The Pre-Test Estimator

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Econ 312

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${f 1} \quad {f The \ Model}$

The True Model

$$Model: Y_i = X_i \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \varepsilon_i; \ i = 1, ..., N; \ \varepsilon_i \sim \mathcal{N}(0, \sigma^2) \quad \text{i.i.d.};$$

$$Y_i = X_{1,i}\beta_1 + X_{2,i}\beta_2 + \varepsilon_i$$

$$X_{i} \equiv \begin{bmatrix} X_{1i} \\ X_{2i} \end{bmatrix} \sim \mathcal{N} \begin{bmatrix} 0, & 1 & \rho \\ \rho & 1 \end{bmatrix} \quad \text{i.i.d.}; \quad \begin{bmatrix} \beta_{1} \\ \beta_{2} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

Parameters estimation

Procedure 1 $\begin{bmatrix} \widehat{\beta}_1 \\ \widehat{\beta}_2 \end{bmatrix} \equiv \widehat{\beta}$ are OLS estimation of β using data from the model above.

Procedure 2 $\overline{\beta}_1$ is OLS estimation of $Y_i = \beta_1 X_{1i} + \xi_i$ using data from the true model above

Procedure 3
$$\widetilde{\beta} = \begin{cases} \widetilde{\beta} = \widehat{\beta} \text{ if } |t|_{\widehat{\beta}_2} > 2 \\ \widetilde{\beta} = \overline{\beta}_1 \text{ otherwise} \end{cases}$$
,

where $|t|_{\widehat{\beta}_2}$ means the absolute value of the random variable t-statistic for the estimator $\widehat{\beta}_2$.

The first procedure estimates the OLS parameters. The second procedure estimates a wrong model in which the parameter β_2 is not computed.

The third procedure uses two estimations and the parameter β_2 belongs to the final one only if it is statistically significant. This is the common method of estimating a OLS with a number of variables in order to observe which of them are statistically significant. The second and final estimation uses only the variables which had significant estimated parameters in the first estimation.

Observe that:

$$\widetilde{\beta}_1 = \begin{cases} \widehat{\beta}_1 \text{ with Prob.= } \Pr\left(|t|_{\widehat{\beta}_2} > 2\right) \\ \overline{\beta}_1 \text{ with Prob.= } 1 - \Pr\left(|t|_{\widehat{\beta}_2} > 2\right) \end{cases}$$

The Variance Analysis in the True Model

The OLS Estimation

$$\widehat{\beta} = (X'X)^{-1} X'Y = \beta + (X'X)^{-1} X'\varepsilon$$

$$= \beta + \left(\frac{X'X}{N}\right)^{-1} \frac{X'\varepsilon}{N} \to_p \beta$$

$$\sqrt{N} \cdot (\widehat{\beta} - \beta) \rightarrow d\mathcal{N} (\mathbf{0}, Q_{XX} \cdot \sigma^2)$$

$$Q_{XX} = \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}^{-1} = \frac{1}{1 - \rho^2} \begin{pmatrix} 1 & -\rho \\ -\rho & 1 \end{pmatrix}$$

For N big enough:

$$\widehat{\beta} \sim \mathcal{N} \left(\beta, \frac{\sigma^2}{N} \cdot \frac{1}{1 - \rho^2} \begin{pmatrix} 1 & -\rho \\ -\rho & 1 \end{pmatrix} \right)$$

$$\rho \to 0 \Rightarrow var(\widehat{\beta}) \to \frac{\sigma^2}{N} \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$\rho \to \pm 1 \Rightarrow var(\widehat{\beta}) \to \infty \text{ (perfect colinearity)}$$

$$\widehat{\beta}_1 \sim \mathcal{N}\left(\beta_1, \frac{\sigma^2}{N} \cdot \frac{1}{1-\rho^2}\right)$$

$$\widehat{\beta}_2 \sim \mathcal{N}\left(\beta_2, \frac{\sigma^2}{N} \cdot \frac{1}{1-\rho^2}\right)$$

3 The Bias and Variance Analysis in the second Procedure

The Exclusion of β_2 Estimation

$$\overline{\beta}_1 = (X_1'X_1)^{-1} X_1'Y = \frac{\sum_{i=1}^N X_{1,i}Y_i}{\sum_{i=1}^N X_{1,i}^2}$$

$$= \frac{\sum_{i=1}^{N} (X_{1,i}\beta_1 + X_{2,i}\beta_2 + \varepsilon_i) X_{1,i}}{\sum_{i=1}^{N} X_{1,i}^2}$$

$$= \beta_1 + \frac{\beta_2 \sum_{i=1}^{N} X_{1,i} X_{2,i} + \sum_{i=1}^{N} \varepsilon_i X_{1,i}}{\sum_{i=1}^{N} X_{1,i}^2}$$

$$= \beta_1 + \frac{\beta_2 \sum_{i=1}^{N} \frac{X_{1,i} X_{2,i}}{N} + \sum_{i=1}^{N} \frac{\varepsilon_i X_{1,i}}{N}}{\sum_{i=1}^{N} \frac{X_{1,i}^2}{N}}$$

But:

$$\sum_{i=1}^{N} \frac{\varepsilon_{i} X_{1,i}}{N} \to_{p} E\left(\varepsilon_{i} X_{1,i}\right) = 0 ;$$

$$\sum_{i=1}^{N} \frac{X_{1,i} X_{2,i}}{N} \to_{p} E\left(X_{1,i} X_{2,i}\right) = \rho \quad \text{(WLLN)}$$

$$\sum_{i=1}^{N} X_{1,i}^{2} = \mathcal{X}_{N}^{2} \Rightarrow \sum_{i=1}^{N} \frac{X_{1,i}^{2}}{N} \to_{p} 1$$

$$\Rightarrow \overline{\beta}_{1} \to_{p} \beta_{1} + \rho \beta_{2}$$

The absolute bias in the second procedure gets bigger as

$$|\rho| \rightarrow_p 1.$$

The variance of the second procedure:

The $\overline{\beta}_1$ Variance: Exclusion of β_2

$$assvar(\overline{\beta}_{1}) = assvar((X'_{1}X_{1})^{-1}X'_{1}Y)$$

$$= assvar((X'_{1}X_{1})^{-1}X'_{1}(X_{1}\beta_{1} + X_{2}\beta_{2} + \varepsilon))$$

$$= assvar(\beta_{1} + \beta_{2}(X'_{1}X_{1})^{-1}X'_{1}X_{2} + (X'_{1}X_{1})^{-1}X'_{1}\varepsilon)$$

$$= p \lim \left(\beta_{2}^{2} \left[\left(\frac{X_{1}'X_{1}}{N} \right)^{-1} \frac{X_{1}'X_{2}}{\sqrt{N}} \right) - \rho \right]^{2} \right)$$

$$+p \lim \left(\left[\left(\frac{X_{1}'X_{1}}{N} \right)^{-1} \frac{X_{1}'\varepsilon}{\sqrt{N}} \right]^{2} \right)$$

$$= E(X_{i,1})^{-1} \cdot \left(\beta_{2}^{2} \cdot var[X_{i,1}X_{i,2}] + \sigma^{2} \right)$$

$$= \beta_{2}^{2} \cdot var[X_{i,1}X_{i,2}] + \sigma^{2}$$

$$= \beta_{2}^{2} \cdot \left(E[X_{i,1}^{2}X_{i,2}^{2}] - \rho^{2} \right) + \sigma^{2}$$

Observe that:

$$X_1 = \mathcal{P}_{X_2}(X_1) + \mathcal{P}_{\perp X_2}(X_1)$$

 $X_1 = \rho X_2 + v; \ v \perp X_2$

But X_2, X_1 are normals $\Rightarrow v$ is normal

$$var[X_1] = 1 = \rho^2 + var[v]$$

$$\Rightarrow var[v] = 1 - \rho^2 \text{ and } E[v] = 0.$$

(The Jacobian method of change of variables can also provide the same result: $v = X_1 - \rho X_2$.)

$$\Rightarrow E(X_1^2 X_2^2) = E((\rho X_2 + v)^2 \cdot X_2^2)$$

but:

$$E\left(X_{2}^{j}\right) = \begin{cases} \frac{j}{(j/2)!} \cdot \frac{\sigma_{\beta_{2}}^{j}}{2^{j/2}} \text{ for } j \text{ even} \\ 0 \text{ for } j \text{ odd} \end{cases}$$

$$\Rightarrow E(X_2^2) = 3\sigma_{\beta_2}^4 = 3$$

$$\Rightarrow assvar(\overline{\beta}_1) = \beta_2^2 \cdot (1 - \rho^2) + \sigma^2$$

For N big enough:

$$\overline{\beta}_1 \approx \mathcal{N}\left(\beta_1 + \rho\beta_2, \frac{\sigma^2}{N} + \frac{\beta_2^2 \cdot (1 - \rho^2)}{N}\right)$$

The Pre-Test Estimator has the distribution:

$$\widetilde{\beta}_1 \sim \begin{cases} \mathcal{N}\left(\beta_1, \frac{\sigma^2}{N} \cdot \frac{1}{1-\rho^2}\right) \text{ with Prob.= } \Pr\left(|t|_{\widehat{\beta}_2} > 2\right) \\ \mathcal{N}\left(\beta_1 + \rho\beta_2, \frac{\sigma^2}{N} + \frac{\beta_2^2 \cdot \left(1-\rho^2\right)}{N}\right) \text{ with Prob.= } 1 - \Pr\left(|t|_{\widehat{\beta}_2} > 2\right) \end{cases}$$

for N large enough

The bias (due to $\rho\beta_2 \neq 0$) depends on

- the level of the correlation ρ ;
- the value of β_2 ;
- the size of $\rho\beta_2$;
- the probability $\Pr\left(|t|_{\widehat{\beta}_2} < 2\right)$

Observe that:

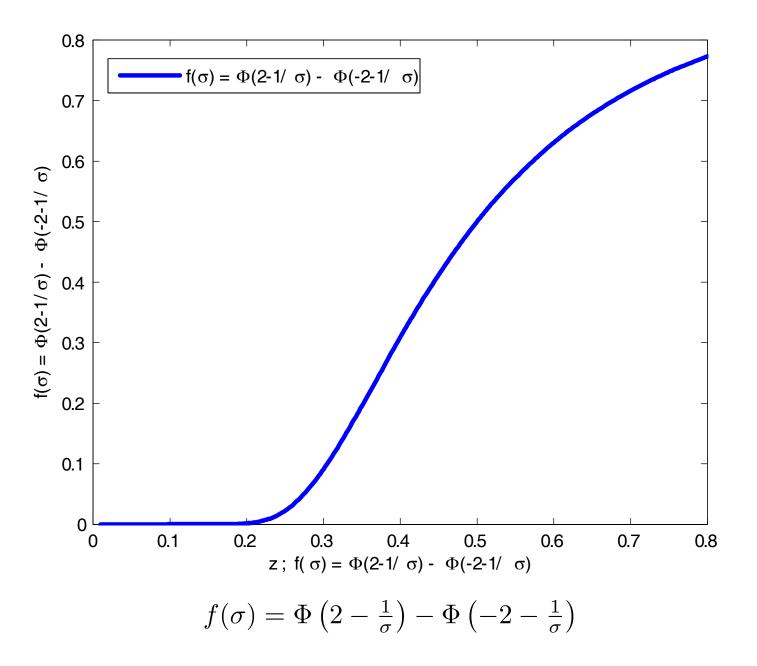
$$\widehat{\beta}_2 \sim \mathcal{N}\left(\beta_2, \frac{\sigma^2}{N} \cdot \frac{1}{1 - \rho^2}\right) \Rightarrow$$

$$\Pr\left(|t|_{\widehat{\beta}_2} < 2\right) \approx \Phi\left(2 - \frac{\beta_2}{\sqrt{\frac{\sigma^2}{N} \cdot \frac{1}{1-\rho^2}}}\right) - \Phi\left(-2 - \frac{\beta_2}{\sqrt{\frac{\sigma^2}{N} \cdot \frac{1}{1-\rho^2}}}\right)$$

The variance of ε_i which affects $|t|_{\widehat{\beta}_2}$; $\sigma^2 \uparrow \Rightarrow \Pr\left(|t|_{\widehat{\beta}_2} < 2\right) \uparrow$ The Expectation of $E(X_i'X_i)$ which affects $|t|_{\widehat{\beta}_2}$.

An increase in $|\rho|$ increases the variance of $\widehat{\beta}_2$ which increases $\Pr\left(|t|_{\widehat{\beta}_2} < 2\right)$

Example of Function



Suppose:

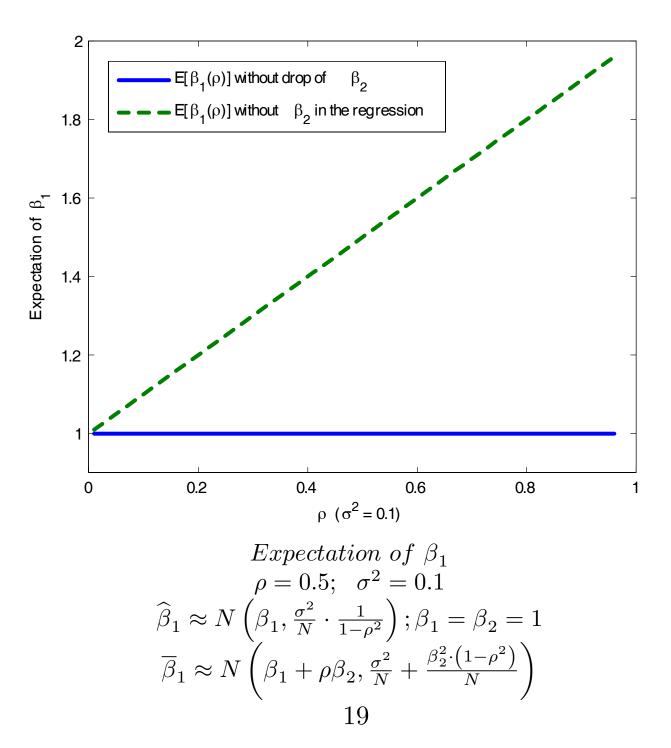
$$Y_{i} = X_{i} \begin{bmatrix} \beta_{1} \\ \beta_{2} \end{bmatrix} + \varepsilon_{i}; i = 1, ..., N;$$

$$\varepsilon_{i} \sim \mathcal{N}(0, \sigma^{2}) \quad \text{i.i.d.}; \quad \sigma^{2} = 0.1$$

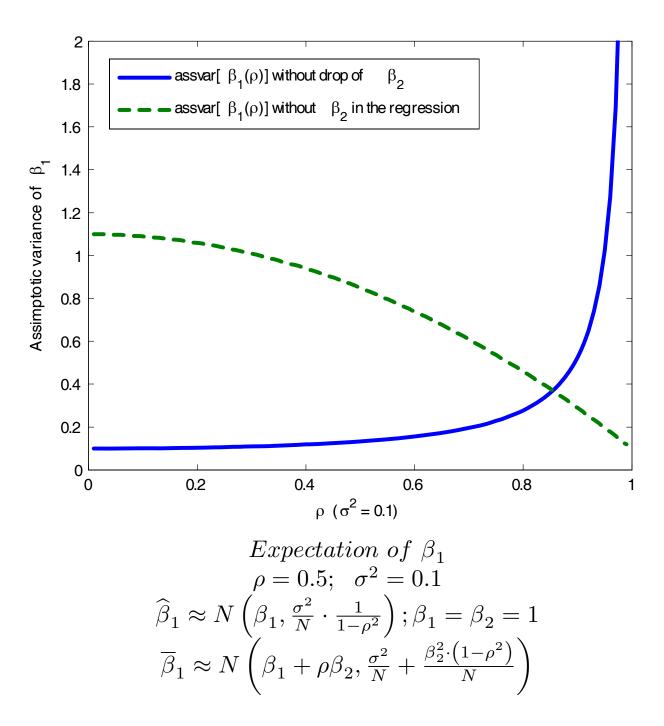
$$X_{i} \equiv \begin{bmatrix} X_{1i} \\ X_{2i} \end{bmatrix} \sim \mathcal{N} \begin{bmatrix} 0, \frac{1}{\rho} & \rho \\ 0 & 1 \end{bmatrix} \quad \text{i.i.d.}; \quad \begin{bmatrix} \beta_{1} \\ \beta_{2} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

We obtain the results:

Expectation of $\widehat{\beta}_1$, $\overline{\beta}_1$,



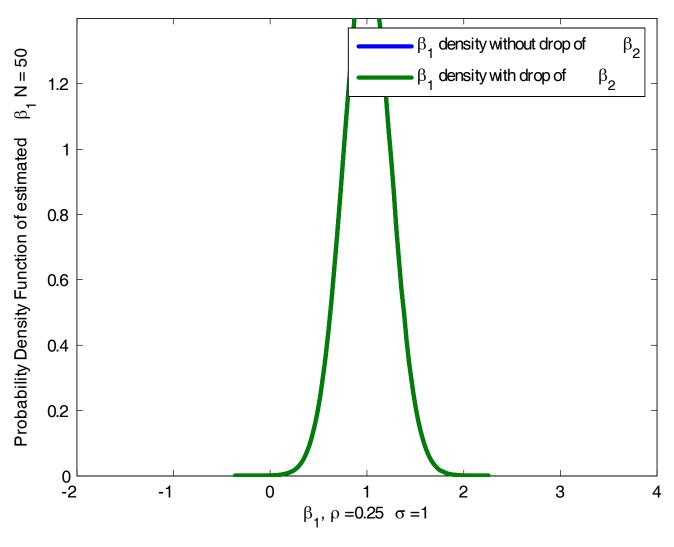
Asymptotic Variance of $\widehat{\beta}_1$, $\overline{\beta}_1$,



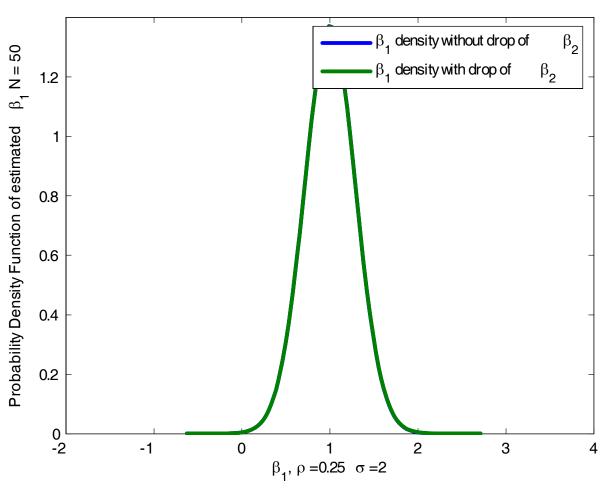
4 The Parameter Analysis of Procedures 1 and 3

This section provides the following facts for $\rho > 0$, $\beta_2 > 0$.

