

# Replication of Empirical Results in 'Finite-Sample Optimal Estimation and Inference on Average Treatment Effects Under Unconfoundedness'

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## 1 Empirical application: NSW

We use a solution path with 3000 steps. At the last step the value of  $\delta$  is given by 668.48739.

Table 2:

|                 | Opt: RMSE | Opt: FLCI | Matching: RMSE | Matching: FLCI |
|-----------------|-----------|-----------|----------------|----------------|
| $M$             |           |           | 1.00           | 18.00          |
| $\delta$        | 1.82      | 3.30      |                |                |
| Estimate        | 0.96      | 0.94      | 1.39           | 1.26           |
| Max. Bias       | 1.64      | 1.78      | 1.48           | 2.21           |
| SE              | 1.01      | 0.94      | 1.09           | 0.89           |
| cv              | 3.26      | 3.55      | 3.01           | 4.13           |
| CI lo           | -2.33     | -2.38     | -1.88          | -2.41          |
| CI hi           | 4.26      | 4.27      | 4.66           | 4.93           |
| SE (PATE)       | 1.06      | 1.00      | 1.14           | 0.94           |
| CI lo (PATE)    | -2.75     | -2.80     | -2.32          | -2.78          |
| CI hi (PATE)    | 4.67      | 4.69      | 5.11           | 5.30           |
| Lindeberg wgt   | 0.07      | 0.06      | 0.19           | 0.06           |
| Rel. efficiency | 101.15    |           | 90.41          | 91.36          |

Efficiency of FLCI in baseline specification (Section 2.5)

```
ATEHonest::ATTEffBounds(op, C = 1, sigma2 = mean(sigma2))
## $onesided
## [1] 0.977761
##
## $twosided
## [1] 0.970476
```

Optimal estimator and CIs as a function of  $C$ . Solid lines mark the CATT CI, while dotted lines mark the PATT CI.

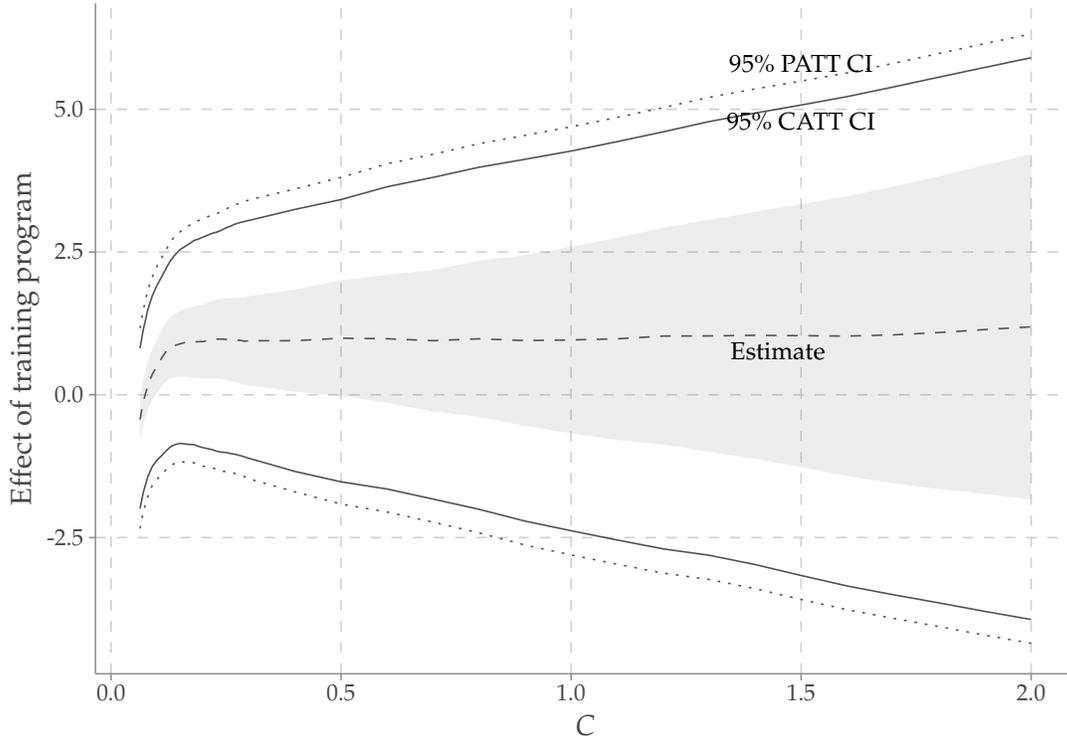


Figure 1: Reproduction of Figure 1 in the paper. Robust SE.

