

w241: Experiments and Causality

Unit 4

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Blocking

- Hard to know if it was due to chance when there are large differences between treatment and control.
- Need to reduce the size of the differences that can arise by chance.
- Increase statistical power given an experiment with same sample and effect size.
- If some variables are related to the outcome, restrict ourselves to randomizations that keep treatment and control similar.

Make Data

```
d ← make_data(effect_size=0)
head(d)
```

```
##      Control Treatment group
## 1:         1          1   Man
## 2:         2          2   Man
## 3:         3          3   Man
## 4:         4          4   Man
## 5:         5          5   Man
## 6:         6          6   Man
```

Randomization

```
d[, assignment := randomize(size=40)][ ,  
  table('Sex' = group, 'Assignment' = assignment)]
```

```
##           Assignment  
## Sex      Control Treatment  
##   Man         12         8  
##   Woman        8        12
```

```
d[, assignment := randomize(size=40)][ ,  
  table('Sex' = group, 'Assignment' = assignment)]
```

```
##           Assignment  
## Sex      Control Treatment  
##   Man         10         10  
##   Woman        10         10
```

Block Randomization

```
block_randomize ← function(size) {  
  ## this function will be executed /within/ the data.table that  
  ## holds the data. It could be run outside, but the assignment  
  ## in place that data.table provides make it clean inside.  
  conditions ← c('Control', 'Treatment')  
  
  if(size %% 2 == 0) {  
    ## if there are an even number of units in each block this is easy  
    urn ← rep(conditions, times = size/2)  
  } else if(size %% 2 == 1) {  
    ## if there are an odd number, then produce conditions to the  
    ## nearest even number that is less than the number of units  
    ## then add one more assignment condition, sampled at random  
    urn ← c(rep(conditions, times = (size/2) - 0.5), sample(conditions, size = 1))  
  }  
  
  ## now, shuffle it up return the shuffled sequence  
  assignment ← sample(urn)  
  return(assignment)  
}
```

Randomization

```
d[ , block_assignment := block_randomize(size=.N), by = group][ ,  
  table('Sex' = group, 'Assignment' = block_assignment)]
```

```
##           Assignment  
## Sex      Control Treatment  
##   Man         10         10  
##   Woman        10         10
```

```
d[ , block_assignment := block_randomize(size=.N), by = group][ ,  
  table('Sex' = group, 'Assignment' = block_assignment)]
```

```
##           Assignment  
## Sex      Control Treatment  
##   Man         10         10  
##   Woman        10         10
```

Conduct Experiment

```
conduct_experiment ← function(potential_control, potential_treatment, assignment) {  
  outcomes ← potential_treatment * I(assignment == "Treatment") +  
    potential_control * I(assignment == "Control")  
  
  return(outcomes)  
}
```

```
d[, Y := conduct_experiment(Control, Treatment, block_assignment)]
```

```
head(d)
```

```
##      Control Treatment group assignment block_assignment Y  
## 1:         1         1   Man  Treatment      Treatment 1  
## 2:         2         2   Man  Treatment      Control 2  
## 3:         3         3   Man   Control      Treatment 3  
## 4:         4         4   Man  Treatment      Control 4  
## 5:         5         5   Man   Control      Treatment 5  
## 6:         6         6   Man   Control      Control 6
```

Estimate ATE

```
estimate_ate <- function(y_values, treatment, verbose=FALSE) {  
  
  treatment_group_mean <- mean(y_values[treatment == 'Treatment'])  
  control_group_mean   <- mean(y_values[treatment == 'Control'])  
  
  ate <- treatment_group_mean - control_group_mean  
  
  if(verbose) {  
    return(  
      list(  
        "tg_mean" = treatment_group_mean,  
        "cg_mean" = control_group_mean,  
        "ate" = ate))  
    } else {  
      return("ate" = ate)  
    }  
  }  
  
  ate <- d[ , estimate_ate(y_values = Y, treatment = block_assignment, verbose=TRUE)]  
  ate  
  
##      tg_mean cg_mean ate  
## 1:    35.95    35.05 0.9
```


Simulate A Normal Study

```
simulate_normal_study <- function(effect_size) {  
  ## create world  
  d <- make_data(effect_size=effect_size)  
  
  ## randomly assign and count the number of women in treatment  
  d[, assignment := randomize()]  
  
  women_in_treatment <- d[group = 'Woman' & assignment = 'Treatment', .N]  
  
  ## measure outcomes  
  d[, Y := conduct_experiment(Control, Treatment, assignment)]  
  
  ## estimate ate  
  ate <- d[, estimate_ate(y_values = Y, treatment = assignment)]  
  
  ## return objects  
  ## - `ate` from the `estimate_ate` function.  
  ## - `women_in_treatment` as a count  
  return(list('ate' = ate, 'women_in_treatment' = women_in_treatment))  
}
```

