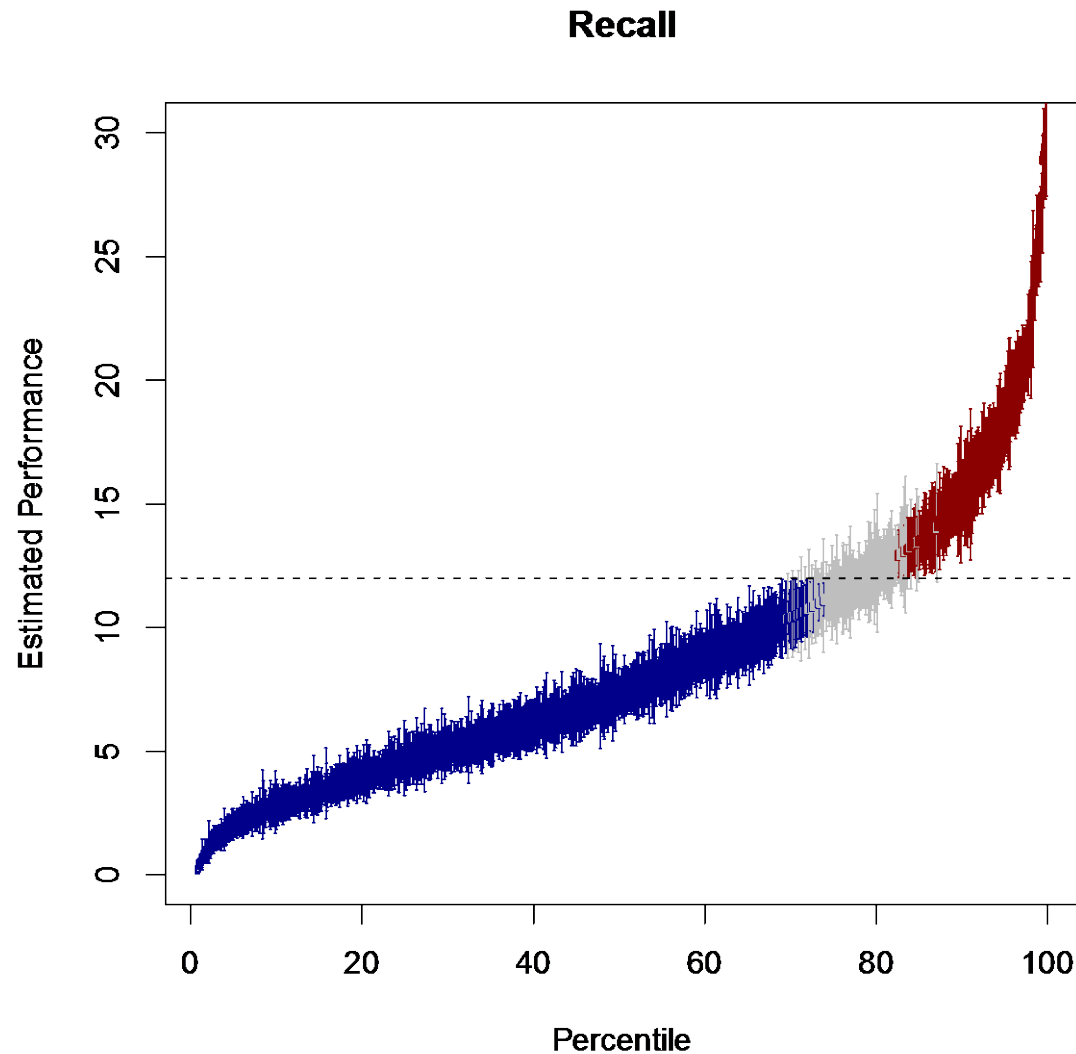
Simulation study design

- ◆ We conducted a statistical simulation study to demonstrate performance of classification relative to guideline targets for recall and cancer detection rate using point estimates and CIs
- ◆ Simulation study parameters chosen to generate data following real-world distribution of radiologist screening volume, CDR and recall
- ◆ Performance of classification evaluated in terms of sensitivity
 - **Sensitivity** = Proportion of radiologists failing to meet target successfully identified as failing to meet target
 - **Specificity** = Proportion of radiologists meeting target successfully identified as meeting target
- ◆ Evaluate sensitivity and specificity for
 - CDR relative to threshold of 2.5 per 1000
 - Recall relative to threshold of 12%