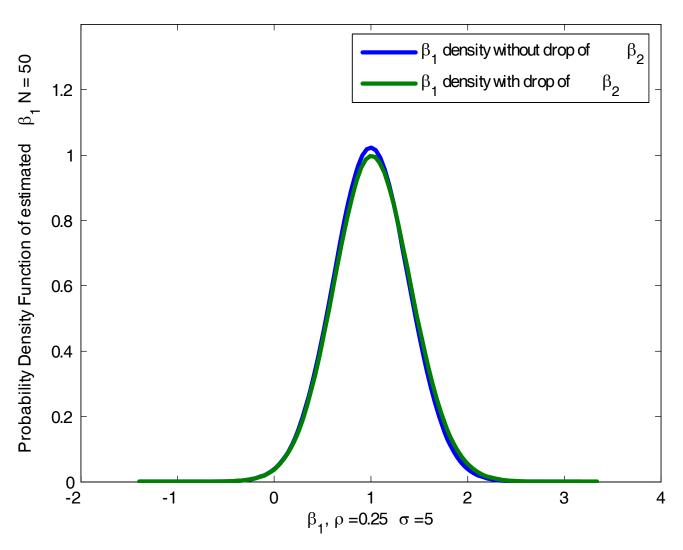
- 1. The increase in the error variance makes the OLS and Pre-Test Estimators less precise.
- 2. The increase in the error variance makes the Pre-Test more biased.
- 3. The increase in the correlation among variables X makes the OLS and Pre-Test Estimators less precise.
- 4. The increase in the correlation among variables X makes Pre-Test Estimators more biased.



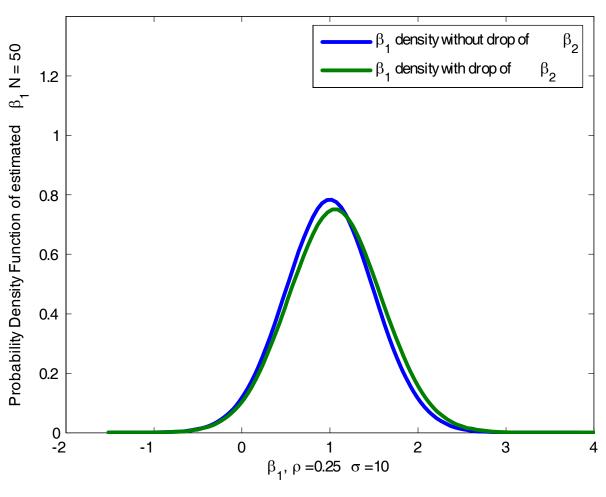
 $\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \rho=0.25; \ \ \sigma^2=1 \end{array}$



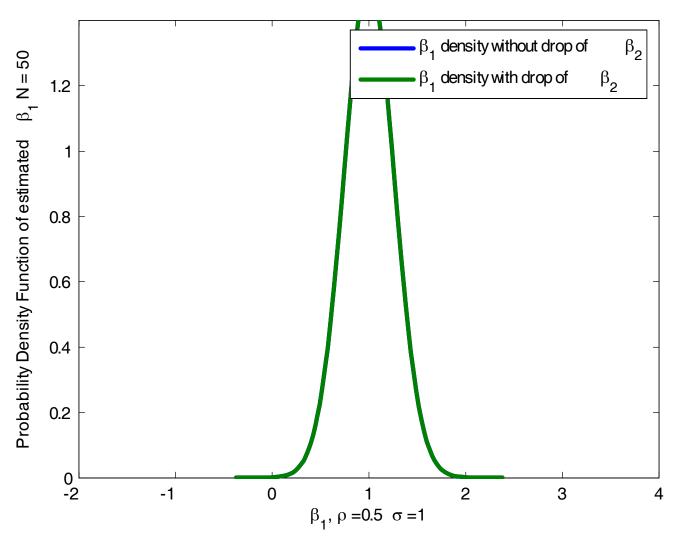
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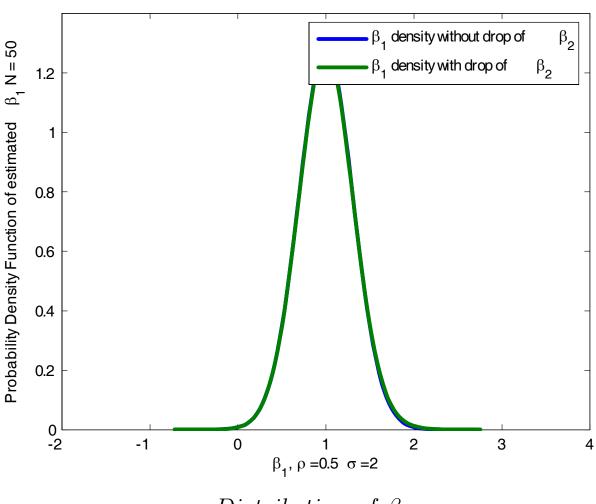
$$\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \rho=0.25; \ \ \sigma^2=5 \end{array}$$



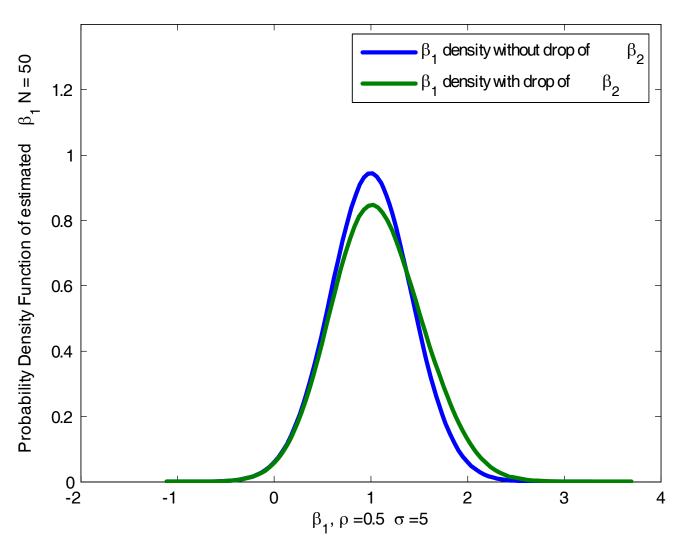
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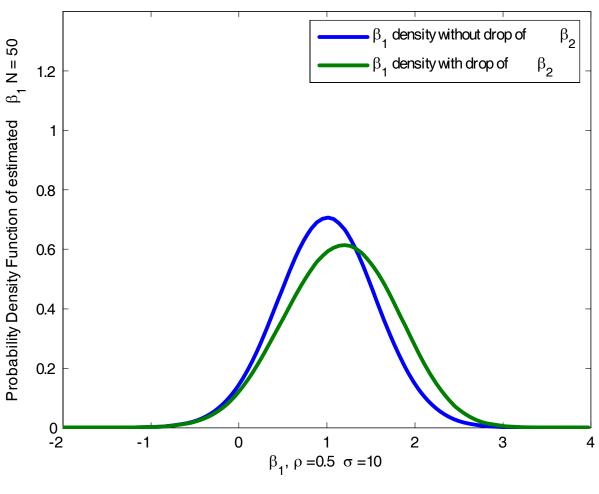
$$\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \rho=0.5; \ \ \sigma^2=1 \end{array}$$



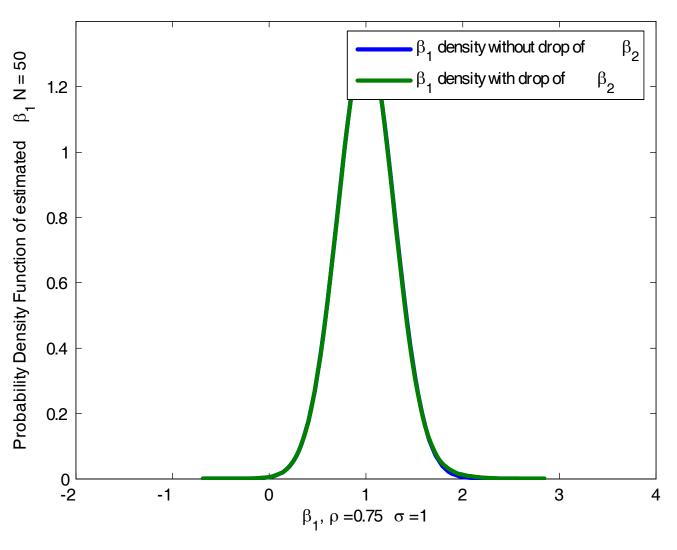
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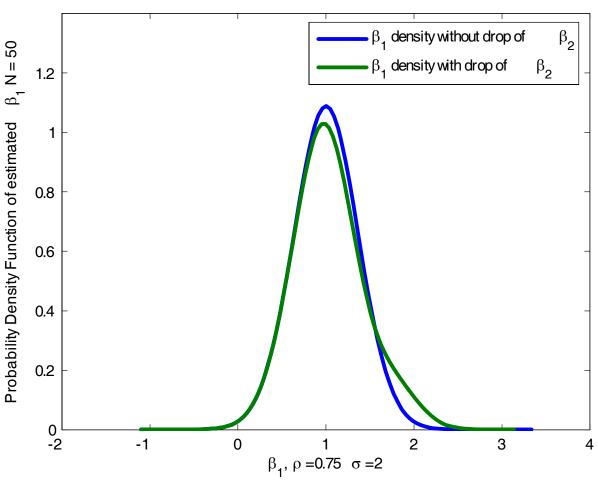
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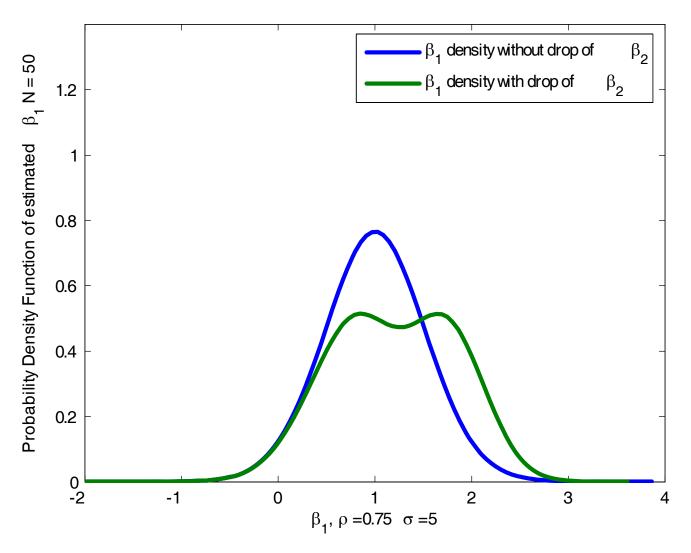
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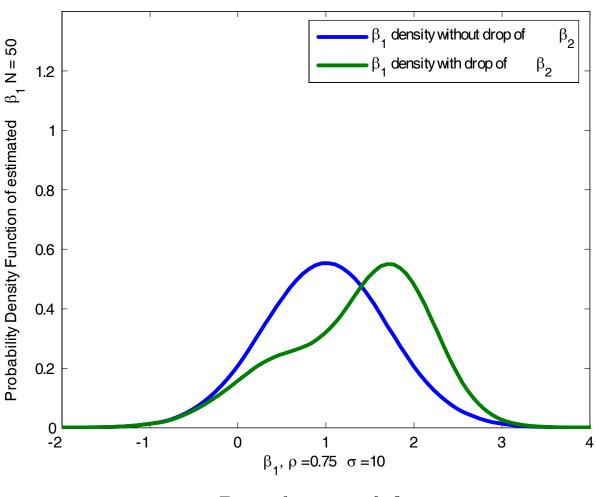
 $\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \rho=0.75; \ \ \sigma^2=1 \end{array}$



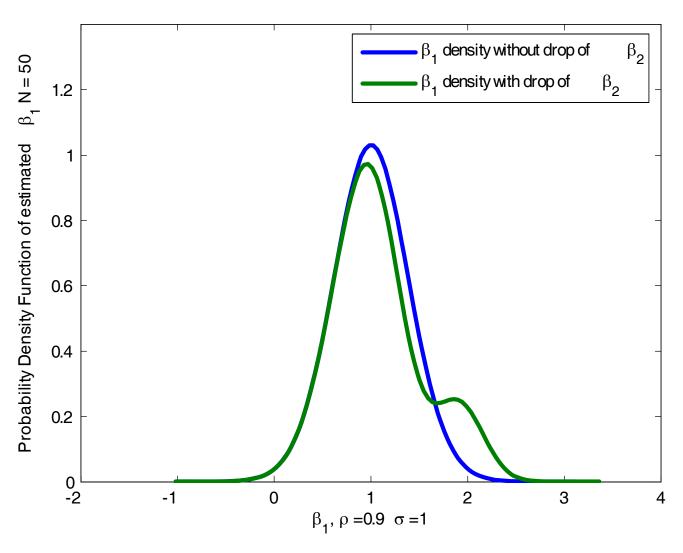
 $\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \rho=0.75; \ \ \sigma^2=2 \end{array}$



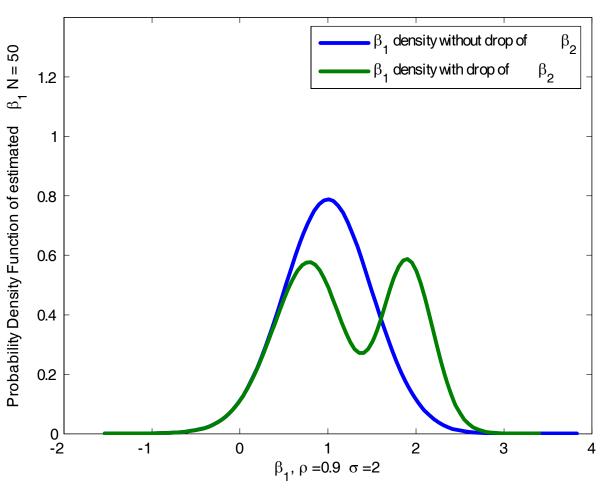
 $Distribution \ of \ \beta_1 \\ N = 50, \ \ \rho = 0.75; \ \ \sigma^2 = 5$



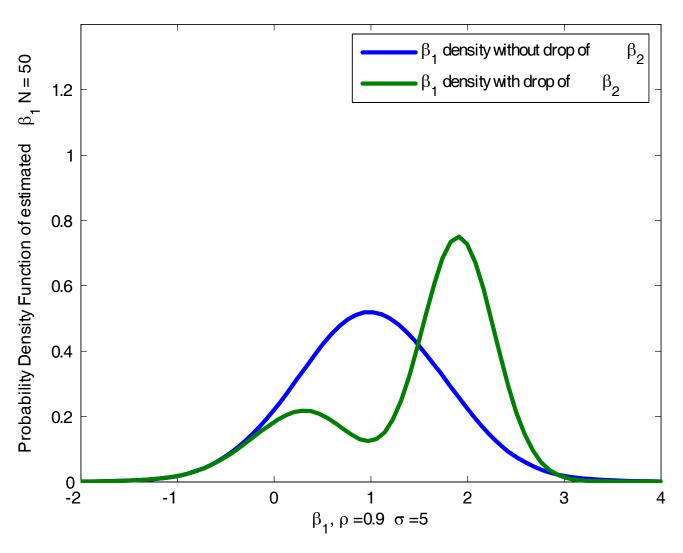
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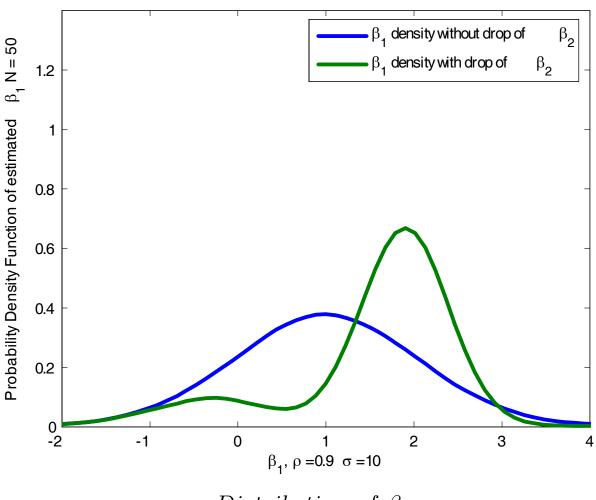
 $\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \rho=0.9; \ \ \sigma^2=1 \end{array}$



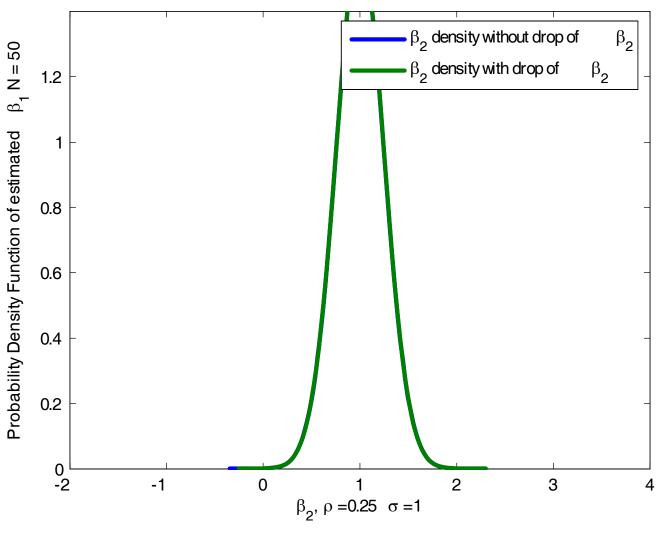
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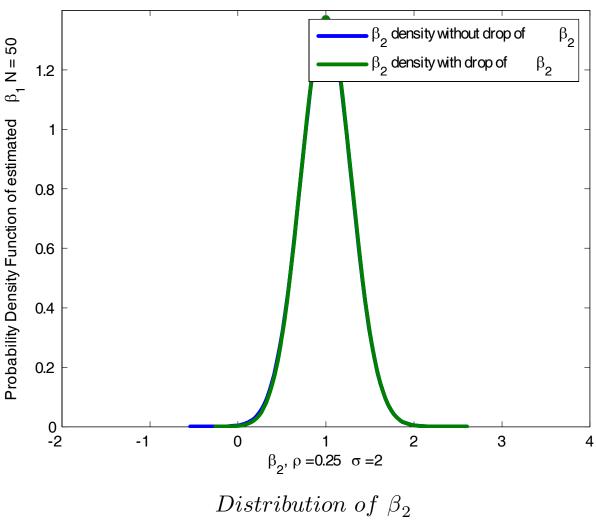
$$\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \rho=0.9; \ \ \sigma^2=5 \end{array}$$



