

# Quantitative Measurement Error Bias and Correction

## Understanding and Correcting Misclassification Bias

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# Outline

- 1 The Problem: Misclassification Bias
- 2 The Solution: Quantitative Bias Correction
- 3 Practical Application
- 4 Conclusions

# What is Misclassification Bias?

## Definition

Misclassification bias occurs when measurement error leads to incorrect categorisation of study participants.

- Common in psychological research
- Affects self-report measures
- Impacts diagnostic categories
- Usually biases toward null if error is not systematic (but not always!)

## Types

- 1 **Non-differential:** Error independent of other variables
- 2 **Differential:** Error varies by group

**Key Point:** Even random error causes systematic bias!

## Example: True Association

### True Association (No Misclassification)

	Disease +	Disease -	Total
Exposed	80	20	100
Unexposed	20	80	100
Total	100	100	200

$$\text{Risk Ratio} = \frac{80/100}{20/100} = 4.0$$

## Example: With Misclassification

### With 20% Non-Differential Misclassification

	Disease +	Disease -	Total
Exposed	68	32	100
Unexposed	32	68	100
Total	100	100	200

$$\text{Observed Risk Ratio} = \frac{68/100}{32/100} = 2.125$$

*True effect attenuated from 4.0 to 2.125 (47% reduction)*

# Impact on Research

## Consequences

- Underestimated effect sizes
- Reduced statistical power
- Failed replications
- False null findings

## Common in Psychology

- Depression screening (PHQ-9)
- Self-reported behaviours
- Cognitive assessments
- Diagnostic interviews

## Critical Insight

Many "null" findings in psychology may actually reflect measurement error rather than absence of true effects!

## Mathematical Foundation

For non-differential misclassification of exposure:

$$P_{\text{true}} = \frac{P_{\text{observed}} - (1 - \text{Specificity})}{\text{Sensitivity} + \text{Specificity} - 1} \quad (1)$$

### Where:

- $P_{\text{true}}$  = True prevalence
- $P_{\text{observed}}$  = Observed prevalence

### Requirements:

- Known sensitivity
- Known specificity
- From validation study

# Correction Process Step-by-Step

## Given Parameters

Parameter	Value
Sensitivity	0.80 (80%)
Specificity	0.80 (80%)
Total with disease	100
Total without disease	100

## Observed Data

	Disease +	Disease -	Total
Exposed	68	32	100
Unexposed	32	68	100



# Applying the Correction

## Step 1: Calculate observed exposure prevalence

$$P_{\text{obs}}(\text{Exposed}|\text{Disease+}) = \frac{68}{100} = 0.68$$

$$P_{\text{obs}}(\text{Exposed}|\text{Disease-}) = \frac{32}{100} = 0.32$$

## Step 2: Apply correction formula

$$P_{\text{true}}(\text{Exposed}|\text{Disease+}) = \frac{0.68 - 0.20}{0.60} = 0.80$$

$$P_{\text{true}}(\text{Exposed}|\text{Disease-}) = \frac{0.32 - 0.20}{0.60} = 0.20$$

# Results of Correction

## Step 3: Calculate corrected cell counts

- True Exposed & Disease+ =  $0.80 \times 100 = 80$
- True Unexposed & Disease+ =  $0.20 \times 100 = 20$
- True Exposed & Disease- =  $0.20 \times 100 = 20$
- True Unexposed & Disease- =  $0.80 \times 100 = 80$

### Corrected Data

	Disease +	Disease -
Exposed	80	20
Unexposed	20	80

### Risk Ratio

Observed: 2.125

Corrected: 4.00

# When to Use Bias Correction

## Good Candidates

- Dichotomised continuous variables
- Self-report measures with known validity
- Diagnostic categories
- Screening tools

## Requirements

- Validation data available
- Non-differential assumption reasonable
- Measurement properties stable
- Same population

## Important Assumptions

- Misclassification is independent of other variables
- Sensitivity/specificity are accurately estimated
- The same error rates apply across all participants

## Simple R Code for Correction

```
# function to correct for misclassification
correct_prevalence <- function(p_obs, sens, spec) {
  p_true <- (p_obs - (1 - spec)) / (sens + spec - 1)
  return(p_true)
}

# apply to diseased group
p_true_diseased <- correct_prevalence(0.68, 0.8, 0.8)
# returns: 0.80
```

See also: `episensr` package for comprehensive bias analysis

# Examples in Psychology

Measure	Sensitivity	Specificity	Expected Bias
PHQ-9 Depression	0.88	0.88	-19%
Self-report Cannabis	0.70	0.95	-31%
ADHD Parent Rating	0.78	0.87	-29%
Memory Test (MCI)	0.85	0.80	-28%

## Interpretation

If true  $RR = 3.0$ , observed  $RR$ s would be:

- PHQ-9: 2.43 (19% attenuation)
- Cannabis use: 2.07 (31% attenuation)
- ADHD rating: 2.13 (29% attenuation)

# Key Takeaways

## ① Measurement error is ubiquitous

- All psychological measures have error
- Even "gold standards" aren't perfect

## ② Random error causes systematic bias

- Always toward the null
- Can be substantial (often 20-50%)

## ③ Correction is possible

- Requires validation data
- Simple mathematical approach
- Can recover true effects

## ④ Implications for psychology

- Many null findings may be false
- Replication "failures" may reflect measurement
- Power calculations should account for error

# References and Resources

## Primary References in This Lecture

Fox, M. P., MacLehose, R. F., & Lash, T. L. (2021). *Applying quantitative bias analysis to epidemiologic data*. New York: Springer.

**Full text:** <https://link.springer.com/content/pdf/10.1007/978-3-030-82673-4.pdf>

Lash, T. L., VanderWeele, T. J., Haneuse, S., & Rothman, K. J. (2021). *Modern Epidemiology* (4th ed.). Philadelphia: Wolters Kluwer.

## Additional Resources

- R package: `episensr` for sensitivity analysis
- R package: `quantbias` for quantitative bias analysis
- Online tools and tutorials at <https://sites.google.com/site/biasanalysis/>

# Thank You!

Questions?