Run One Normal Study

```
normal_study ← simulate_normal_study(effect_size = 0)  
normal_study
```

```
## $ate  
## [1] -5.3  
##  
## $women_in_treatment  
## [1] 9
```

Simulate Many Normal Studies

```
many_normal_studies <- replicate(  
  n = 1000,  
  expr = simulate_normal_study(effect_size = 0))
```

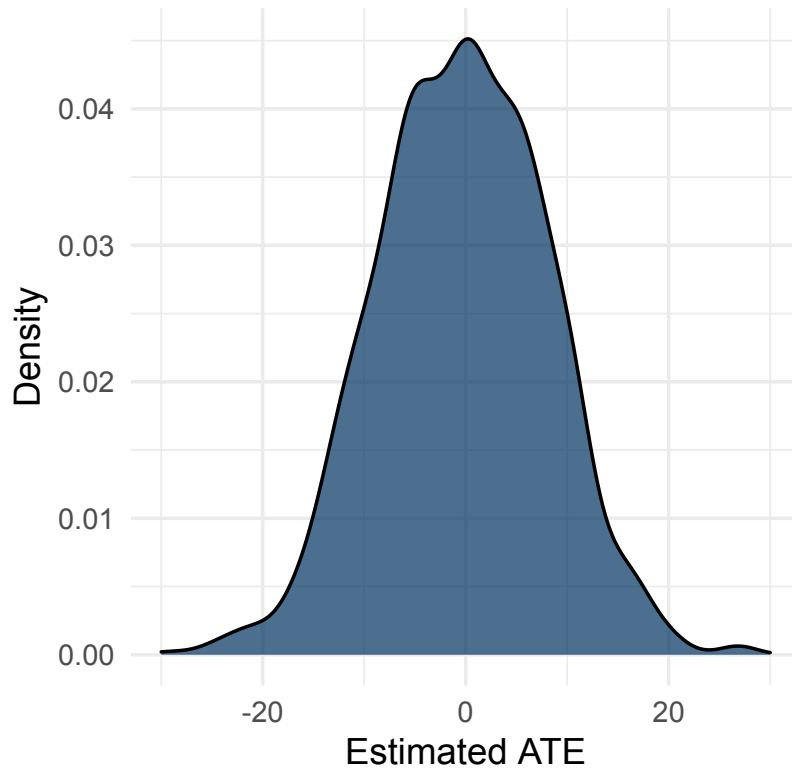
```
many_normal_studies <- t(many_normal_studies)  
head(many_normal_studies)
```

```
##      ate  women_in_treatment  
## [1,] 1.8    10  
## [2,] 0.6    10  
## [3,] 4.3    11  
## [4,] -6     9  
## [5,] -9.8   8  
## [6,] 2.7    10
```

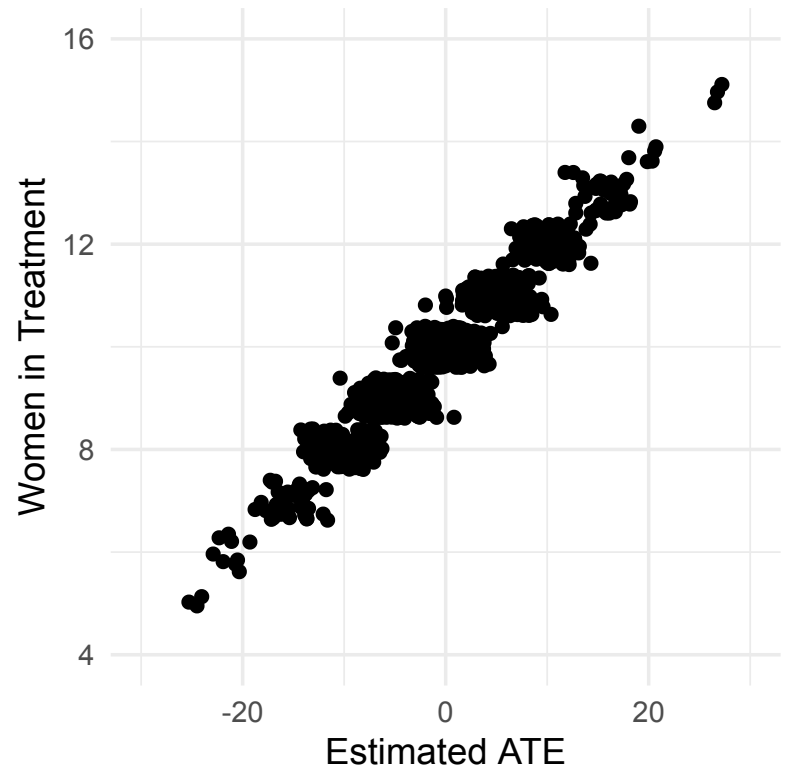
Plot Normal ATE

```
## Warning: Removed 1 rows containing missing values (geom_point).
```