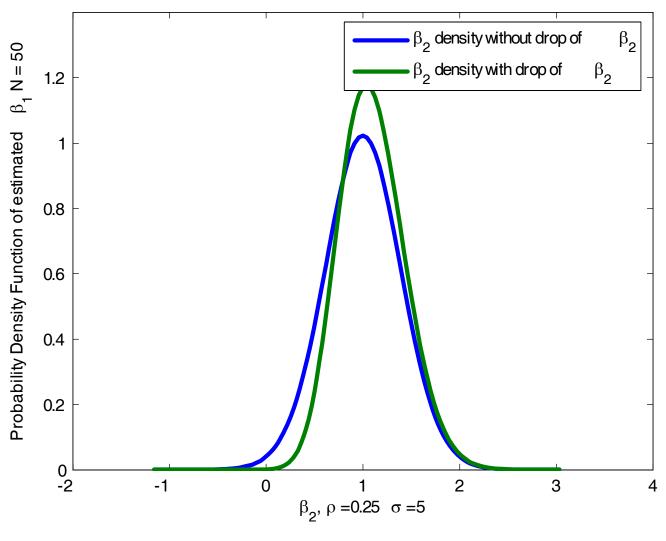
 $Distribution \ of \ \beta_1 \\ N = 50, \ \ \rho = 0.9; \ \ \sigma^2 = 10$



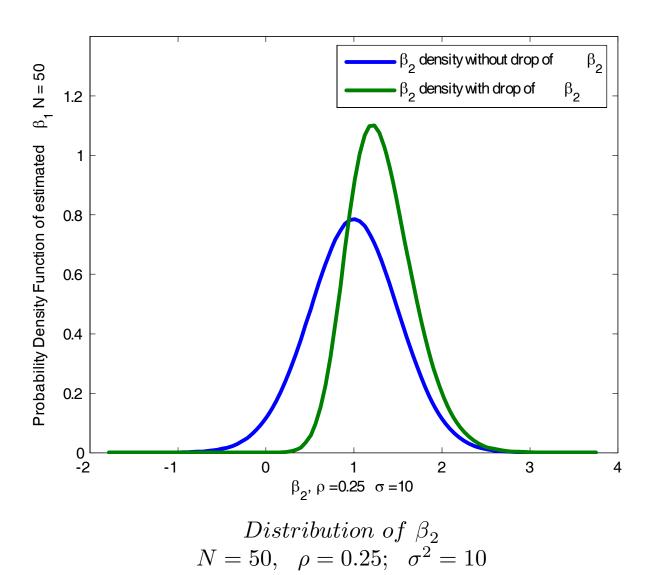
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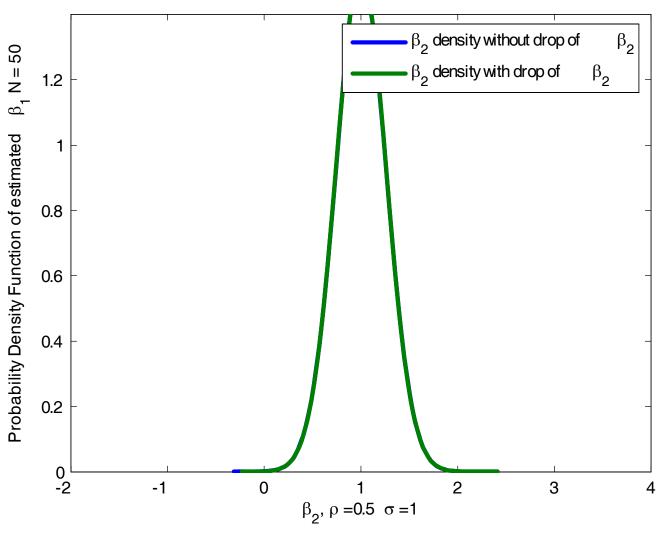


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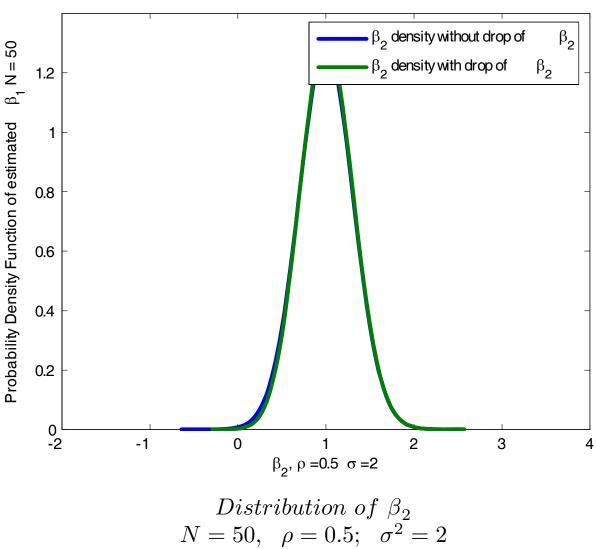


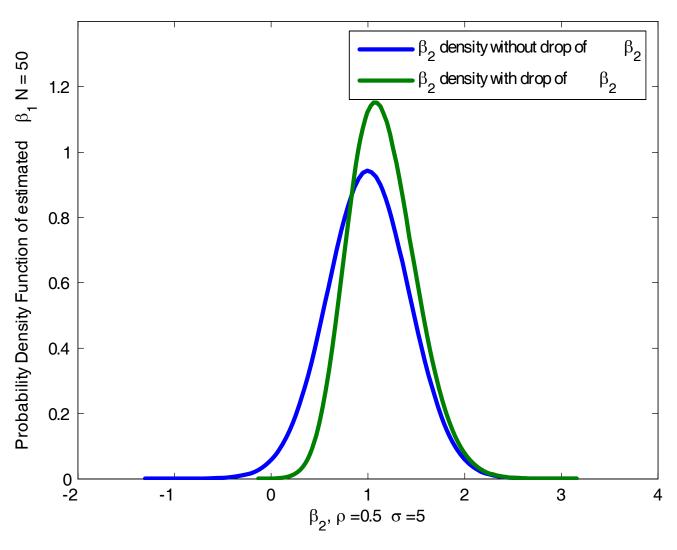
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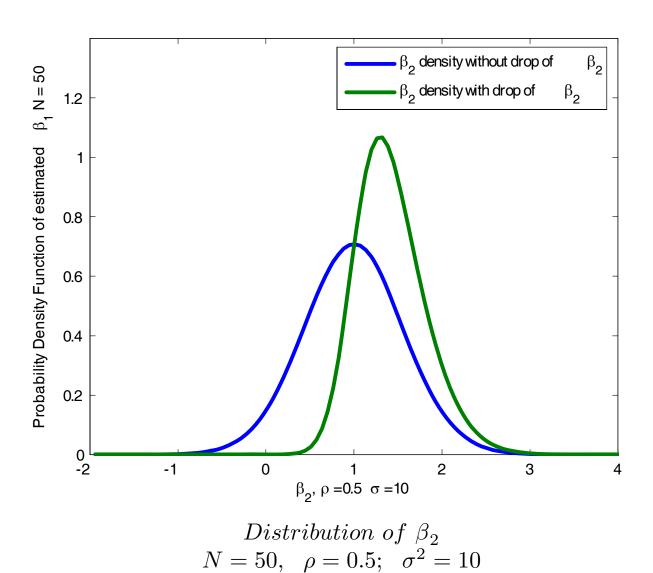


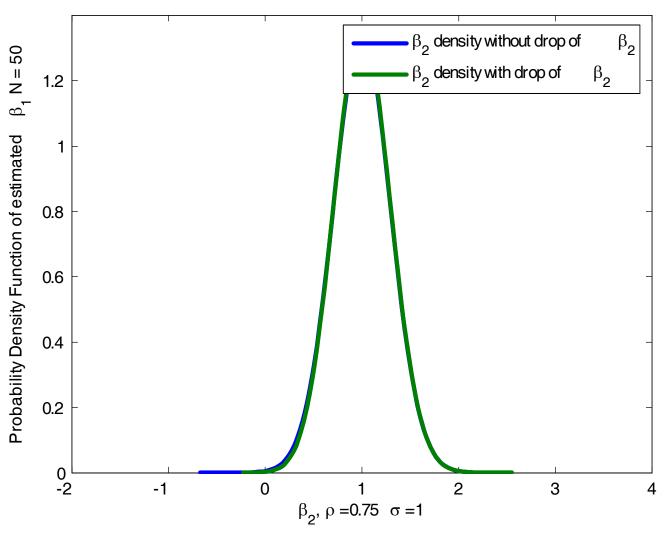
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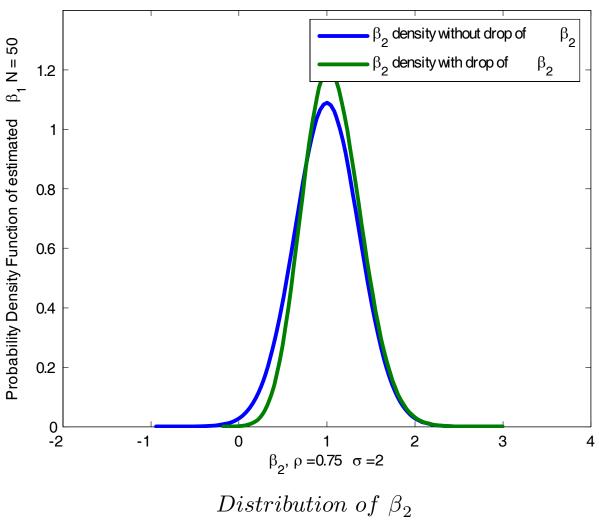


$$Distribution \ of \ \beta_2 \\ N=50, \ \ \rho=0.5; \ \ \sigma^2=5$$

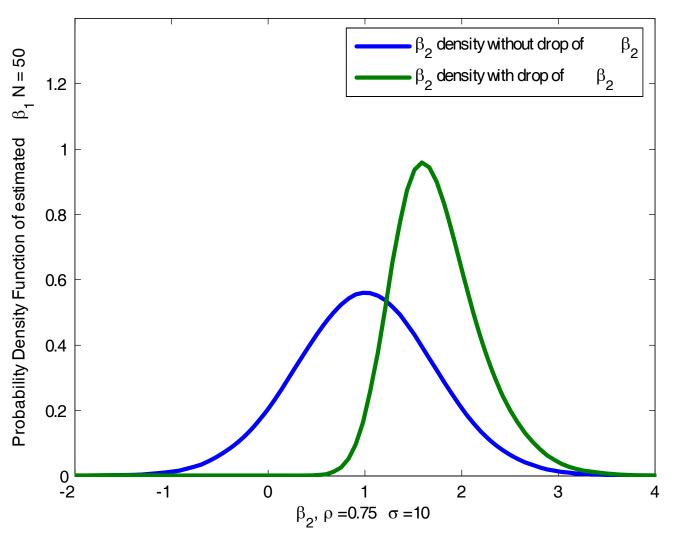




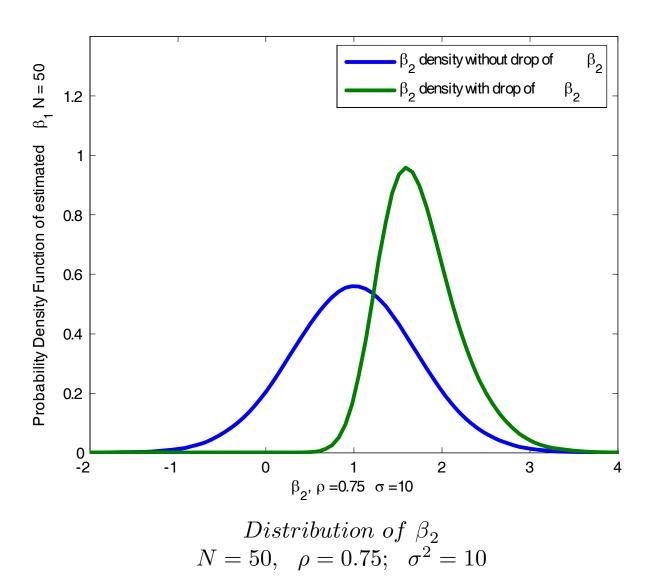
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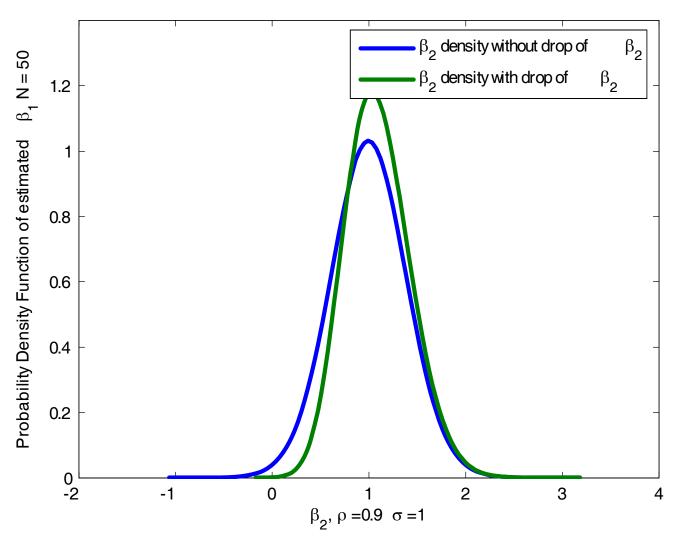


 $Distribution \ of \ \beta_2 \\ N = 50, \ \ \rho = 0.75; \ \ \sigma^2 = 2$

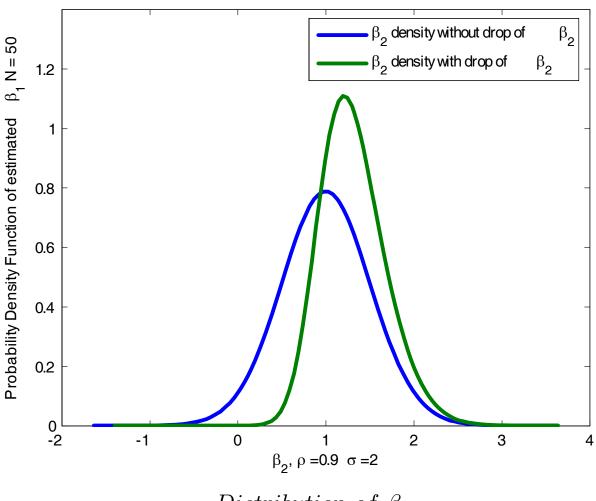


 $Distribution \ of \ \beta_2 \\ N = 50, \ \ \rho = 0.75; \ \ \sigma^2 = 5$

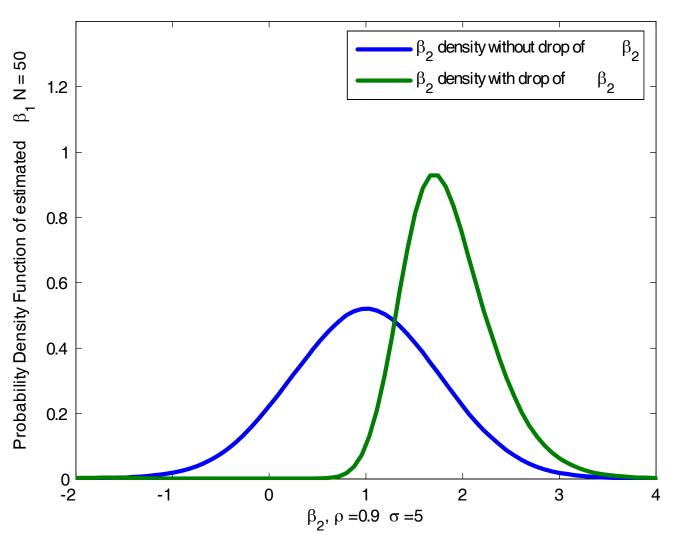




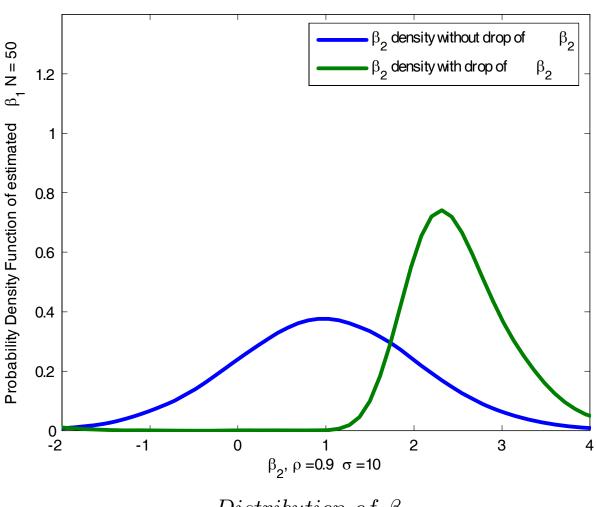
$$\begin{array}{c} Distribution \ of \ \beta_2 \\ N=50, \ \ \rho=0.9; \ \ \sigma^2=1 \end{array}$$



 $\begin{array}{c} Distribution \ of \ \beta_2 \\ N=50, \ \ \rho=0.9; \ \ \sigma^2=2 \end{array}$



 $\begin{array}{c} Distribution \ of \ \beta_2 \\ N=50, \ \ \rho=0.9; \ \ \sigma^2=5 \end{array}$



$$Distribution \ of \ \beta_2 \\ N = 50, \ \ \rho = 0.9; \ \ \sigma^2 = 10$$

5 The β_1 Analysis of Procedures 1, 2 and 3 for a specific Model

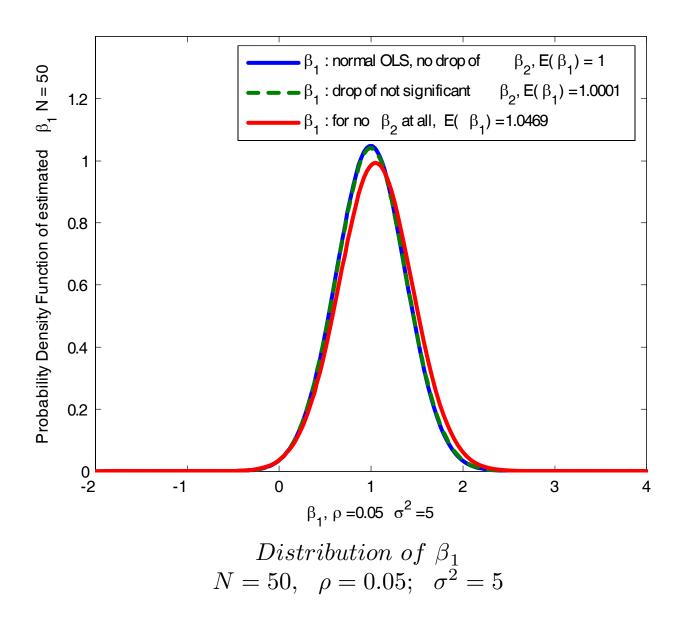
Let the model be:

