

Simulation

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Data-Generating Process for Simulation

Bootstrap samples by cluster

```
DGP <- function(clusters, Y, cluster=F){  
  G = length(clusters)  
  if(cluster){  
    return(Y[unlist(sapply(1:G, function(g){  
      lag(cumsum(clusters), default=0)[g]+  
        floor(clusters[g]*runif(clusters[g]))+1  
    })])])  
  } else {  
    return(Y[floor(length(Y)*runif(length(Y)))+1])  
  }  
}
```

```

}

g.map <- women.cleaned %>% group_by(countryid) %>%

  summarize(

    total = n()

  ) %>% ungroup() %>% mutate(

    g.index = cumsum(total)

  )

```

Two-Stage Inference with Homoskedasticity Assumption

```

twosls <- function(X, Y, Z){

  n <- nrow(X)

  if(ncol(X) == ncol(Z)){

    #Just Identified

    beta.hat <- solve(t(Z)%*%X)%*%t(Z)%*%Y

  } else {

    #Assume homoskedasticity

    W <- solve(t(Z)%*%Z)

    beta.hat <- solve(t(X)%*%Z)%*%W)%*%t(Z)%*%X)%*%

```

```

      t(X)%*%Z)%*%W)%*%t(Z)%*%Y
    }

    sigma2 = (1/n)*sum((Y-X)%*%beta.hat)^2)

    #Variance-Covariance Matrix

    V <- sigma2*solve((t(X)%*%Z)%*%(solve(t(Z)%*%Z))%*%
                      t(t(X)%*%Z))

    #95% C.I.

    beta.hat[1,]+c(-1,1)*qnorm(0.975)*sqrt(V[1,1])
  }

  #These two are the same via the Frisch-Waugh Theorem

  #twosls(X, Z, Y)

  twosls(X.tilde, Y.tilde, Z.tilde)

```

```
[1] -133.36956 -82.31544
```

Cluster Robust Variance Function

```

cluster.robust.iv <- function(X, Y, Z, clusters){ #Assumes data is sorted

  n <- nrow(X)

```

```

k <- ncol(X)

G <- length(clusters)

if(ncol(X) == ncol(Z)){

  #Just Identified

  beta.hat <- solve(t(Z)%*%X)%*%t(Z)%*%Y

}

u.hat <- X%*%beta.hat - Y


xtz <- t(X)%*%Z

Sigma <- matrix(0, k, k)

for(g in 1:G){

  cluster = (lag(cumsum(clusters), default = 0)[g]+1):cumsum(clusters)[g]

  M.z <- solve(diag(1, clusters[g]))-

    Z[cluster,]%*%solve(t(Z)%*%Z)%*%t(Z[cluster,]))

  zeta.g <- t(Z[cluster,])%*%M.z%*%u.hat[cluster]

  Sigma <- Sigma+zeta.g%*%t(zeta.g)

}

V <- (G-1)/G*solve(n^-1*xtz%*%solve(n^-1*Sigma)%*%t(n^-1*xtz))

#95% C.I.

```

```

    beta.hat[1,]+c(-1,1)*qnorm(0.975)*sqrt(n^-1*V[1,1])
}

```

3. Cluster Robust Variance #3 (Jackknife)

$$\frac{G-1}{G}(\hat{X}'\hat{X})^{-1}\left(\sum_{g=1}^G \acute{s}_g \acute{s}_g'\right)(\hat{X}'\hat{X})^{-1}$$

$$\acute{s}_g = \hat{X}_g' M_{gg}^{-1} \hat{u}_g$$

Proposed CRVE for Two-Stage Least Squares

$$\sqrt{n}(\hat{b} - \beta) \xrightarrow{d} N(0, (\mathbb{E}[X_i Z_i'] \Sigma^{-1} \mathbb{E}[X_i Z_i'])^{-1})$$

