Propensity Score Lweighting within complex survey

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SPPH 504/007

Propensity score Weighting (ATE + ATT)

IPW (inverse probability weighting)

How to conduct propensity score weighting?

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Step 1: Specify PS & fit model  
→ Exposure model (RA)
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Step 2: Match subjects by PS Convert PS to IPW

Step 3: Covariate balance in matched weighted sample

Step 4: Estimate treatment effect
→ Outcome model (MI)

For the purposes of illustration, we will first assume that our <u>data was collected</u> <u>via SRS</u>.

PW ERROR

Step 1: Fit PS model

Step 2: Convert PS = IPW(ATE)

IPW = 1/ps, if A = 1

IPW = 1/(1-ps), if A = 0

Step 3: Check balance

SMD in IPW-weighted data

Step 4: Outcome model with

Weight = IPW

IPW in complex survey

Step 1: Fit PS model

A~L (survey-weights as design variable / covariate)

complex surveys

Step 2: Convert PS = IPW(ATE)

IPW =
$$1/ps$$
, if $A = 1$

IPW =
$$1/(1-ps)$$
, if $A = 0$

Step 3: Check balance

Prop<mark>ensity score analysis with survey weighted data

<u>G Ridgeway, SA Kovalchik, BA Griffin</u>... - Journal of Causal ..., 2015 - degruyter.com</mark>

Generalizing observational study results: applying propensity score methods to

EH DuGoff, M Schuler, EA Stuart - Health services research, 2014 - Wiley Online Library Objective To provide a tutorial for using propensity score methods with complex survey data.

subclassification), and propensity score methods in combination with survey weighting ...

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Data Sources Simulated data and the 2008 Medical Expenditure Panel Survey. Study Design Using simulation, we compared the following methods for estimating the treatment effect: a naïve estimate (ignoring both survey weights and propensity scores), survey weighting, propensity score methods (nearest neighbor matching, weighting, and

"sampling weights in the propensity score estimation stage (as weights, not as a covariate)"

SMD (data weighted by w = IPW * survey-weights)

Step 4: Outcome model with

Weight = IPW * survey-weights

PW ESTA

Step 1: Fit PS model

A~L

Step 2: Convert PS = IPW(ATT)

IPW = 1, if A = 1

IPW = ps/(1-ps), if A = 0

Step 3: Check balance

SMD in IPW-weighted data

Step 4: Outcome model with

Weight = IPW

IPW in complex survey (ATT)

Step 1: Fit PS model

A~L (survey-weights as design variable / covariate)

Step 2: Convert PS = IPW(ATT)

IPW = 1, if A = 1

IPW = ps/(1-ps), if A = 0

Step 3: Check balance

SMD (data weighted by w = IPW * survey-weights)

Step 4: Outcome model with

Weight = IPW * survey-weights

Reasonable approach (my summary)

- PS model: (population-level)
 - use <u>design variables</u> (cluster + strata + weight) to estimate ps (not as covariate)
 - Combined weight = ipw * survey weight
- Outcome model: (population-level)
 - use design features (strata+psu as well as combined weight) to get <u>population level</u> estimates

Estimates and conclusion=

Adult patients with RA are at increased risk for MI in US (based on 2007-08 data)?

Risk of cardiovascular mortality in patients with rheumatoid arthritis: a metaanalysis of observational studies

JA Aviña-Zubieta, HK Choi... - Arthritis Care & ..., 2008 - Wiley Online Library

Objective To determine the magnitude of risk of cardiovascular mortality in patients with rheumatoid arthritis (RA) compared with the general population through a meta-analysis of observational studies. Methods We searched Medline, EMBase, and Lilacs databases from

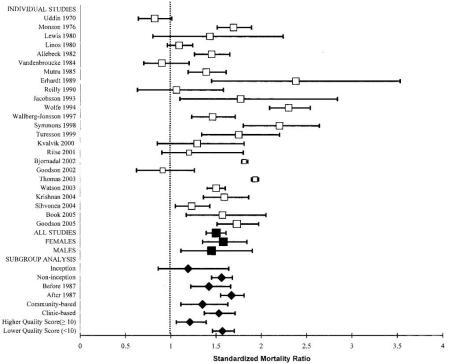


Figure 1. Meta-analysis of 24 studies on cardiovascular disease mortality in patients with rheumatoid arthritis.

50% increased risk of CVD death in patients with RA 9

Estimates from NHANES (2007-08) and conclusion =

OR: population-based estimates, sample-based not shown

	Adjusted Regression	Matching (Zanutto)	Matching (DuGoff)	Matching (design in both stages)	Weighting (Ridgeway)	Weighting (DuGoff)
PATT		1.87 (0.86 4.07)	1.26 (0.55, 2.88)	1.66 (0.65, 4.28)	1.38 (0.71, 2.71)	1.37 (0.71, 2.67)
PATE	1.66* (0.71, 3.89)				1.51 (0.68, 3.35)	1.43 (0.62, 3.28)

^{*} Also conditional estimates if further adjustment made;

SE / CI width is a function of n.

NHANES VS. CCHS 💝

- In the public release data, NHANES provides
 - o masked variance <u>pseudo-PSUs</u>, and
 - masked variance <u>pseudo-stratum</u>

to account for the complex survey design.

 CCHS public use microdata file (PUMF) does not contain PSU / Stratum information. Any SE calculation <u>assumes</u> <u>SRS</u> even if weights are used. <u>RDC</u> provides access to master data with these necessary information.

Short Reference and Textbook List

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