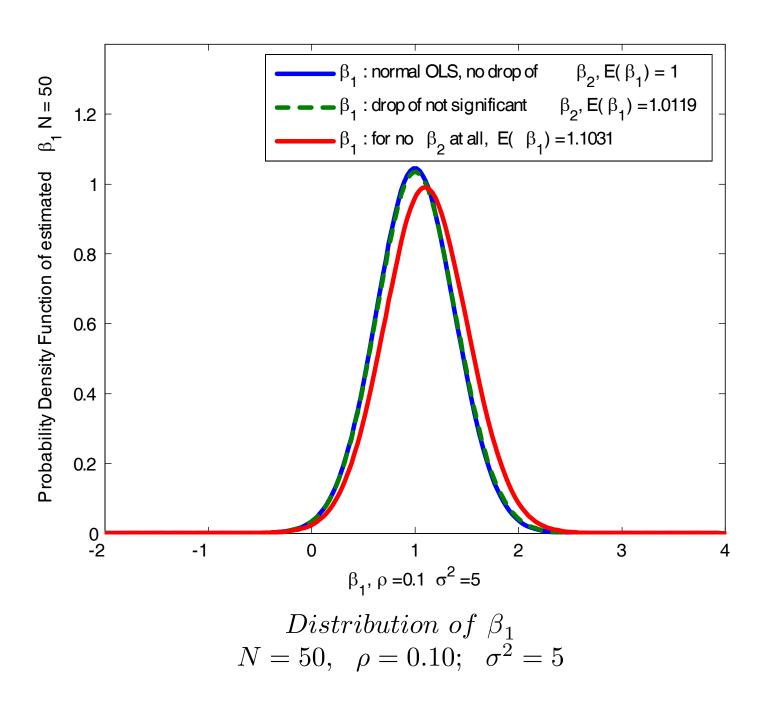
The True Model:

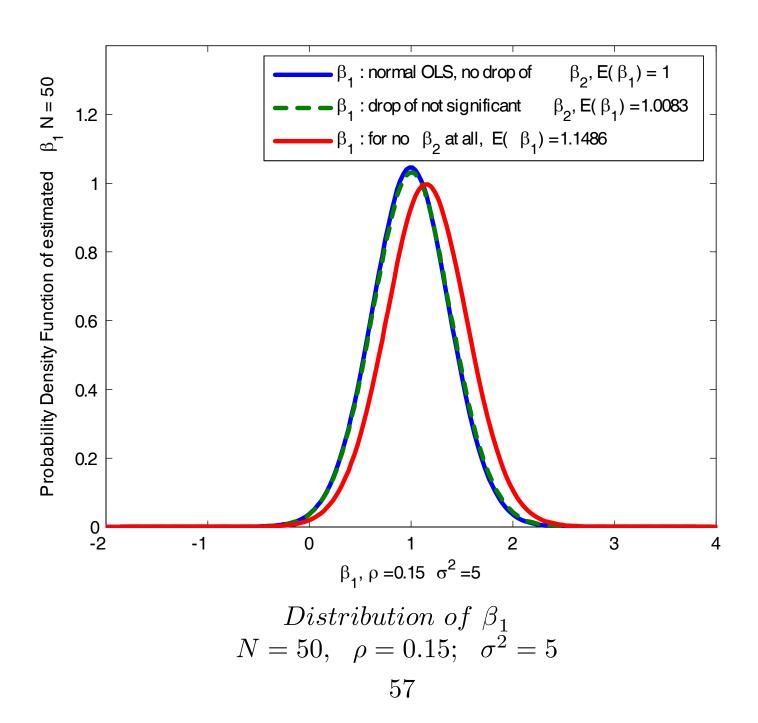
$$Y_{i} = X_{i} \begin{bmatrix} 1 \\ 1 \end{bmatrix} + \varepsilon_{i}; \ i = 1, ..., 50; \quad \varepsilon_{i} \sim \mathcal{N}(0, 5) \quad \text{i.i.d.};$$

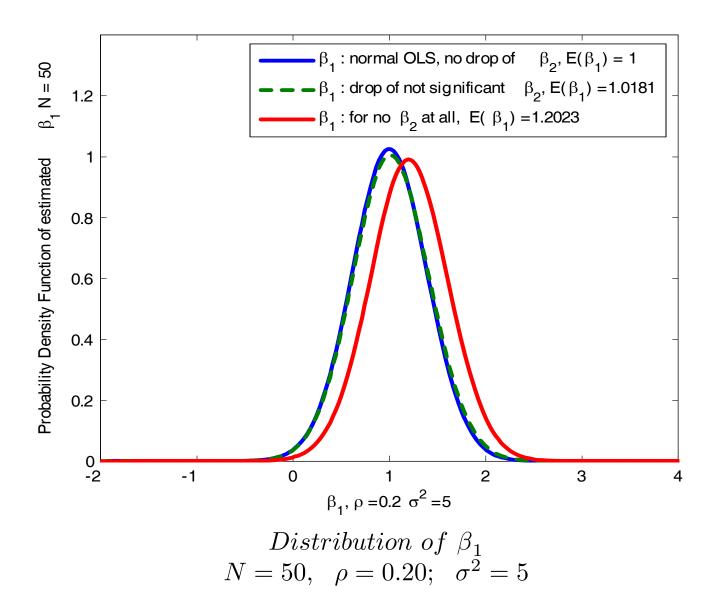
$$Y_{i} = X_{1,i} + X_{2,i} + \varepsilon_{i}; \quad \sigma^{2} = 5; \quad N = 50;$$

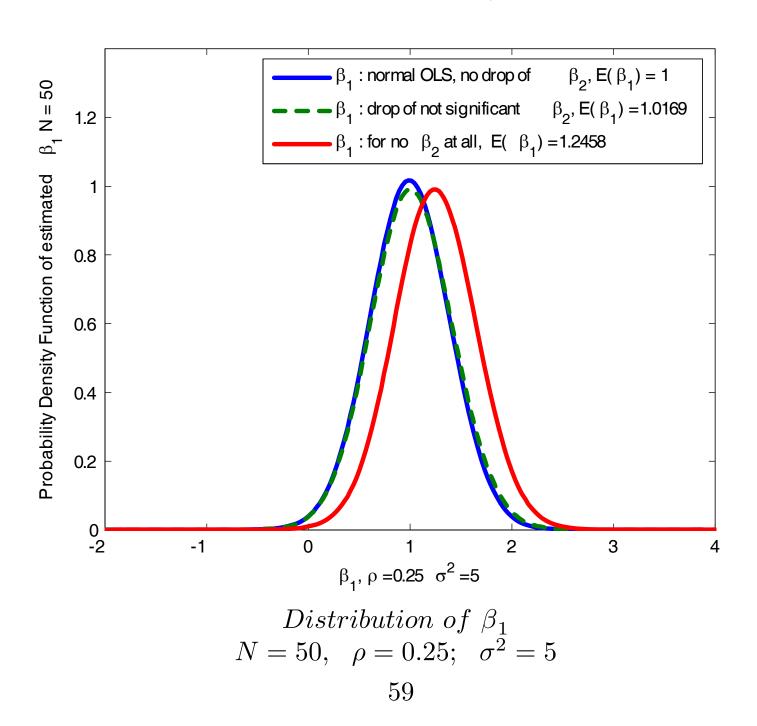
$$X_{i} \equiv \begin{bmatrix} X_{1i} \\ X_{2i} \end{bmatrix} \sim \mathcal{N} \begin{bmatrix} 0, \frac{1}{\rho} & \rho \\ 0 & 1 \end{bmatrix} \quad \text{i.i.d.}; \quad \begin{bmatrix} \beta_{1} \\ \beta_{2} \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