Zero Effect, Simple Randomization
Some extreme effects



Treatment Effect vs. Women in Treat
Linear, increasing relationship



Simulate a Block Randomized Study

```
simulate_blocked_study ← function(effect_size) {  
  ## create world  
  d ← make_data(effect_size=effect_size)  
  
  ## randomly assign and count the number of women in treatment  
  d[ , assignment := block_randomize(20), by = group]  
  
  women_in_treatment ← d[group = 'Woman' & assignment = 'Treatment', .N]  
  
  ## measure outcomes  
  d[ , Y := conduct_experiment(Control, Treatment, assignment)]  
  
  ## estimate ate  
  ate ← d[ , estimate_ate(y_values = Y, treatment = assignment)]  
  
  ## return objects  
  ## - `ate` from the `estimate_ate` function.  
  ## - `women_in_treatment` as a count  
  return(list('ate' = ate, 'women_in_treatment' = women_in_treatment))  
}
```

Simulate a Block Randomized Study

```
blocked_study ← simulate_blocked_study(effect_size = 0)
blocked_study

## $ate
## [1] -1.4
##
## $women_in_treatment
## [1] 10
```

Simulate Many Block Randomized

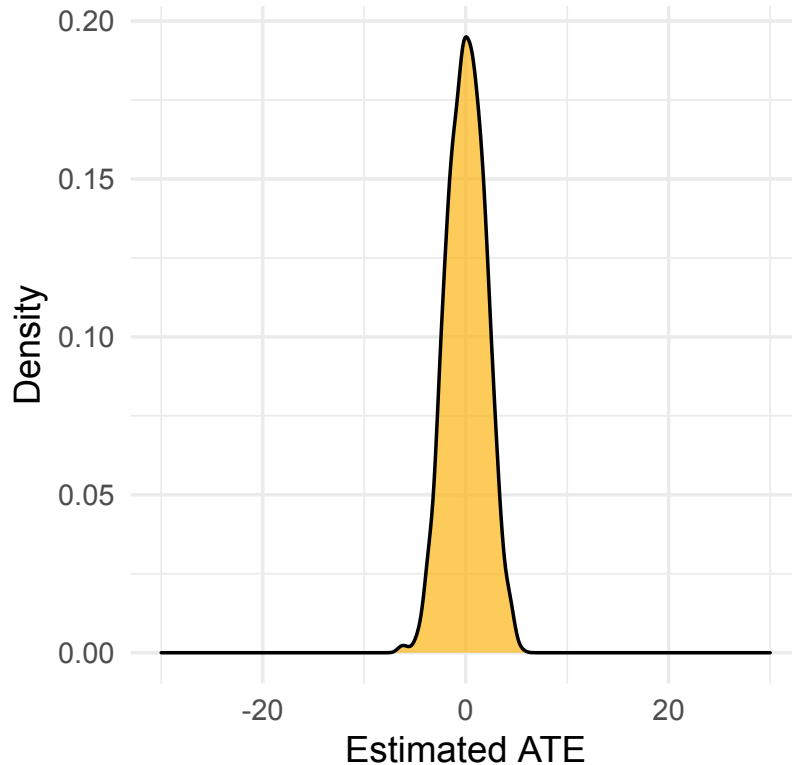
```
many_blocked_studies <- replicate(  
  n = 1000,  
  expr = simulate_blocked_study(effect_size = 0)  
)
```

```
many_blocked_studies <- t(many_blocked_studies)  
head(many_blocked_studies)
```

