

## 1. Simple regression simulation

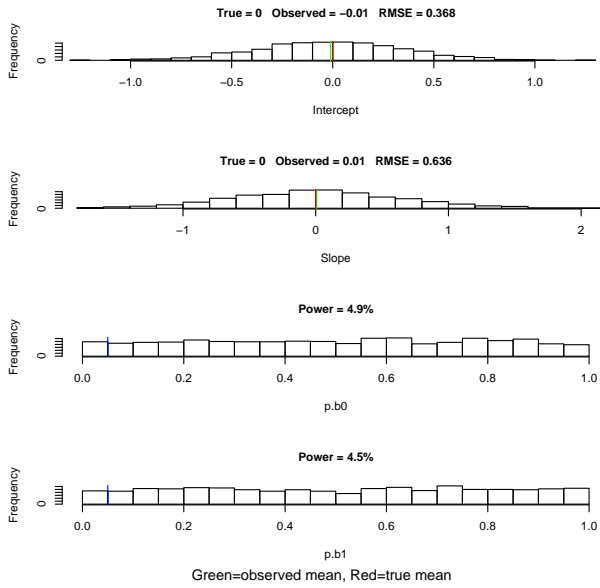
`eiv()` is a function that generates random data for a simple regression with “n” subjects, optionally adding measurement error to the x data (violating the fixed-x assumption and resulting in a so-called “errors in variables” model). It then calculates the standard regression results for the simulated data. The whole process is repeated “nsim” times, and the results are returned in a “simulation” object.

The x data is generated using a uniform [0,1] distribution. You may set parameters for n, nsim, b (true intercept and slope), sdy (true error s.d.), and sdx (true x measurement error s.d.).

Note: RMSE (root mean square error) is an estimate of  $\sigma$ .

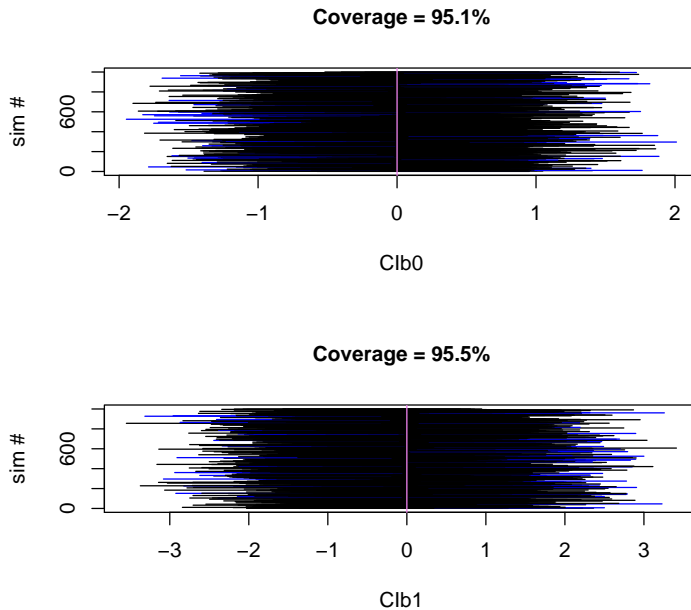
```
> s1 = eiv(nsim=1000, n=30, b=c(0,0), sdy=1, sdx=0)
> print(s1)
Simulation: s1
1000 simulations with estimation of Intercept and Slope.
True values: 0, 0.
Tests: p.b0 and p.b1.
CIs: CIb0 and CIb1.
Options: nsim=1000, n=30, sdx=0, sdy=1.
> summary(s1)
$nsim
[1] 1000
$alpha
[1] 0.05
$estimates
      Statistic
Parameter  True Value      Mean      Bias      Variance      RMSE
Intercept  0.000000000 -0.010462687 -0.010462687  0.135681577  0.368498364
Slope      0.000000000  0.007420337  0.007420337  0.404349842  0.635928379
$power
  p.b0  p.b1
0.049  0.045
$coverage
      MeanWidth  MeanCenter  True Coverage
CIb0  1.517241 -0.010462687    0    0.951
CIb1  2.639221  0.007420337    0    0.955
> plot(s1)
```

1000 Simulations:  $n=30$ ,  $sd_x=0$ ,  $sd_y=1$



```
> plot(s1, doParam=FALSE, doP=FALSE, doCI=TRUE)
```

1000 Simulations:  $n=30$ ,  $sd_x=0$ ,  $sd_y=1$



**Question 1:** What model was simulated, and how many times? What can you learn from the output, and how does it match your expectations?