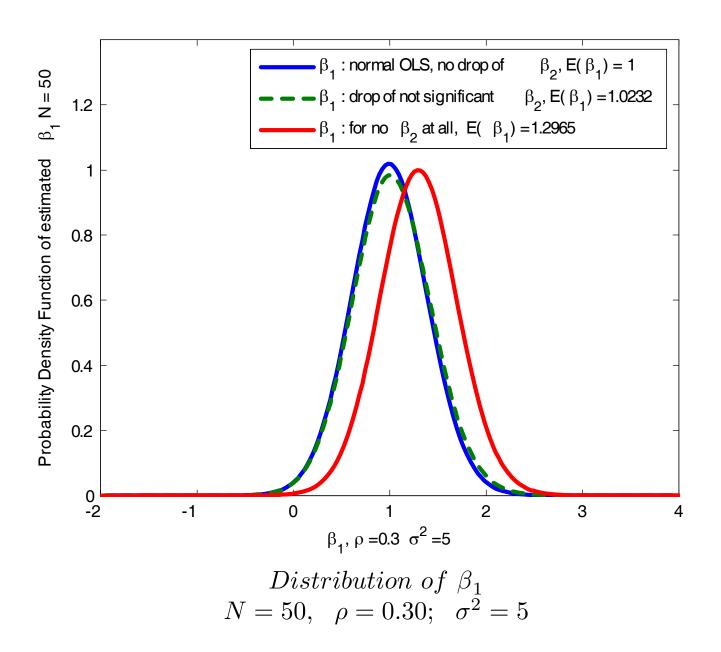


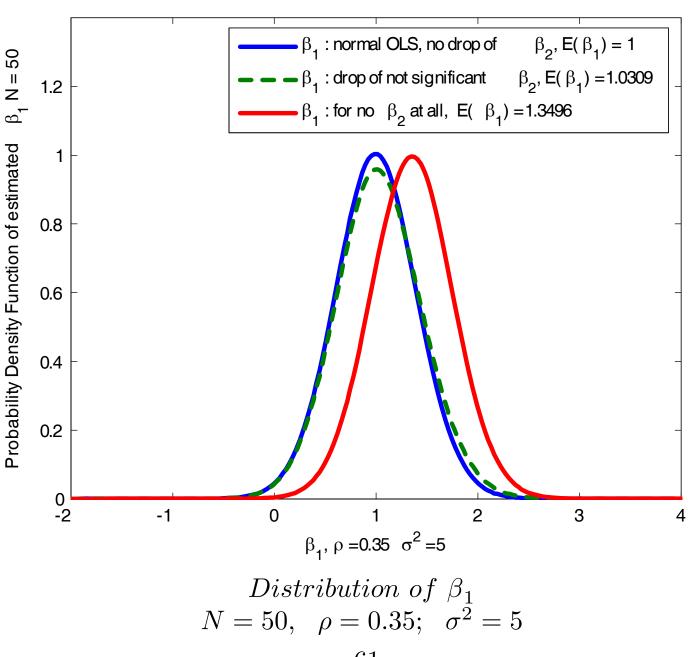


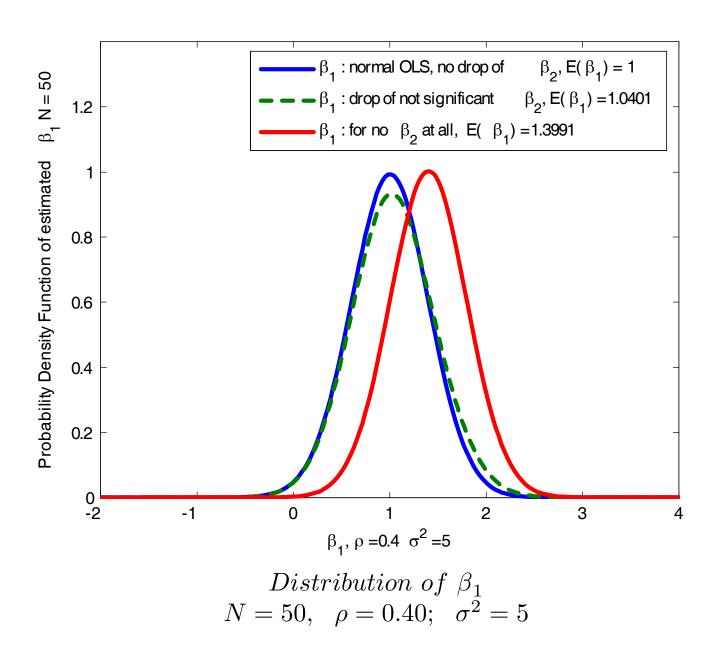


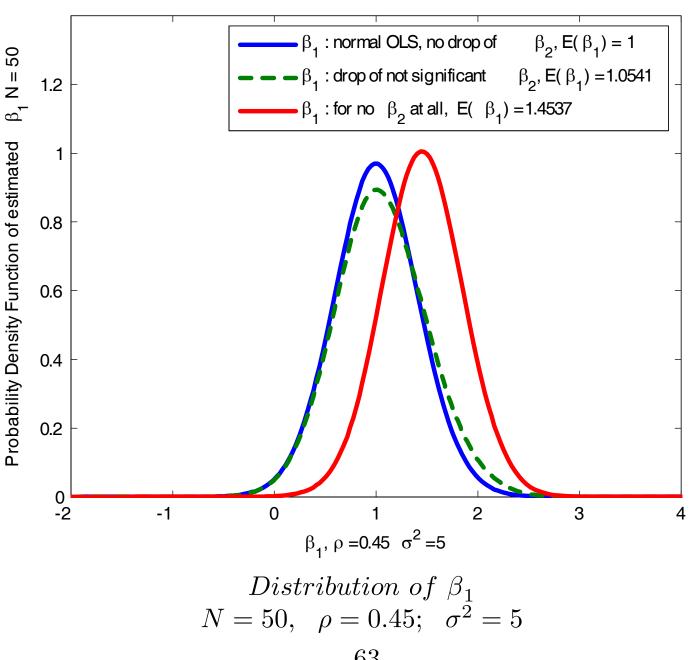


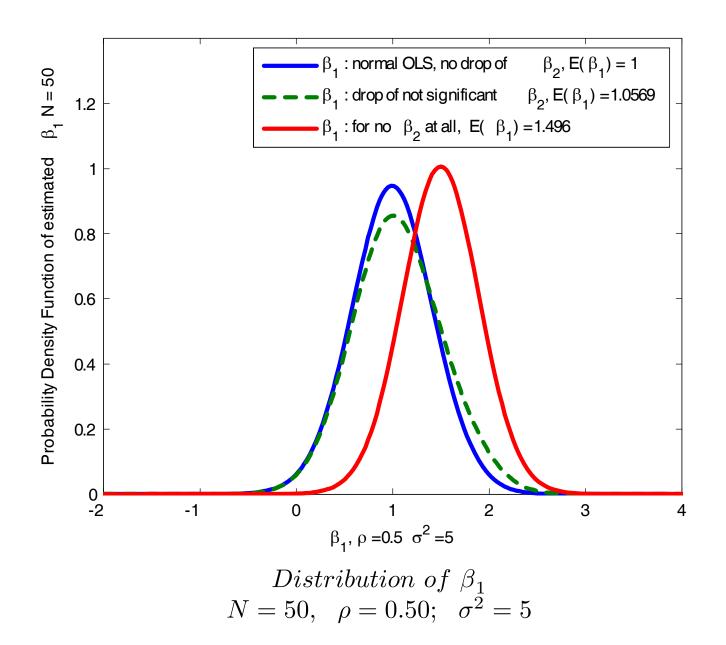


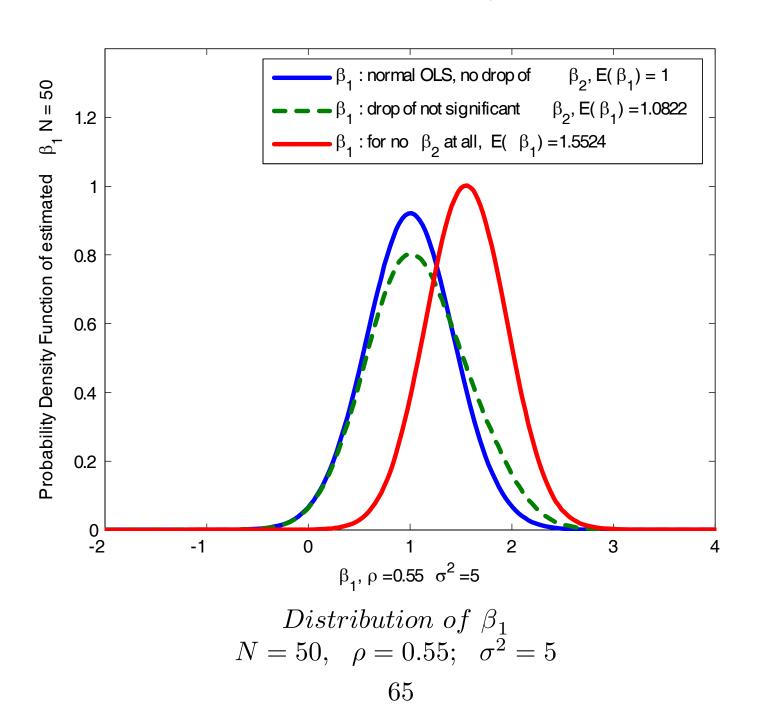


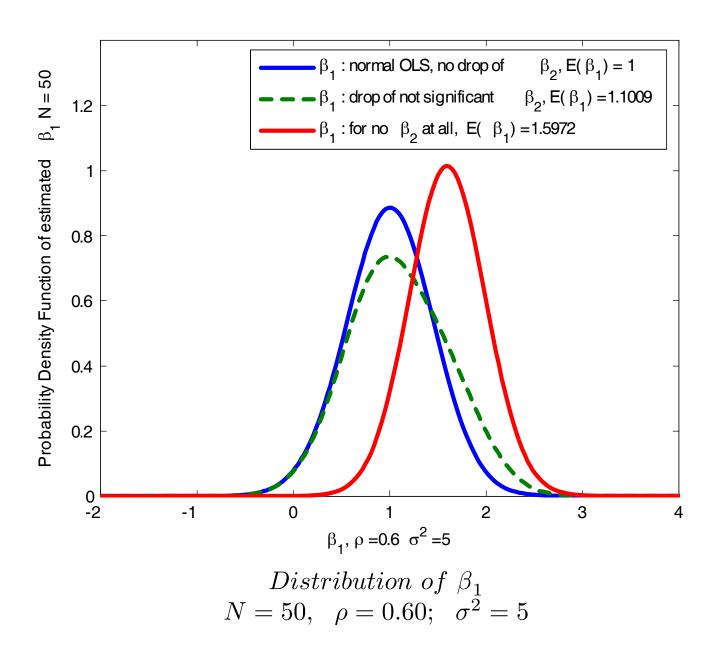


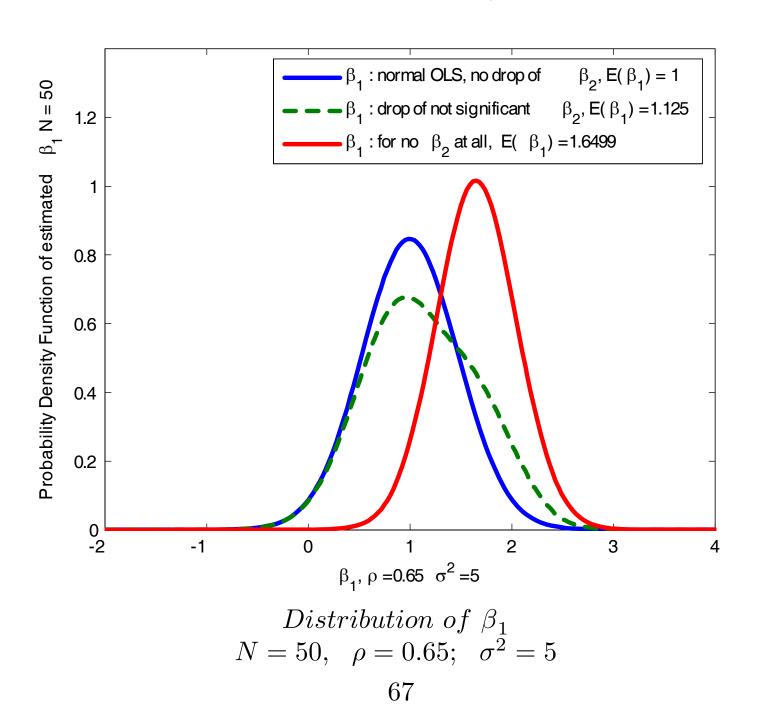


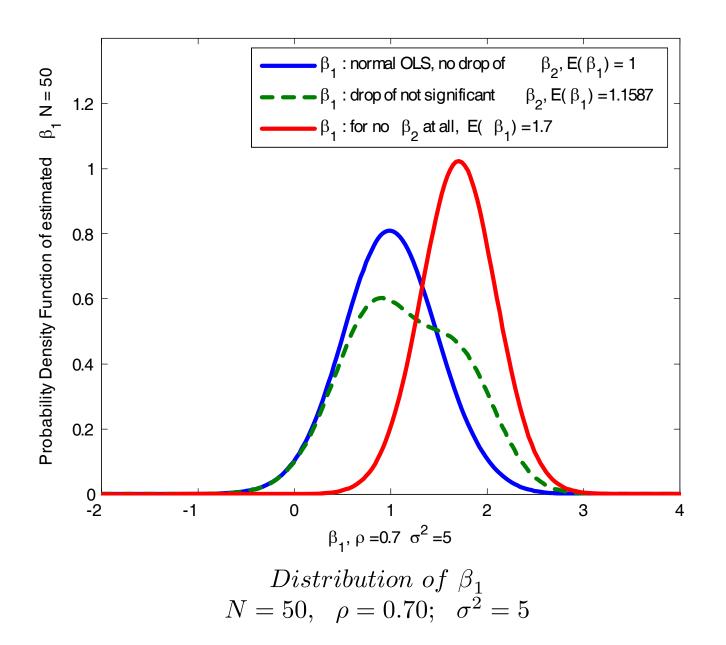


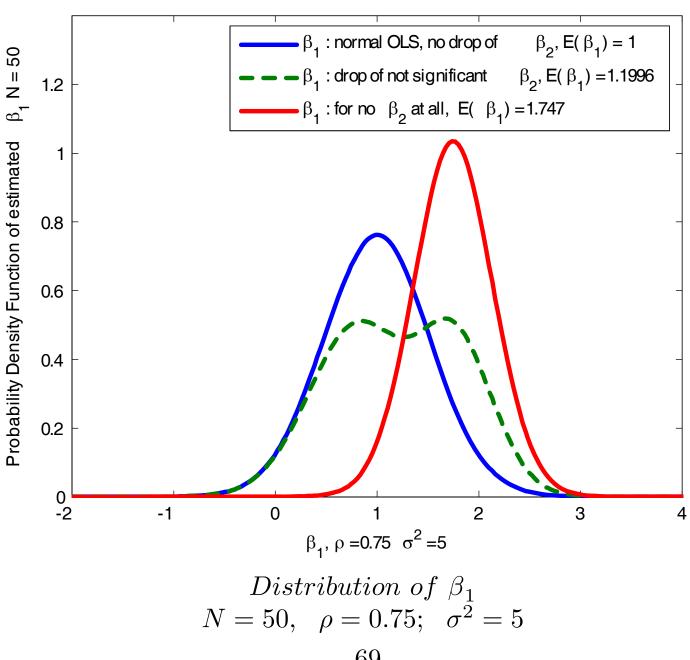


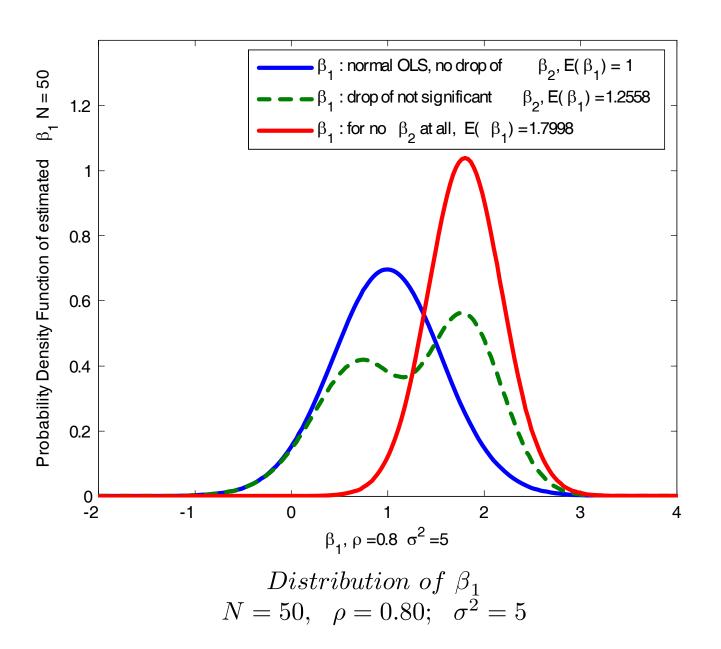


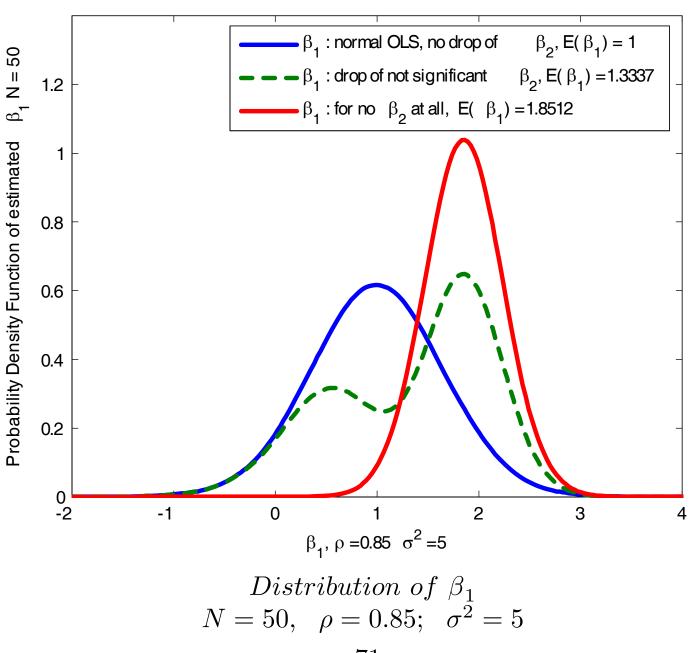


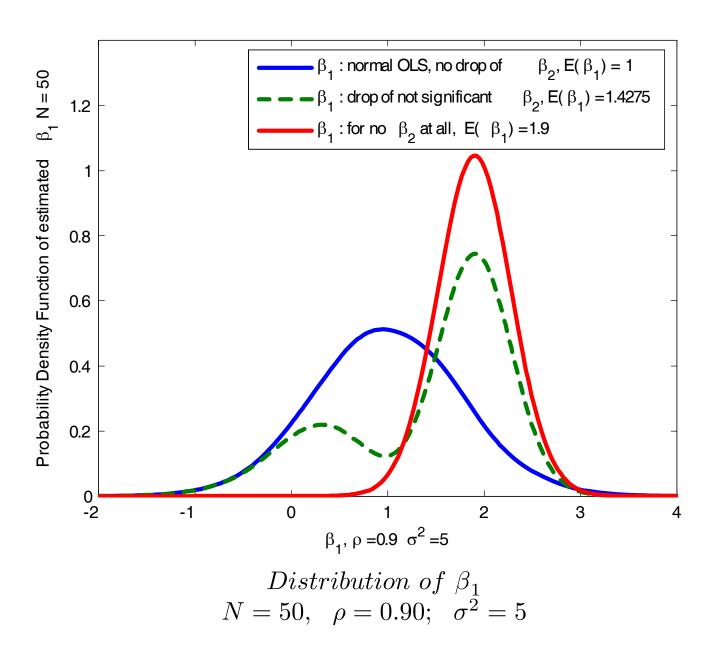




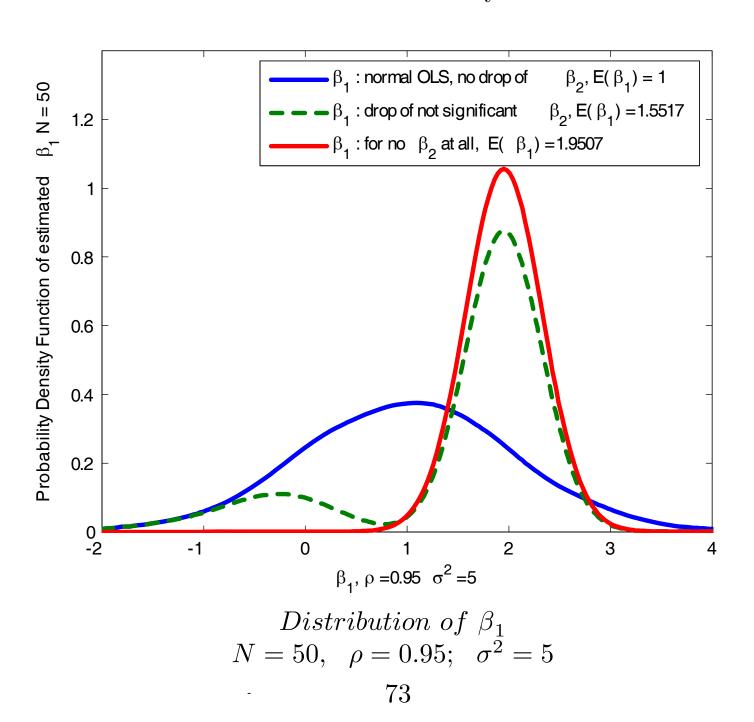




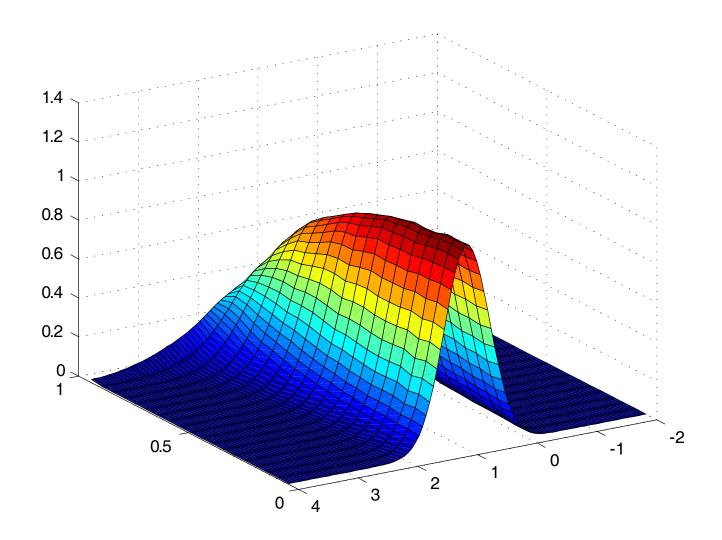




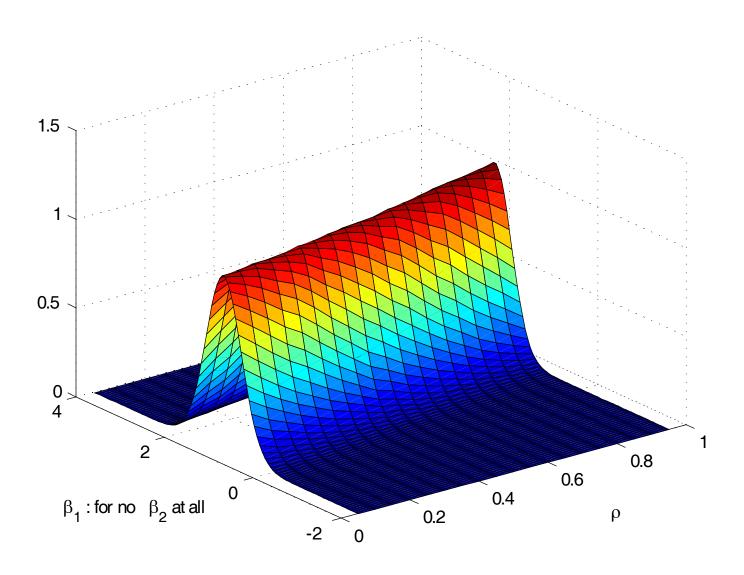
Pre-Test Estimator Density Function



OLS $\widehat{\beta}_1$ Estimator Density Function

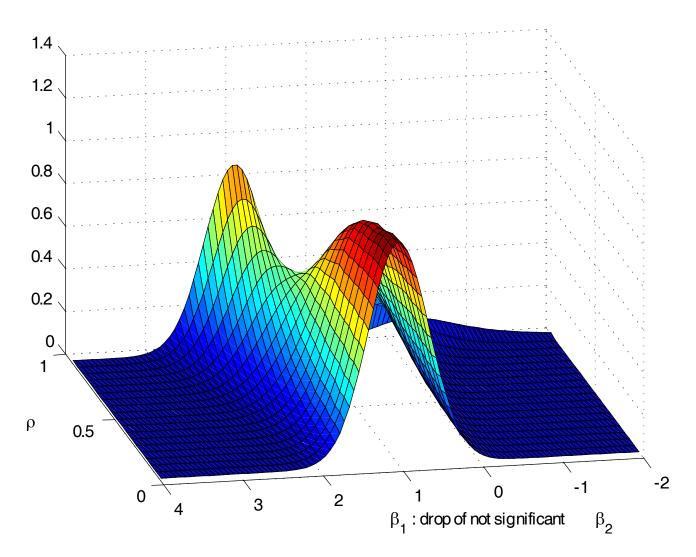


Procedure 2 $\overline{\beta}_1$ Estimator Density Function



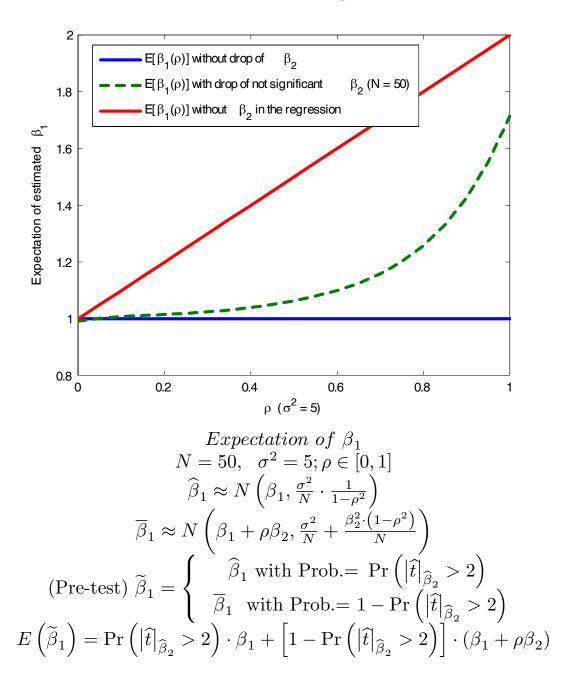
 $Distribution \ of \ \beta_1 \\ N = 50, \ \ \sigma^2 = 5; \rho \in [0,1] \\ 75$

Pre-test $\widetilde{\boldsymbol{\beta}}_1$ Estimator Density Function

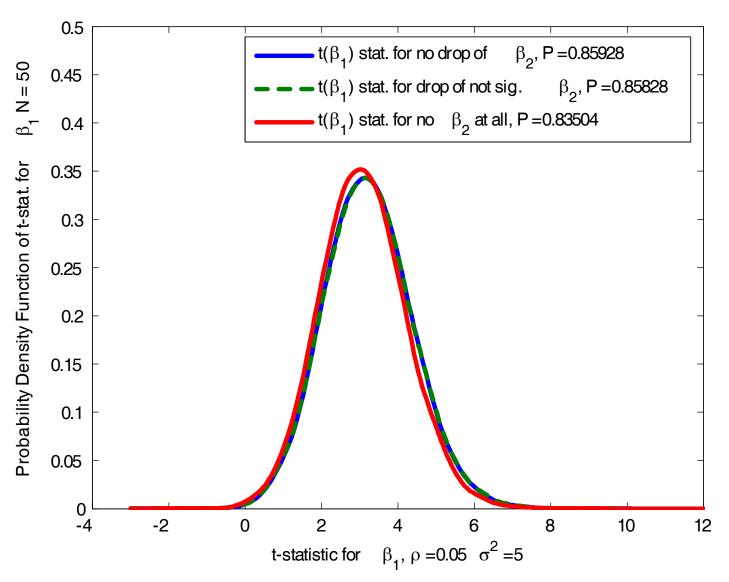


$$\begin{array}{c} Distribution \ of \ \beta_1 \\ N=50, \ \ \sigma^2=5; \rho \in [0,1] \end{array}$$