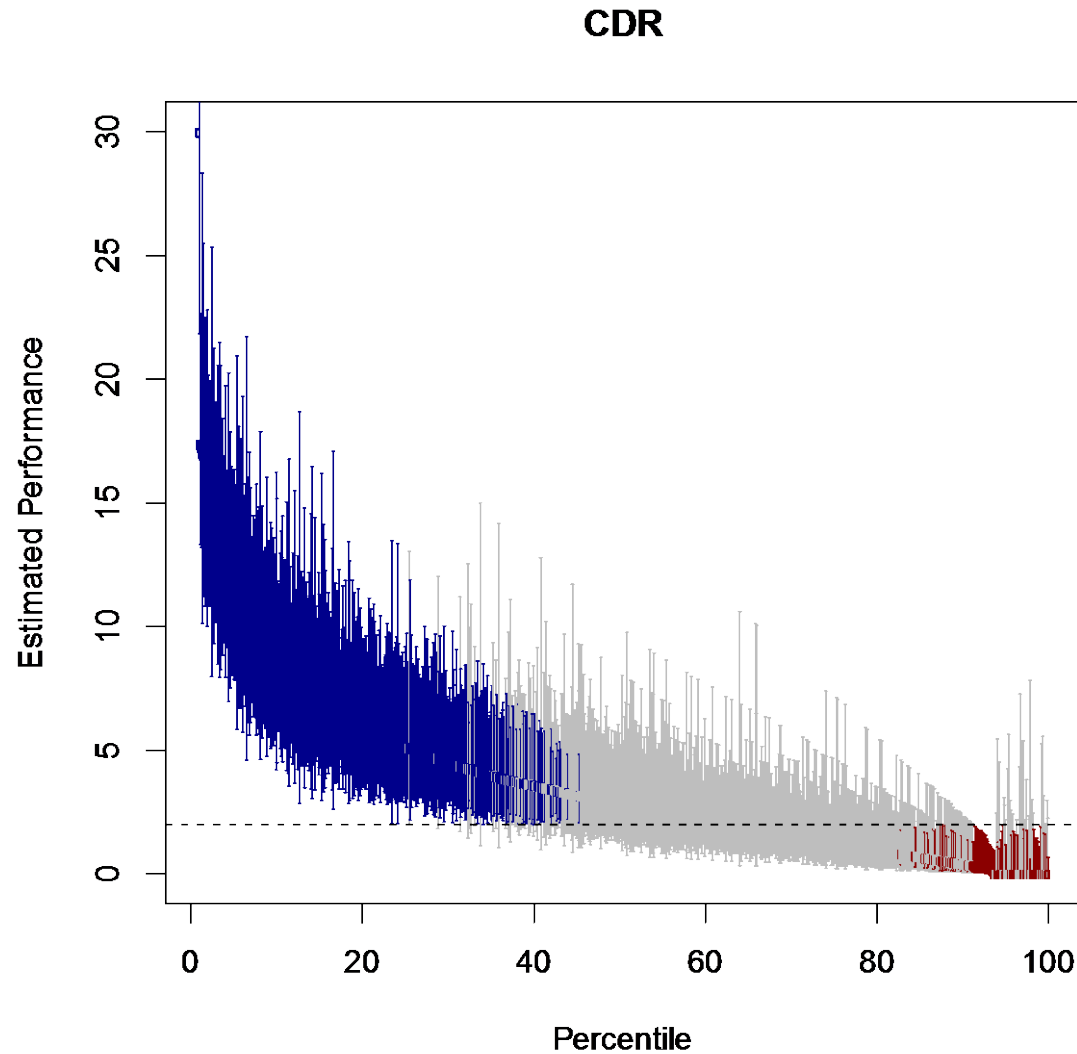
Simulation study design

- ♦ Average of $N = 1000$ radiologists per simulation
- ♦ Patient volumes $\sim \text{Gamma}(3.4, 458.4)$, truncated at 480
 - Mean = 1557, SD = 845
- ♦ True radiologist performance measures
 - Recall $\sim \text{Beta}(2.5, 26.4)$
 - Mean = 8.5%, SD = 5.1%
 - Reliability = 0.981
 - CDR $\sim \text{Beta}(1.36, 372.69)$
 - Mean = 3.6/1000, SD = 3.1/1000
 - Reliability = 0.807
- ♦ Repeat simulations 1000 times