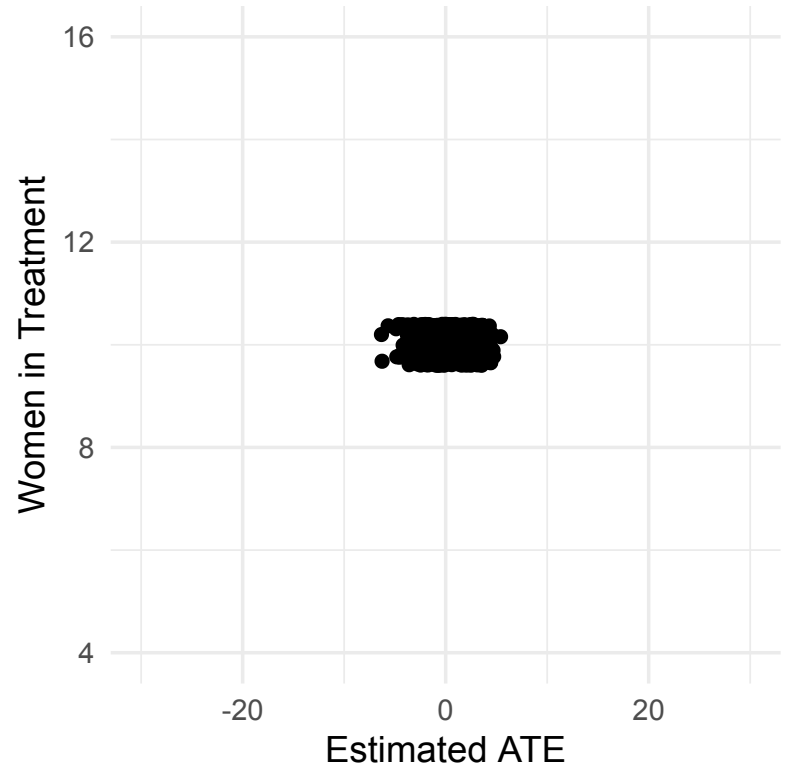
```
##      ate  women_in_treatment  
## [1,] 0.6   10  
## [2,] -1.2  10  
## [3,] -2.3  10  
## [4,] 1.4   10  
## [5,] 1.1   10  
## [6,] 1.7   10
```

Plot Blocked ATE

Zero Effect, Blocked Randomization
Results tightly centered at true treatment effect

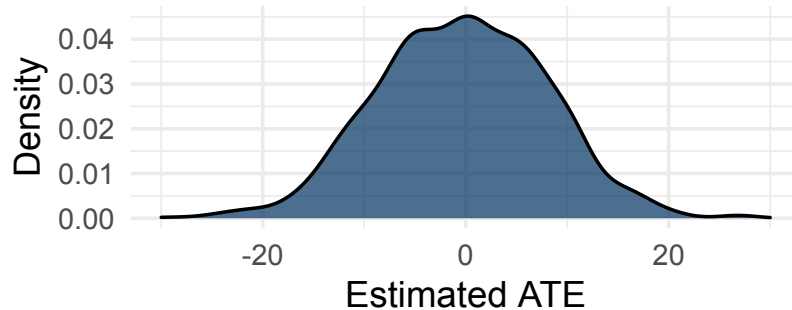


Treatment Effect vs. Women in Treat

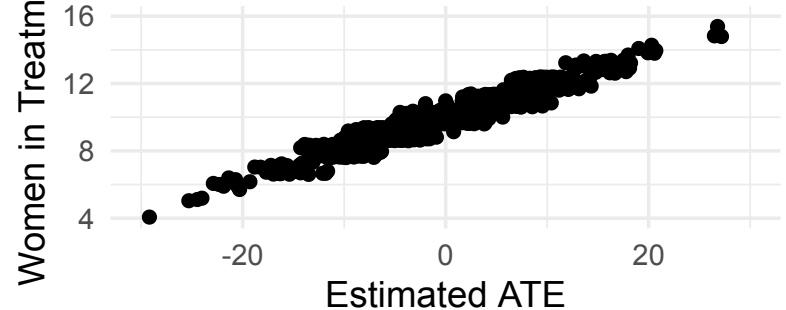


Plot Unblocked and Blocked

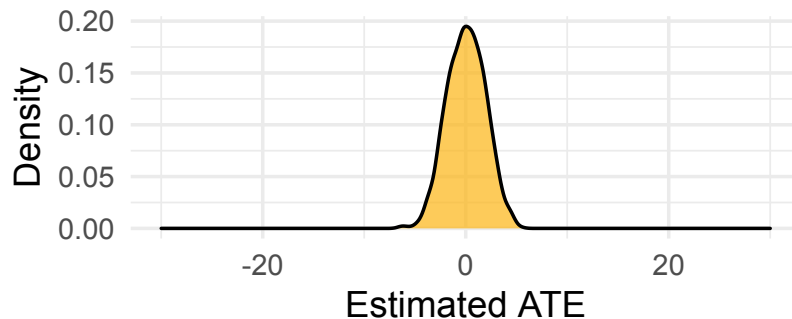
Zero Effect, Simple Randomization
Some extreme effects



Treatment Effect vs. Women in Treat
Linear, increasing relationship



Zero Effect, Blocked Randomization
Results tightly centered at true treatment effect



Treatment Effect vs. Women in Treat