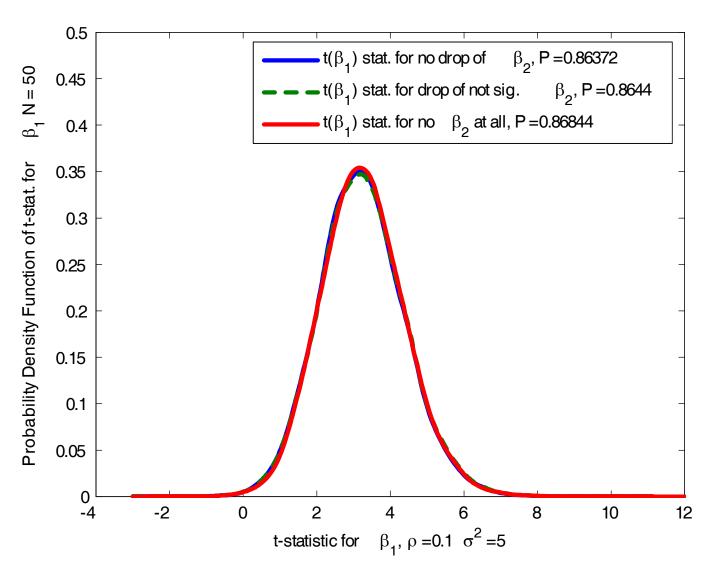
Pre-Test Estimator Expectation



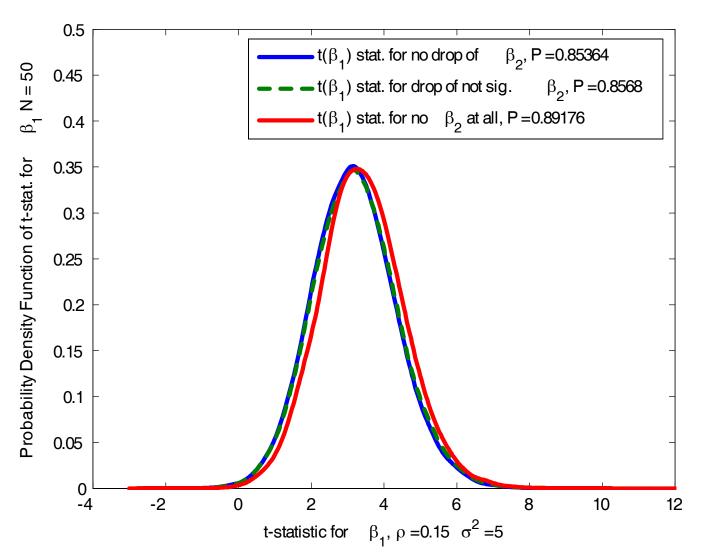
6 The β_1 T-statistic Analysis of Procedures 1, 2 and 3 for a specific Model



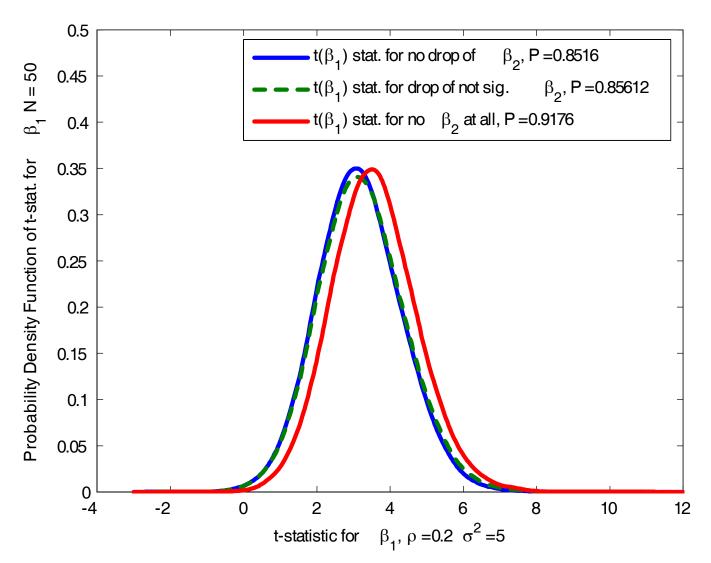
Distribution of T – statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N = 50, \quad \rho = 0.05; \quad \sigma^2 = 5$ 79



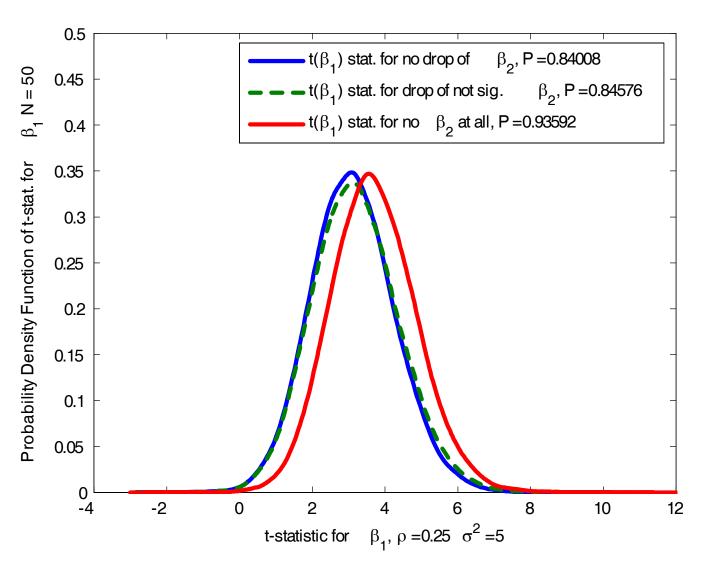
Distribution of T – statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N = 50, \ \rho = 0.10; \ \sigma^2 = 5$



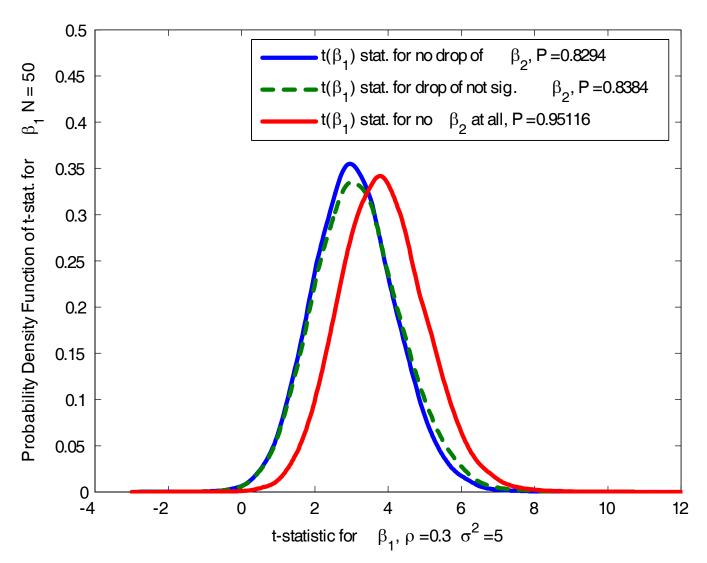
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.15; \ \sigma^2=5$



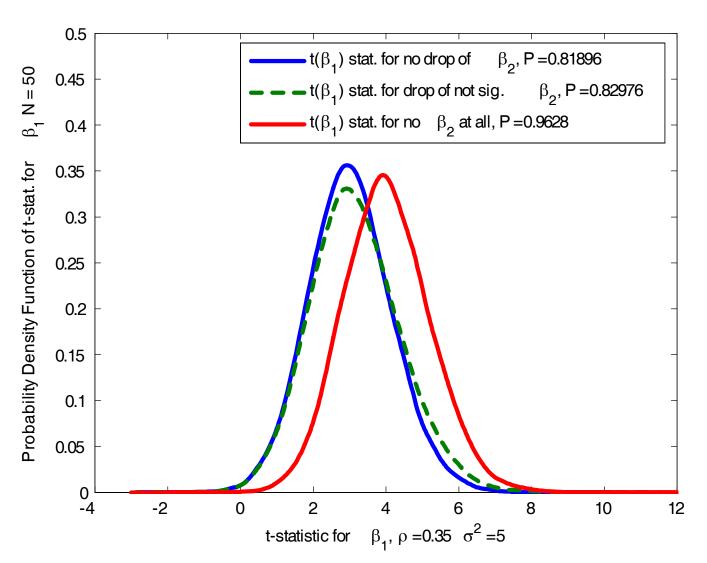
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.20; \ \sigma^2=5$



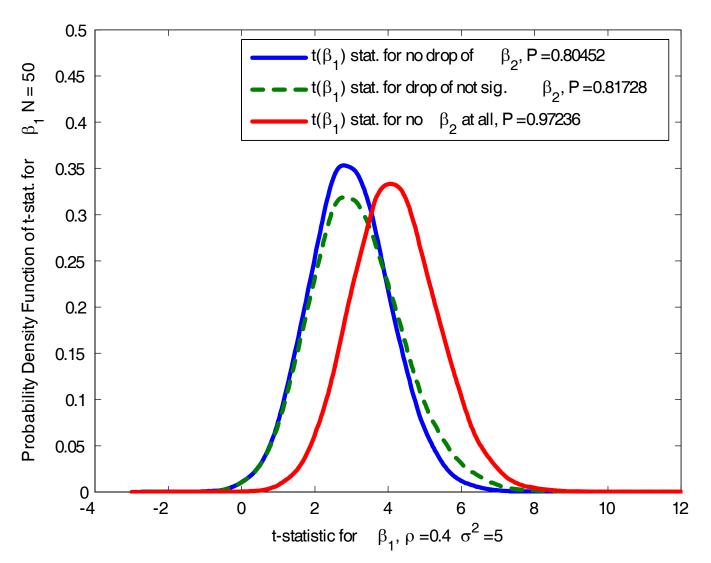
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.25; \ \sigma^2=5$



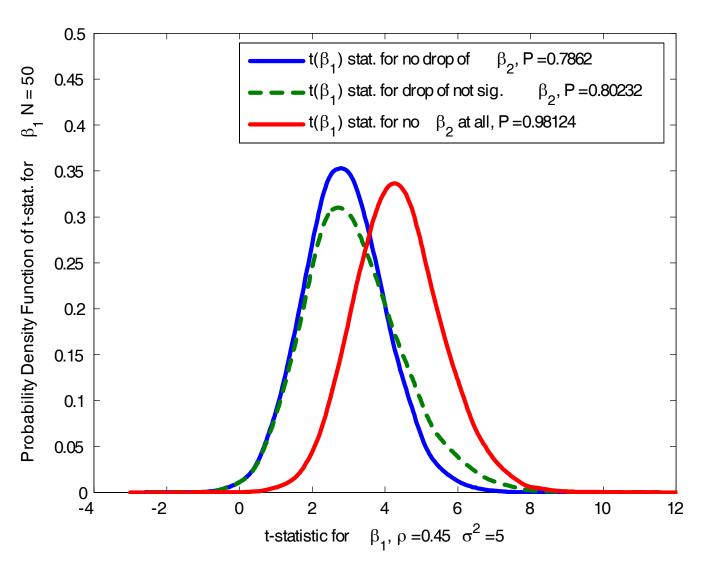
Distribution of T – statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N = 50, \ \rho = 0.30; \ \sigma^2 = 5$



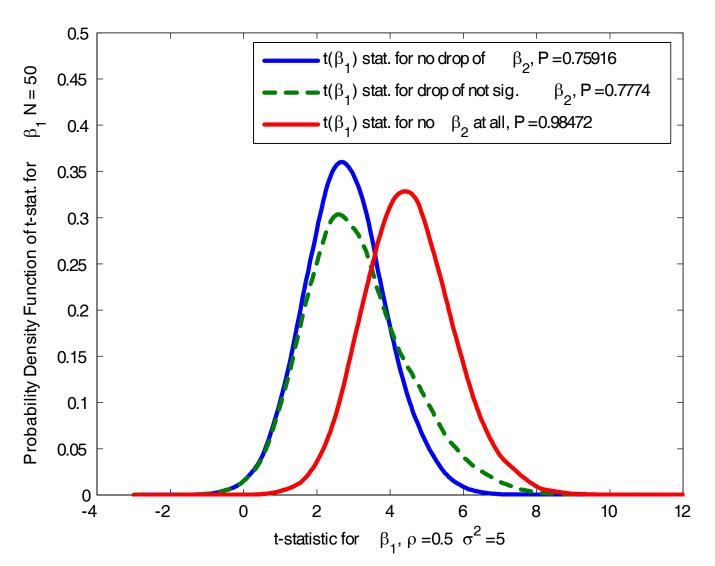
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.35; \ \sigma^2=5$



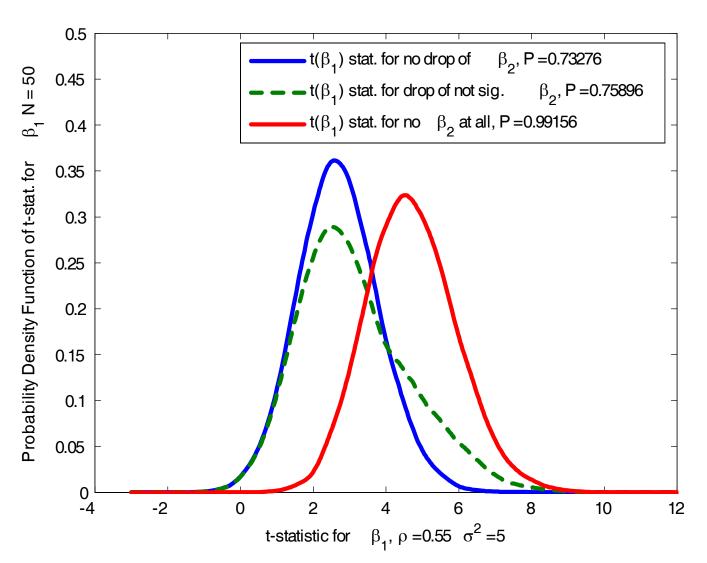
Distribution of T – statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N = 50, \ \rho = 0.40; \ \sigma^2 = 5$



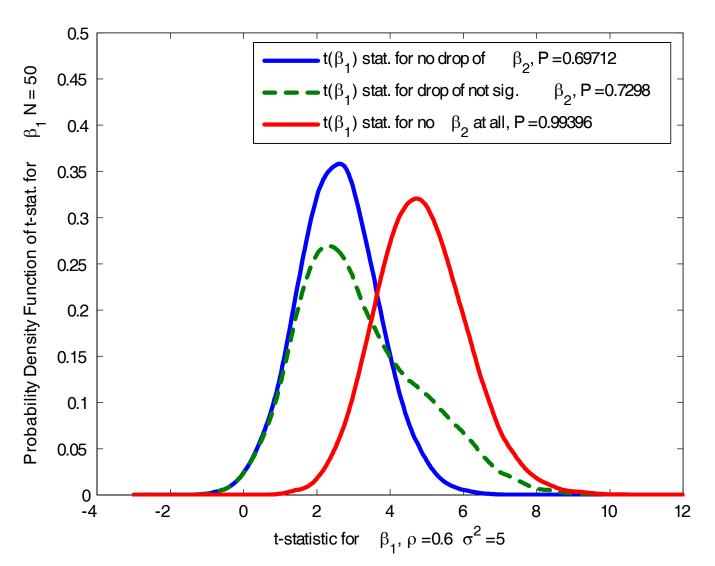
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.45; \ \sigma^2=5$



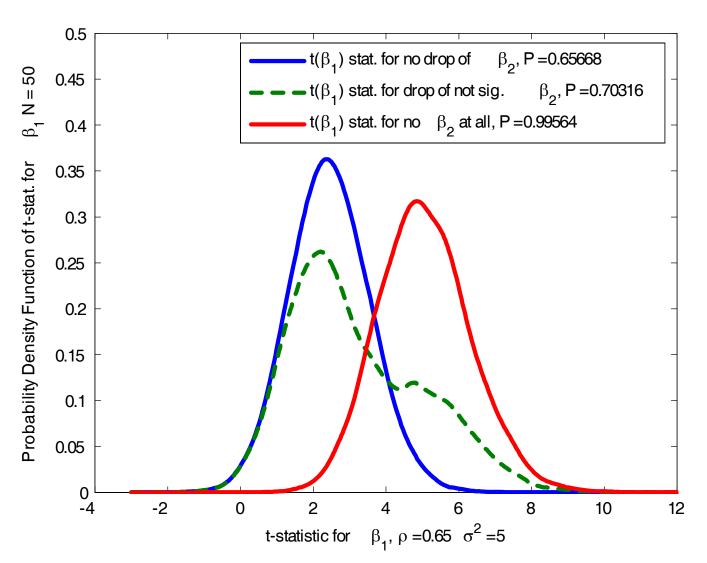
Distribution of T – statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N = 50, \ \rho = 0.50; \ \sigma^2 = 5$



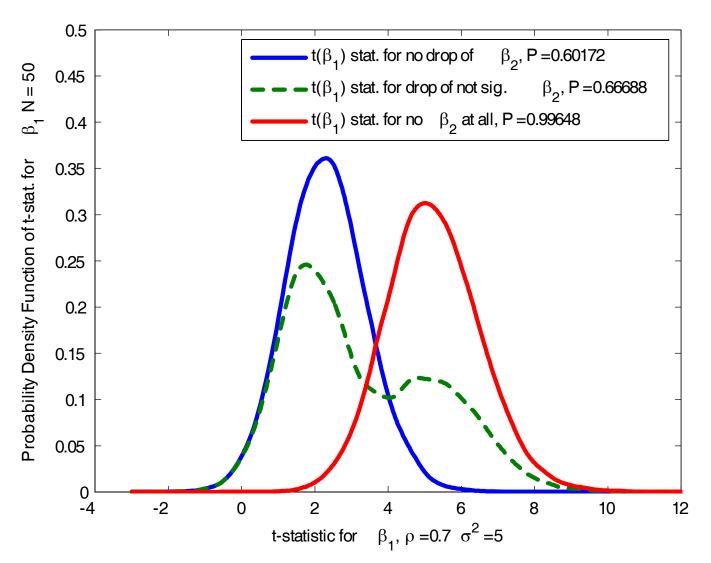
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.55; \ \sigma^2=5$



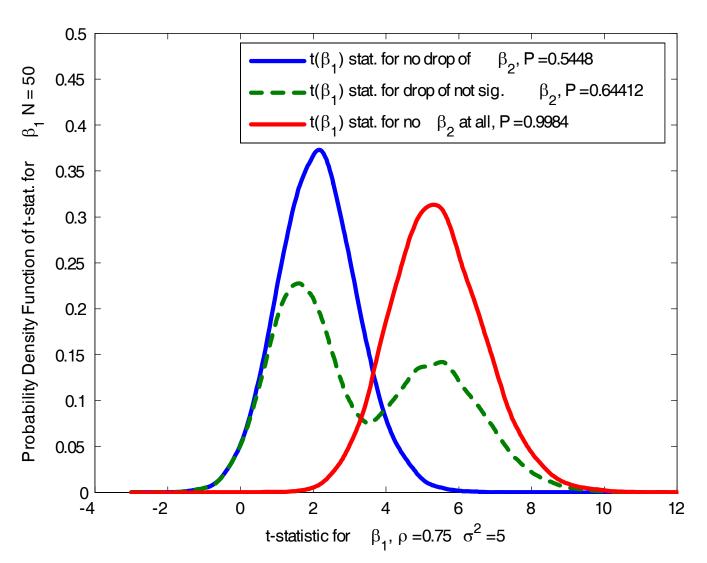
Distribution of T – statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N = 50, \ \rho = 0.60; \ \sigma^2 = 5$