$$\Rightarrow \hat{V} = \left(\frac{1}{n} X' Z \hat{\Sigma}^{-1} \left(\frac{1}{n} X' Z\right)'\right)^{-1}; \quad \Sigma^{-1} = \mathbb{E}[Z_i Z_i' u_i^2] \Rightarrow \hat{\Sigma}^{-1} = \left(\frac{1}{n} \sum_{g=1}^G Z_g' Z_g \hat{u}_g^2\right)^{-1}$$

$$\frac{G-1}{G}(X'Z)^{-1}\left(\sum_{g=1}^G \hat{\zeta}_g \hat{\zeta}_g'\right)(X'Z)^{-1}$$

$$\hat{\zeta}_g = Z_g' \tilde{M}_{gg}^{-1} \hat{u}_g; \quad \tilde{M}_{gg} = I_{N_g} - Z_g(Z_g' Z_g)^{-1} Z_g'$$

```

cluster.robust.iv(X.tilde, Y.tilde, Z.tilde, g.map$total)

```

```
[1] -166.85855 -48.82644
```

Simulated CRVE Intervals

```
simulate.CRVE <- function(n, bootstrap.cluster=F){  
  n.sims <- n  
  sims.ci <- matrix(0, nrow=n.sims, ncol=2*2)  
  for(i in 1:n.sims){  
    Y.new = DGP(g.map$total, Y.tilde, cluster=bootstrap.cluster) %>% as.matrix()  
    for(j in 1:2){  
      if(j == 2){  
        sims.ci[i,(j+(j-1)):(j+(j-1)+1)] = twosls(X.tilde, Y.new, Z.tilde)  
      } else {  
        sims.ci[i,(j+(j-1)):(j+(j-1)+1)] =  
          cluster.robust.iv(X.tilde, Y.new, Z.tilde, g.map$total)  
      }  
    }  
  }  
  colnames(sims.ci) <- c("JackKnife.025", "JackKnife.975",  
                        "2SLS.025", "2SLS.975")  
  return(sims.ci)  
}
```

```

coverage <- function(sims){

  coverages = c()

  k = 1

  for(i in 1:(ncol(sims)/2)){

    coverages[i] = sum(0 < sims[,k] & 0 < sims[,k+1] |

      (0 > sims[,k] & 0 > sims[,k+1]))/nrow(sims)

    k = k+2

  }

  names(coverages) = unlist(lapply(str_split(colnames(sims), "[.]"), first)) %>%

    unique()

  return(coverages)

}

simulate.CRVE(10) %>% round(2) %>% knitr::kable()

```

JackKnife.025	JackKnife.975	2SLS.025	2SLS.975
-52.84	34.26	-28.84	10.26
-27.50	29.14	-15.54	17.18
-24.78	37.16	-13.10	25.48
-27.70	52.43	-11.39	36.12

JackKnife.025	JackKnife.975	2SLS.025	2SLS.975
-32.18	43.12	-15.83	26.77
-27.09	15.15	-28.52	16.58
-17.45	9.31	-20.79	12.64
-22.20	13.02	-24.25	15.06
-13.52	20.22	-14.28	20.98
-13.06	23.20	-16.02	26.15

Simulated CRVE Intervals with Clustered Bootstrap Sampling

```
sims.cluster = simulate.CRVE(1000, bootstrap.cluster = F)
```

```
sims.cluster %>%
```

```
round(2) %>% head(10) %>% knitr::kable()
```

JackKnife.025	JackKnife.975	2SLS.025	2SLS.975
-13.22	25.85	-12.49	25.13
-42.16	52.77	-14.56	25.18
-36.76	16.55	-30.92	10.71
-34.08	2.39	-33.56	1.87
-21.61	3.07	-27.12	8.58
-28.63	36.88	-17.26	25.51
-18.35	22.95	-16.44	21.04
-15.58	28.74	-12.59	25.74
-77.46	28.70	-45.39	-3.38
-12.51	27.83	-11.68	27.00

```
coverage(sims.cluster)
```

JackKnife 2SLS

0.016 0.047