If you remember something about the binomial distribution, see if you can explain how the following calculation relates to the uncertainty in the simulation results.

```
> p=0.05
> simVar = p*(1-p)/1000
> plusMinus = 2 *sqrt(simVar)
> plusMinus # [1] 0.01378405
> 0.05 + c(-1,1)*plusMinus
[1] 0.03621595 0.06378405
> 0.95 + c(-1,1)*plusMinus
[1] 0.936216 0.963784
```

## 2. Simulation when standard null hypotheses ( $\beta_i = 0$ ) are false

```
> s2 = eiv(nsim=1000, n=30, b=c(0.25,0.5), sdy=1, sdx=0)
> summary(s2) # (abbreviated results:)
$estimates
      Statistic
Parameter True Value      Mean      Bias      Variance      RMSE
  Intercept 0.25000000 0.26333971 0.01333971 0.14033487 0.37485039
   Slope    0.50000000 0.46705637 -0.03294363 0.42228688 0.65067055
$power
  p.b0 p.b1
0.098 0.113
$coverage
      MeanWidth MeanCenter True Coverage
CIb0 1.520409 0.2633397 0.25 0.948
CIb1 2.642626 0.4670564 0.50 0.953
```

**Question 2:** What model was simulated? Are  $H_0$ 's true, false, or undetermined? What can you learn from the output, and how does it match your expectations?

## 3. Errors in variables with $b=c(0,0)$

```
> s3 = eiv(nsim=1000, n=30, b=c(0,0), sdy=1, sdx=1)
> summary(s3)
$estimates
      Statistic
Parameter True Value      Mean      Bias      Variance      RMSE
  Intercept 0.00000000 0.003937849 0.003937849 0.043071478 0.207574047
   Slope    0.00000000 0.005046648 0.005046648 0.036418210 0.190902274
$power
```

```

p.b0 p.b1
0.053 0.054
$coverage
      MeanWidth MeanCenter True Coverage
CIb0 0.8400203 0.003937849    0    0.947
CIb1 0.7428122 0.005046648    0    0.946

```

**Question 3:** What model was simulated? What can you learn from the output, and how does it match your expectations?

4. **Errors in variables with  $b=c(0.25,0.5)$**

```

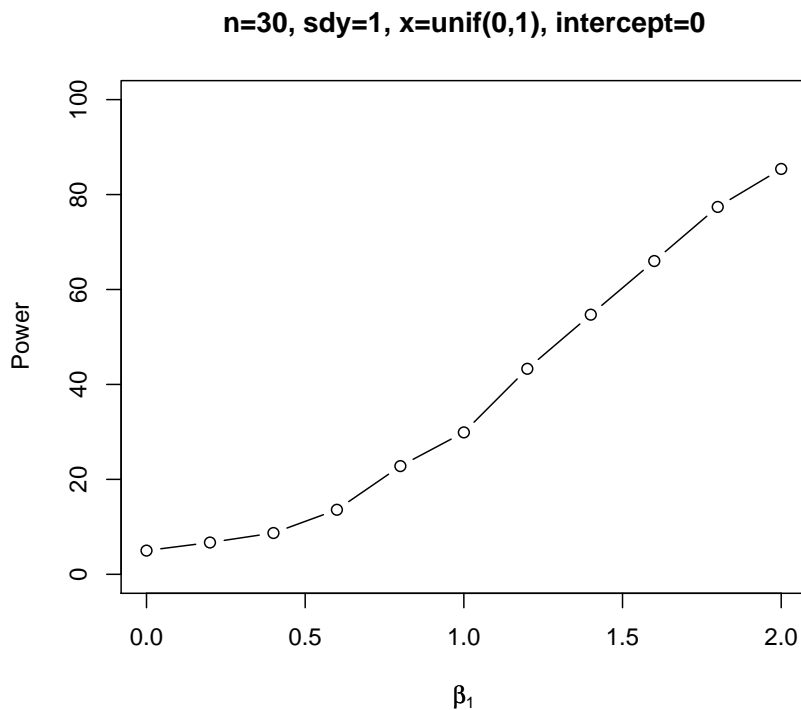
> s4 = eiv(nsim=1000, n=30, b=c(0.25,0.5), sdy=1, sdx=1)
> summary(s4)
$estimates
      Statistic
Parameter  True Value      Mean      Bias      Variance      RMSE
Intercept  0.25000000  0.46990074  0.21990074  0.04658917  0.30813228
Slope      0.50000000  0.04452477 -0.45547523  0.03760090  0.49503392
$power
p.b0 p.b1
0.572 0.052
$coverage
      MeanWidth MeanCenter True Coverage
CIb0 0.8502235 0.46990074 0.25    0.810
CIb1 0.7596810 0.04452477 0.50    0.342