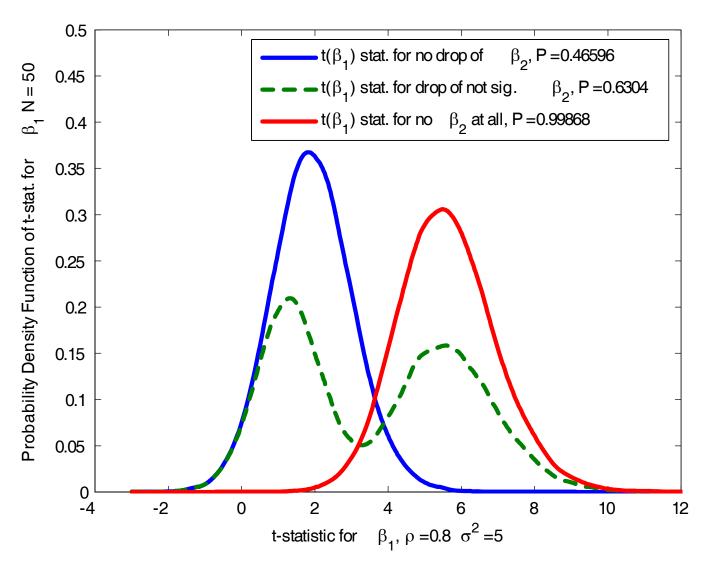
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.65; \ \sigma^2=5$



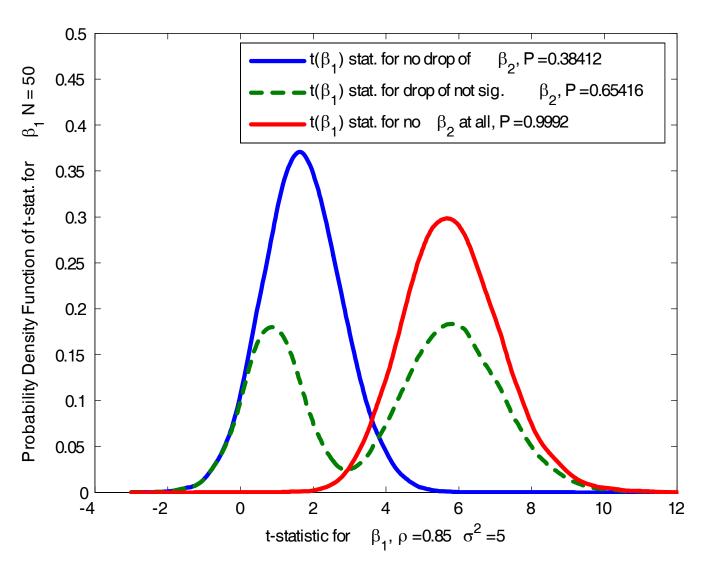
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.70; \ \sigma^2=5$



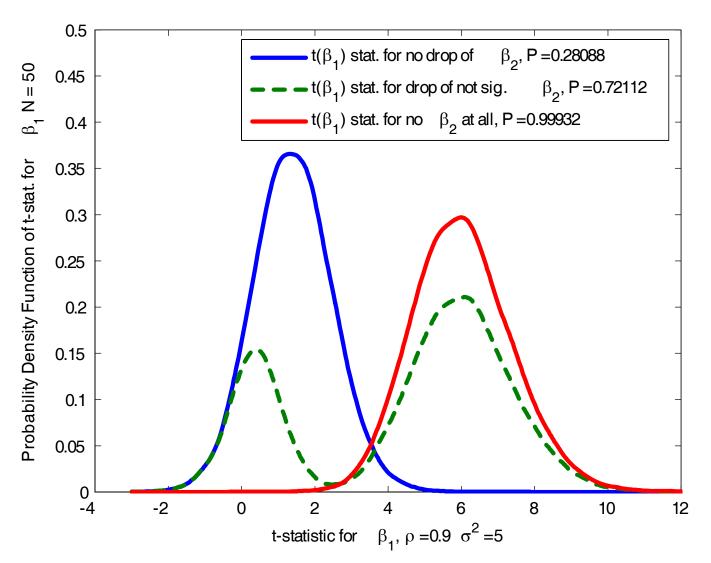
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.75; \ \sigma^2=5$



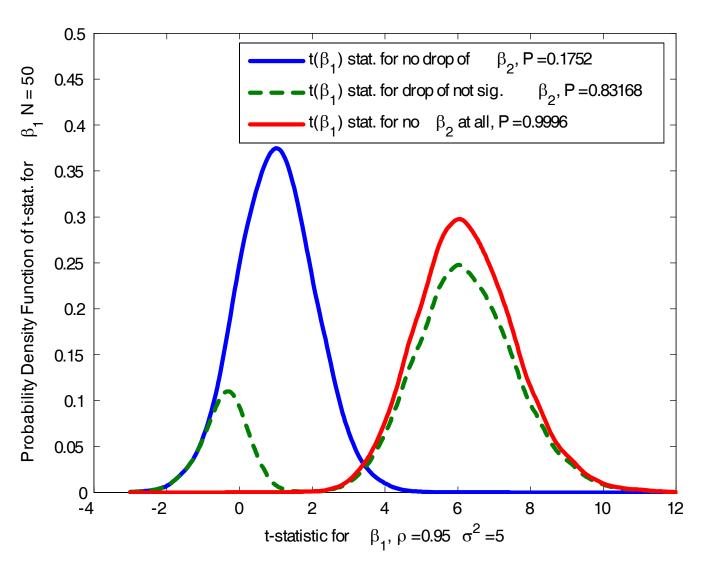
 $\begin{array}{c} Distribution \ of \ T-statistics \ \ for \ \beta_1 \\ P \ \text{is the test power} \equiv \Pr(|t|_{\widehat{\beta}_2} > 2) \\ N = 50, \ \ \rho = 0.80; \ \ \sigma^2 = 5 \end{array}$



Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.85; \ \sigma^2=5$

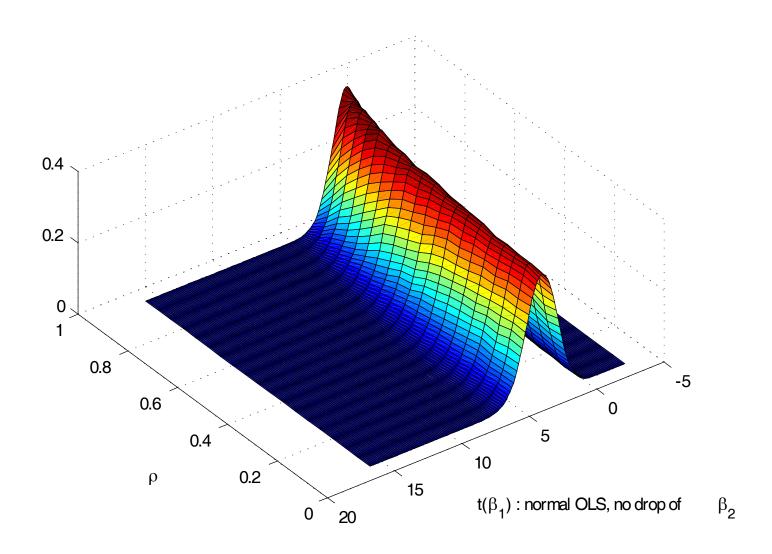


Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.90; \ \sigma^2=5$



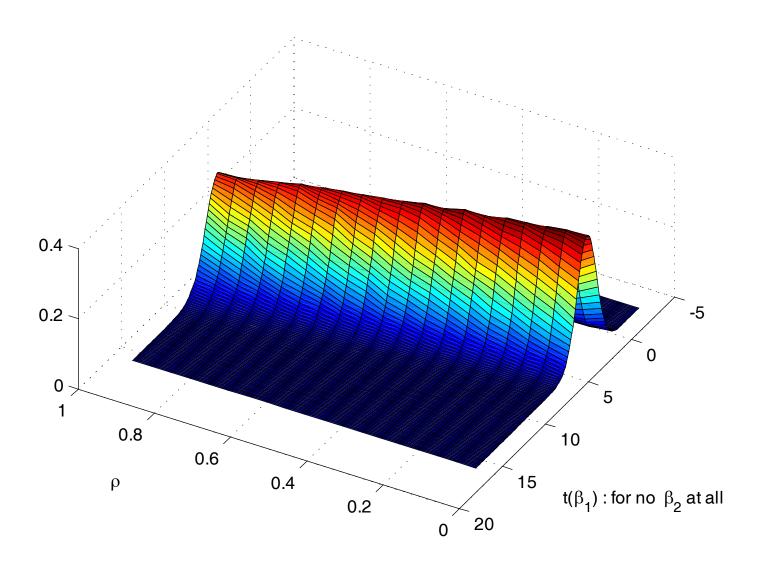
Distribution of T- statistics for β_1 P is the test power $\equiv \Pr(|t|_{\widehat{\beta}_2} > 2)$ $N=50, \ \rho=0.95; \ \sigma^2=5$

T-statistic Density Function for OLS $\widehat{\beta}_1$ Estimator



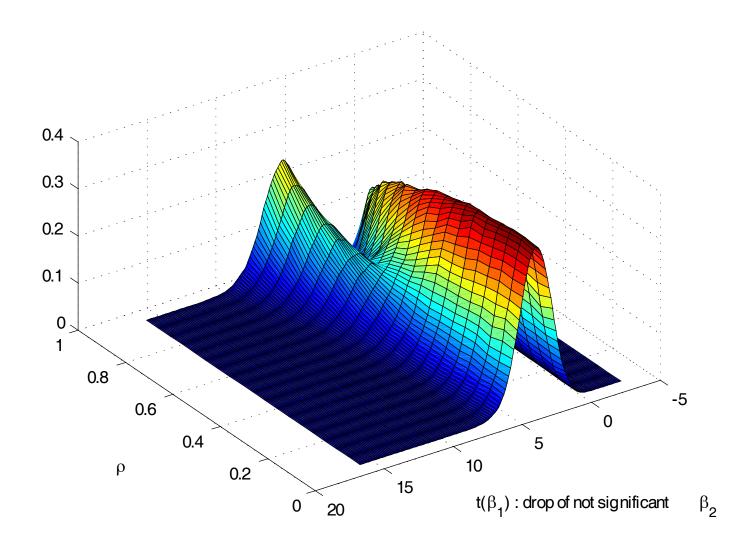
 $Distribution \ of \ T-statistics \ for \ \beta_1 \\ N=50, \ \ \sigma^2=5; \rho \in [0,1]$

T-statistic Density Function for Procedure 2 $\overline{\beta}_1$ Estimator



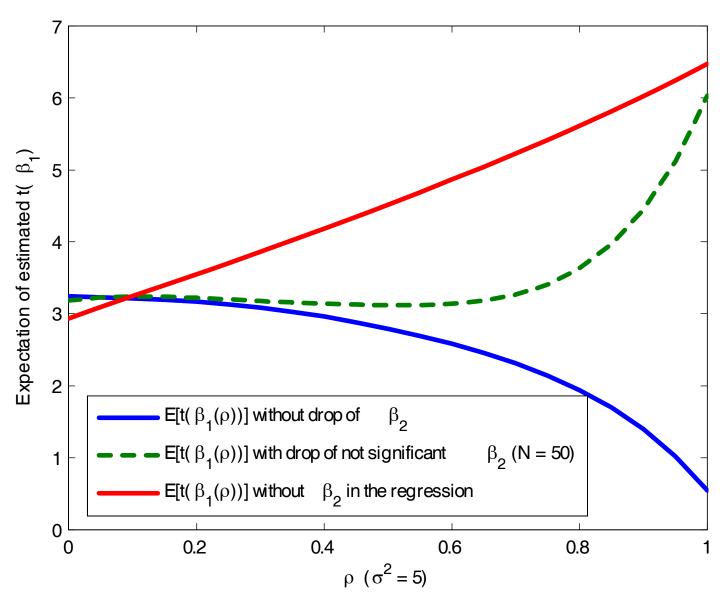
 $\begin{array}{c} Distribution \ of \ T-statistics \ \ for \ \beta_1 \\ N=50, \ \ \sigma^2=5; \rho \in [0,1] \end{array}$

T-statistic Density Function for Pre-test $\widetilde{\boldsymbol{\beta}}_1$ Estimator

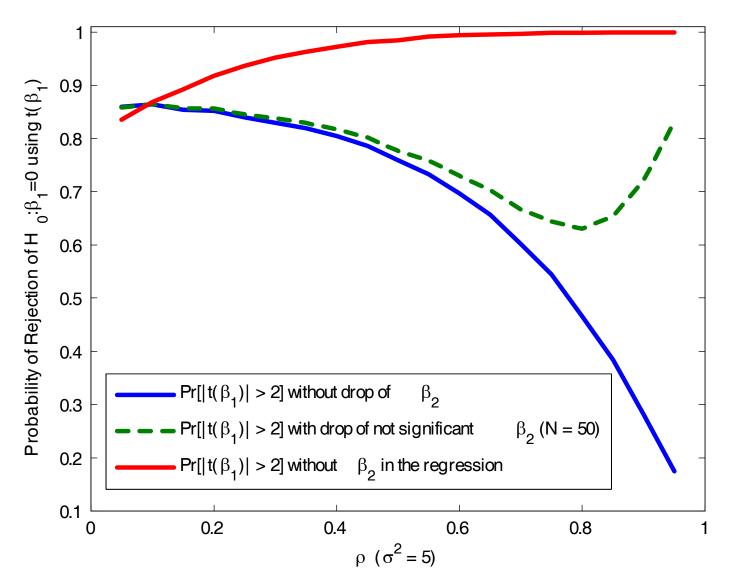


$$Distribution \ of \ T-statistics \ for \ \beta_1 \\ N=50, \ \ \sigma^2=5; \rho \in [0,1]$$

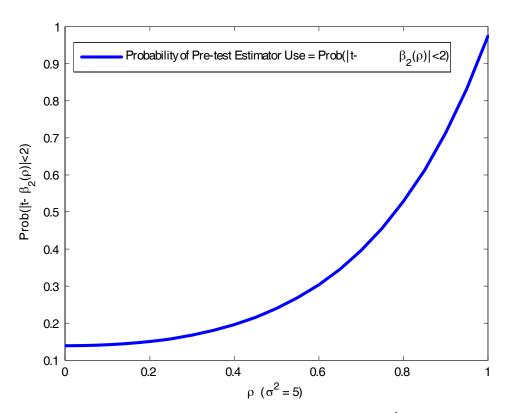
T-statistic Expectation for β_1



 $Expectation \ of \ T-statistics \ \ for \ \beta_1 \\ N=50, \ \ \sigma^2=5; \rho \in [0,1] \\ 101$

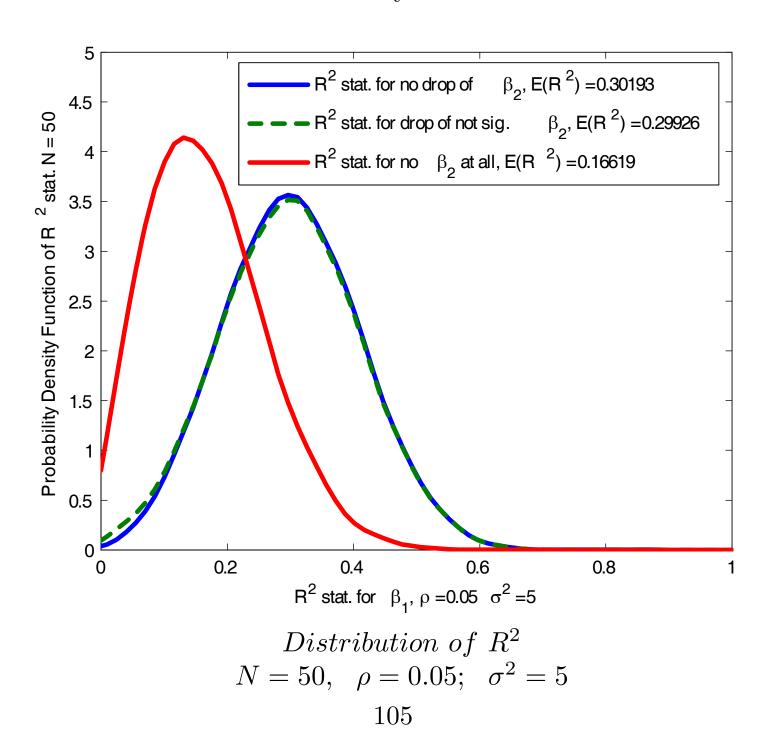


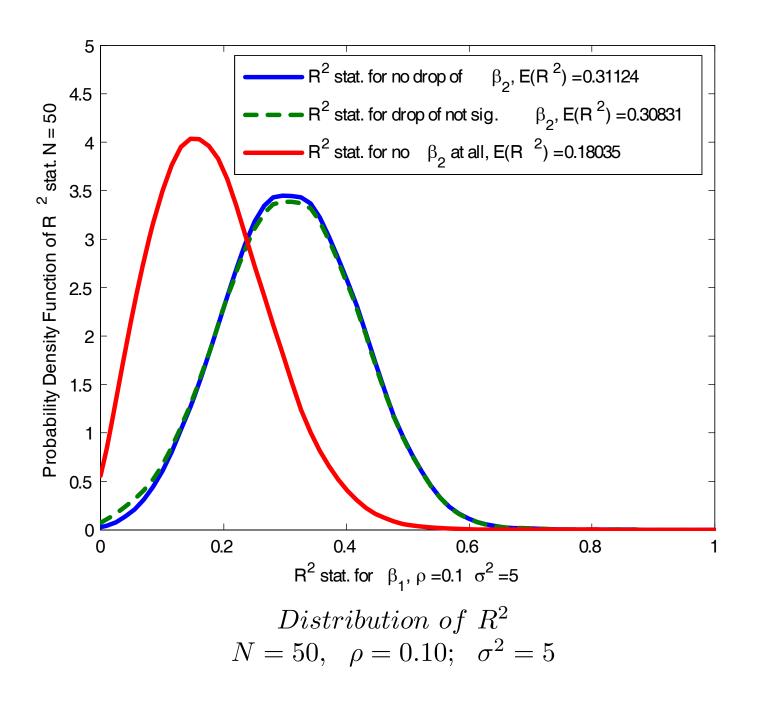
Probability of Rejection $H_0: \beta_1 = 0$. $\Pr(|t|_{\widehat{\beta}_1} > 2)$ $N = 50, \quad \sigma^2 = 5; \rho \in [0, 1]$