Range of length penalty of PATT relative to CATT in the four columns of Table 2: 1.102, 1.135

Range of point estimate for reasonable  $C$ :

```
range(oe$rmse$att[oe$rmse$C > 0.2 & oe$rmse$C < 2])
## [1] 0.939249 1.142967
```

### 1.1 Matching efficiency

Efficiency at  $C = 1$ :

| C | type | eff     |
|---|------|---------|
| 1 | RMSE | 90.4132 |
| 1 | FLCI | 85.9666 |

Minimum  $C$  such that efficiency is at least 95%:

| C   | type | eff     |
|-----|------|---------|
| 1.5 | RMSE | 95.6341 |
| 2.8 | FLCI | 95.0211 |

## 1.2 Lindeberg weights

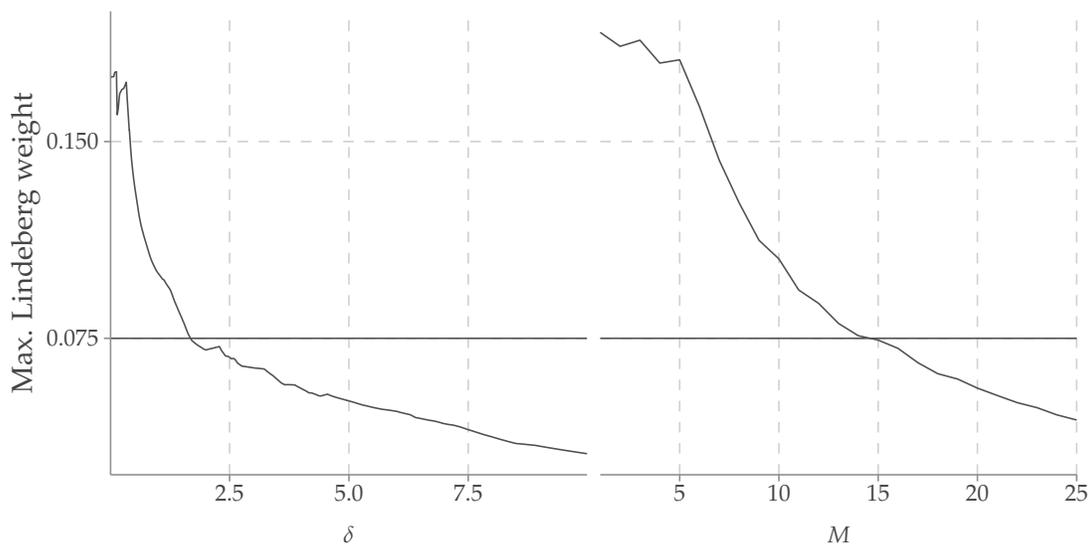


Figure 2: Lindeberg weights for optimal and matching estimator.

For optimal estimator, we need  $\delta$  to be at least 1.686452. For matching estimator, we need at least 15 matches to be below the 0.075 cutoff.

## 1.3 Experimental data

```
Xe <- as.matrix(ATEHonest::NSWexper[, 2:10])
de <- ATEHonest::NSWexper$treated
ye <- ATEHonest::NSWexper$re78
s2 <- ATEHonest::nnvar(ATEHonest::distMat(Xe, chol(solve(cov(Xe))),
  method = "euclidean"), de, ye, J = 2)
```

Difference in means estimate:

```
RCTvariance(ye, de, s2, ATEHonest::distMat(Xe, Amain, method = "manhattan",
  de))
##      att  maxbias Cond rob Cond hom CATT rob CATT hom PATE HC2 PATE EHW
## 1.794342 1.737712 0.611274 0.578961 0.641517 0.610806 0.670997 0.669315
## PATE hom
## 0.632853
```

## 1.4 Alternative choice of distance (Supplement)

Now consider the norm in Abadie and Imbens (2011), with  $p = 2$ , and  $C$  described in the text. Optimal FLCI:

```
oe2 <- ATEHonest::ATTOptEstimate(y = y, d = d, D0 = D2, C = 1/sqrt(2 *
  0.07^2), sigma2 = sigma2, opt.criterion = "FLCI")
## Increasing length of solution path to 100
```

```

oe2
##
##
## |      | Estimate| Max. bias|      SE|CI          | delta|
## |-----|-----:|-----:|-----:|-----:|-----:|
## |CATT | 1.72259| 6.25009| 0.842248|(-5.91288, 9.35805) | 3.28971|
## |PATT | 1.72259|          | 0.913810|(-6.31854, 9.76371) |          |
oe2$e["rhl"]
##      rhl
## 7.63546

```

Now consider a matching estimator that matches on this norm:

```

mp2 <- ATEHonest::ATTMatchPath(y, d, D2, M = 1)
## Now calculate worst-case bias under D0
mp2$ep$maxbias <- ATEHonest::ATTbias(mp2$K[, d == 0], D0)
mp2$D0 <- D0
me2 <- ATEHonest::ATTMatchEstimate(mp2, sigma2 = sigma2, C = 1)
me2$e[c("hl", "rhl")]
##      hl      rhl
## 5.38073 3.37502

```

```

meff2 <- MatchEfficiency(oe, mp2, mean(sigma2), mineff = 95)
knitr::kable(meff2$effC1, digits = 4, row.names = FALSE)

```

| C | type | eff     |
|---|------|---------|
| 1 | RMSE | 77.8138 |
| 1 | FLCI | 74.9252 |

```

## Efficiency loss relative to using correct norm
meff$effC1$eff - meff2$effC1$eff
## [1] 12.5994 11.0414
## Maximum efficiency
max(meff2$d[meff2$d$type == "RMSE", "eff"])
## [1] 80.1513

```

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