```

**Question 4:** What model was simulated? What can you learn from the output, and how does it match your expectations? Make educated guesses about what would happen if “sdx” were increased, and if  $\beta_1$  were set to +1.0 or -0.5.

## 5. Breakout #2: Power curve

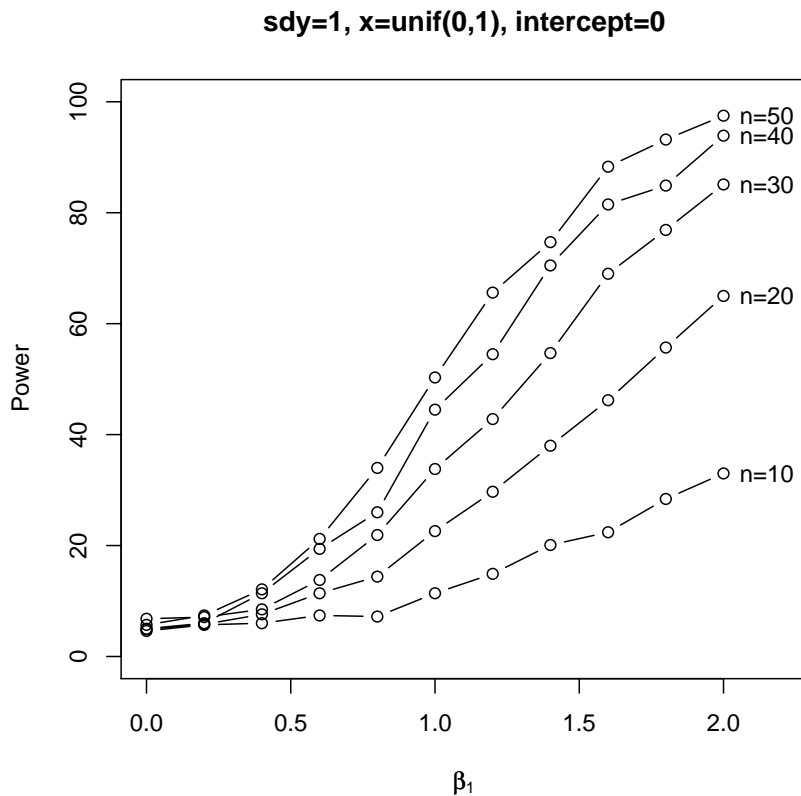
```
> summary(eiv(nsim=1000, n=30, b=c(0,0), sdy=1, sdx=0))$power[2]
p.b1
0.058
> b1s = seq(0,2,0.2)
> nb1 = length(b1s)
> pwr = rep(NA, nb1)
> for (i in (1:nb1))
+ pwr[i]=summary(eiv(nsim=1000, n=30, b=c(0,b1s[i]), sdy=1, sdx=0))$power[2]*100
> plot(b1s, pwr, type="b", xlab=expression(beta[1]), ylab="Power", ylim=c(0,100),
+ main=paste("n=30, sdy=1, x=unif(0,1), intercept=0"))
```



**Question 5:** How does the code work? What is the explanation for the shape of the curve? What are the asymptote values?

## 6. Set of power curves

```
# Set of power curves using an alternate approach:
> fpwr = function(beta1, n, nsim=1000)
+ return(summary(eiv(nsim=nsim, n=n, b=c(0,beta1), sdy=1, sdx=0))$power[2]*100)
> plot(c(0, 2.2), c(0,100), type="n", xlab=expression(beta[1]), ylab="Power",
+ main="sdy=1, x=unif(0,1), intercept=0")
> for (n in seq(from=10,to=50,by=10)) {
+   pwr = sapply(b1s, fpwr, n=n)
+   lines(b1s, pwr, type="b")
+   text(b1s[nb1], pwr[nb1], paste("n=",n,sep=""), adj=-0.3)
+ }
```



**Question 6:** How many simulated datasets are used in drawing this plot? How can we use this plot for designing an experiment?