Recall classification for simulated population



CDR classification for simulated population



Sensitivity and specificity for radiologist classification

	Point estimate		95% CI	
	Sens	Spec	Sens	Spec
Recall	94.8	98.4	76.4	99.9
CDR	86.4	88.7	22.3	99.9

Considerations under outcome misclassification

- ◆ We also explored classification based on Medicare claims
- ◆ This introduces an additional challenge since classification of an exam as resulting in recall or CDR is imperfect
- ◆ Algorithm operating characteristics for proxy recall and CDR are known
 - Sensitivity: probability of event based on claims given truly was an event
 - Specificity: probability of no event based on claims given truly was no event
- ◆ How does using an imperfect proxy for outcomes affect radiologist classification relative to targets?

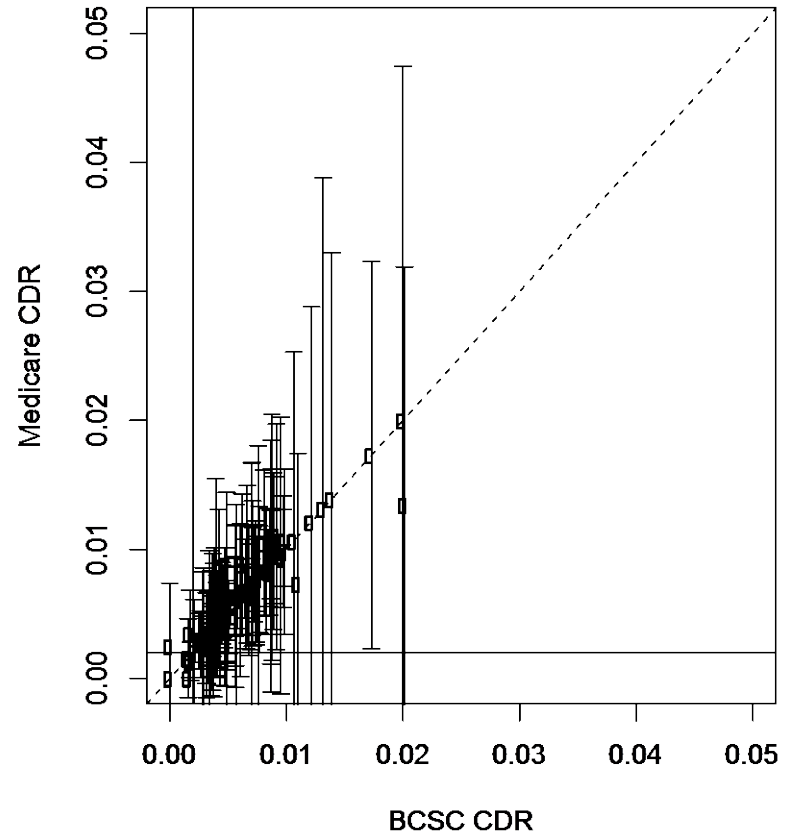
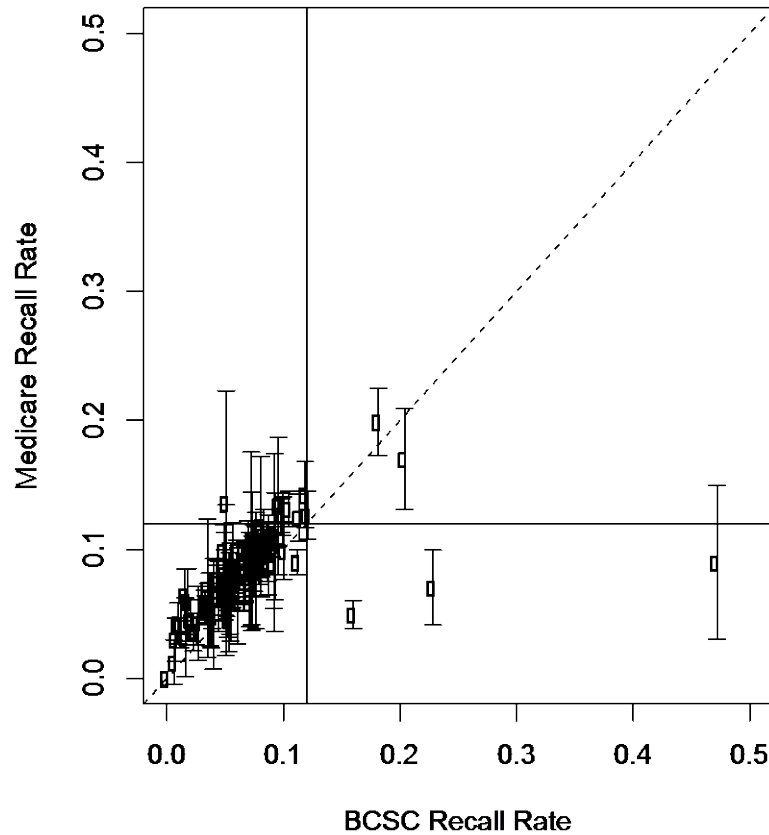
Claims-based algorithms

- ◆ Claims-based algorithms for recall and screen-detected cancer
 - Outcomes based on ICD-9 and HCPCS codes for breast imaging and breast cancer diagnosis around time of screening mammogram
- ◆ Algorithm performance:
 - Recall: Sensitivity = 82.6%, Specificity = 96.7%
 - CDR: Sensitivity = 94.0%, Specificity = 99.9%

Medicare-linked BCSC data

- ◆ Clinical data on mammography interpretation and cancer outcomes available from the Breast Cancer Surveillance Consortium
- ◆ Linked to Medicare claims
- ◆ Data on 134,330 screening mammograms from 2003 – 2005 performed at 106 mammography facilities
- ◆ Volume ranged from 52 to 5,925 mammograms per facility

Claims-based performance estimates



ML point estimates and 95% confidence intervals

Comparison of Medicare and BCSC estimates

- ◆ Imperfect specificity results in slight inflation of recall and CDR estimates
- ◆ Provider-level estimates based on claims agree well with gold-standard
- ◆ However, agreement between the two sources does not ensure correct classification of providers
- ◆ Evaluation of claims-based measures often includes only operating characteristics, but this does not address error in profiling due to sampling variability
- ◆ We repeated the simulation study incorporating error due to imperfect classification

Sensitivity and specificity under misclassification

	Point estimate		95% CI	
	Sens	Spec	Sens	Spec
Recall	98.9	92.9	86.8	99.1
CDR	54.8	95.7	2.7	99.9

Conclusions

- ◆ Reliability provides a good first indication of the likely success of profiling
- ◆ With or without misclassification of outcomes, performance is reasonable for recall because it is relatively common
- ◆ Classification of radiologists on CDR is challenging because outcome is rare
 - Profiling based on point estimates works reasonably well when there is no misclassification of events
 - Misclassification of events at the level of our Medicare claims-based algorithm resulted in low sensitivity
- ◆ Incorporating uncertainty through CIs results in decrease in sensitivity, increase in specificity
- ◆ The purpose of profiling should be considered when choosing an approach/determining acceptable levels of radiologist misclassification

Acknowledgments

**NCI-funded grant R21CA158510 and Breast Cancer
Surveillance Consortium (HHSN261201100031C)**

Rhondee Benjamin-Johnson

Rebecca Smith-Bindman

Weiwei Zhu

Tracy Onega

Joshua Fenton

References

Hubbard RA, Benjamin-Johnson R, Onega, T, Smith-Bindman R, Zhu W, Fenton JJ. 2015. Classification accuracy of Medicare claims-based methods for identifying providers failing to meet performance targets. *Statistics in Medicine*. 34(1):93-105.

Thank you