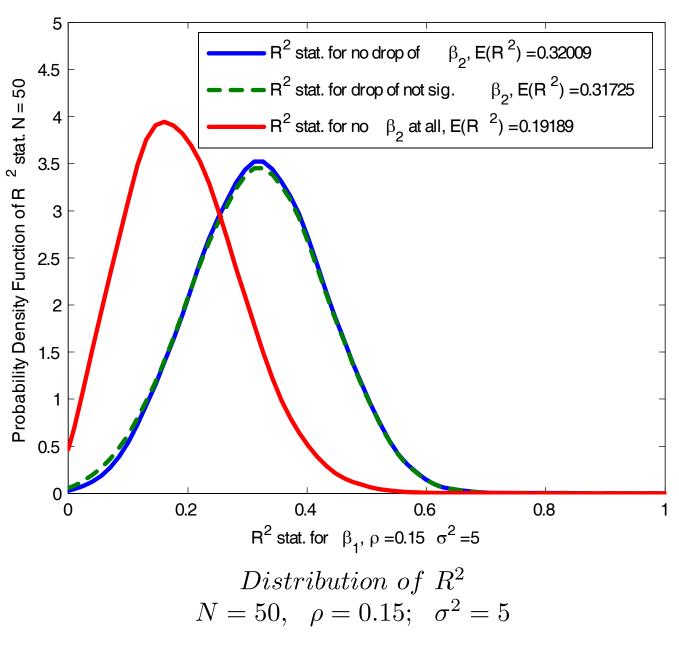


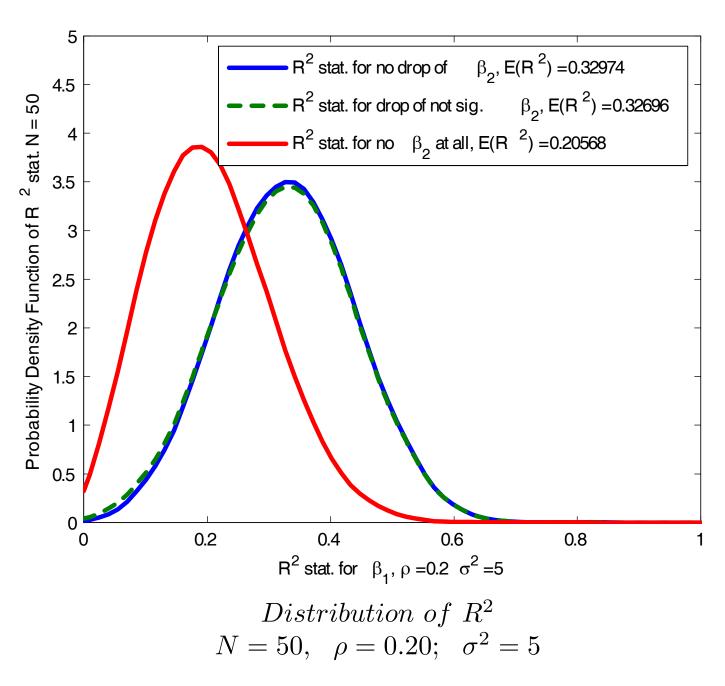
Probability of Non Rejection of $H_0: \widehat{\beta}_2 = 0$ $\Pr(|t|_{\widehat{\beta}_2} < 2)$ $N = 50, \quad \sigma^2 = 5; \rho \in [0, 1]$ $\widehat{\beta}_1 \approx N \left(\beta_1, \frac{\sigma^2}{N} \cdot \frac{1}{1 - \rho^2}\right)$ $\overline{\beta}_1 \approx N \left(\beta_1 + \rho \beta_2, \frac{\sigma^2}{N} + \frac{\beta_2^2 \cdot (1 - \rho^2)}{N}\right)$ $(\text{Pre-test}) \quad \widetilde{\beta}_1 = \begin{cases} \widehat{\beta}_1 \text{ with Prob.} = \Pr\left(|\widehat{t}|_{\widehat{\beta}_2} > 2\right) \\ \overline{\beta}_1 \text{ with Prob.} = \Pr\left(|\widehat{t}|_{\widehat{\beta}_2} < 2\right) \end{cases}$ $E\left(\widetilde{\beta}_1\right) = \Pr\left(|\widehat{t}|_{\widehat{\beta}_2} > 2\right) \cdot \beta_1 + \left[\Pr\left(|\widehat{t}|_{\widehat{\beta}_2} < 2\right)\right] \cdot (\beta_1 + \rho \beta_2)$

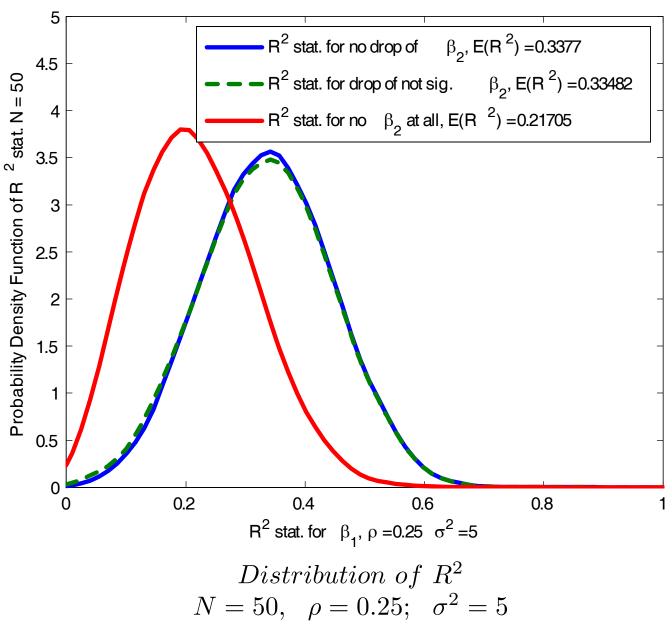
7 The R^2 Analysis

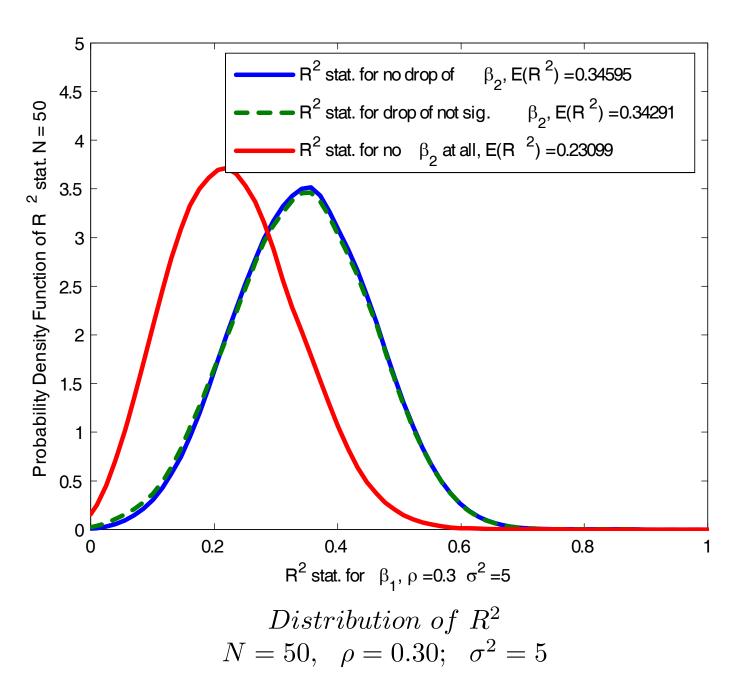


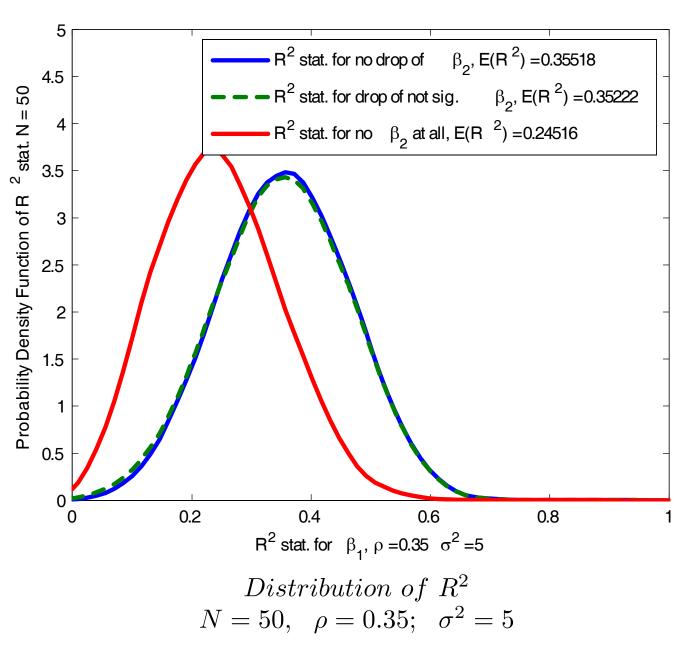


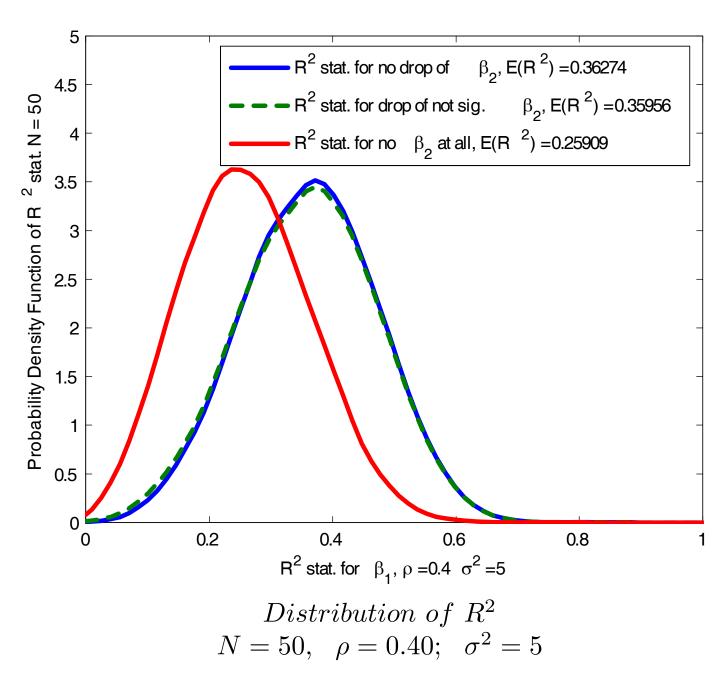


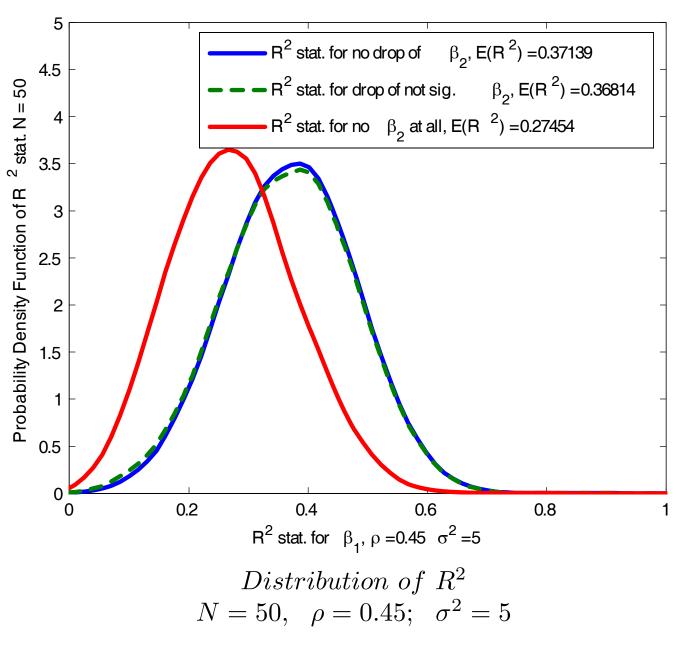


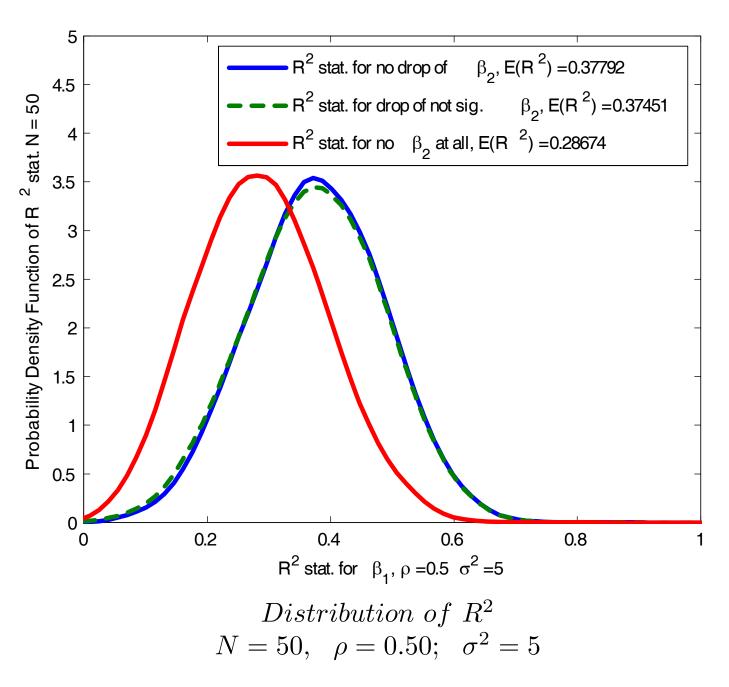


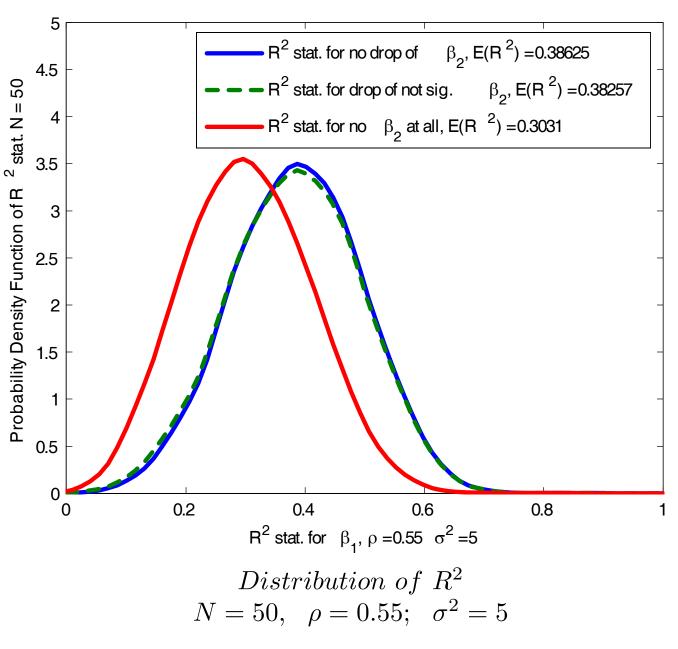


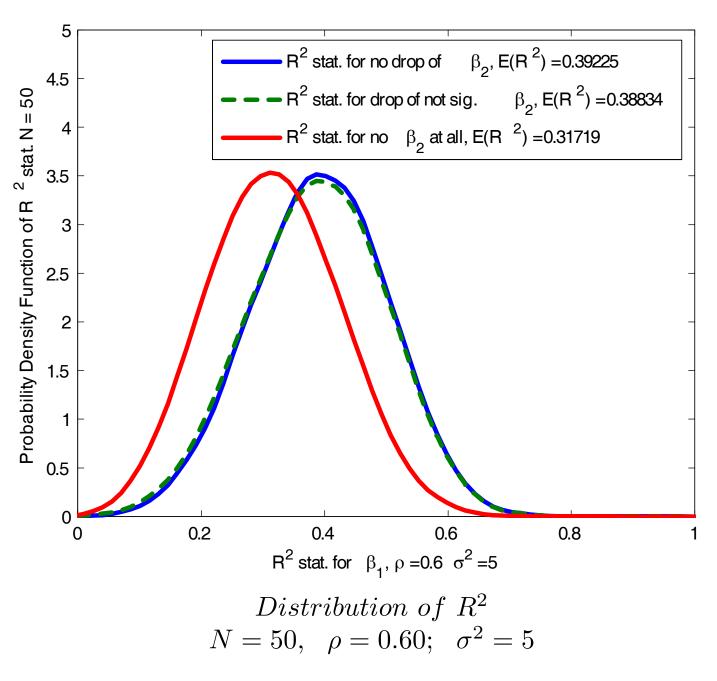


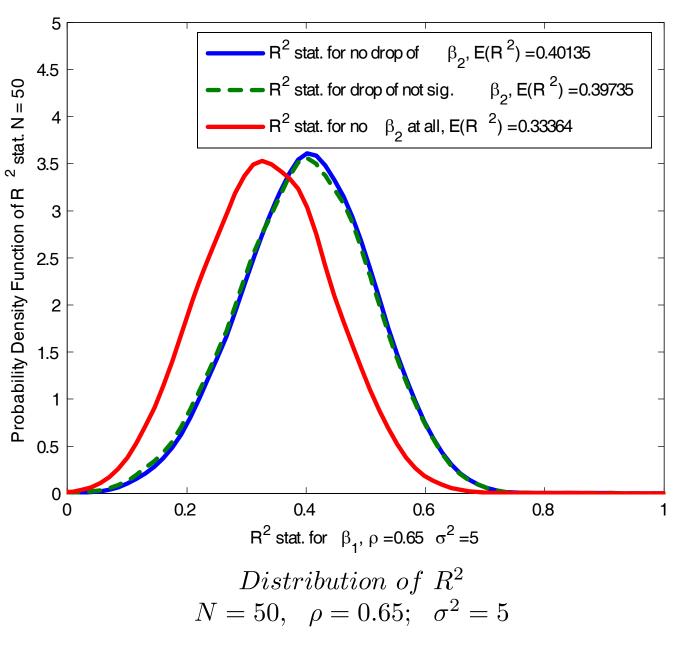


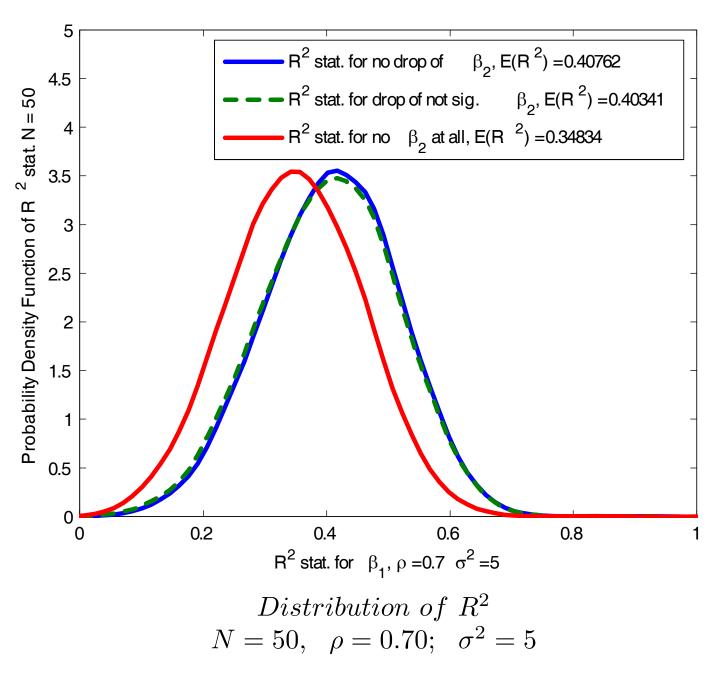


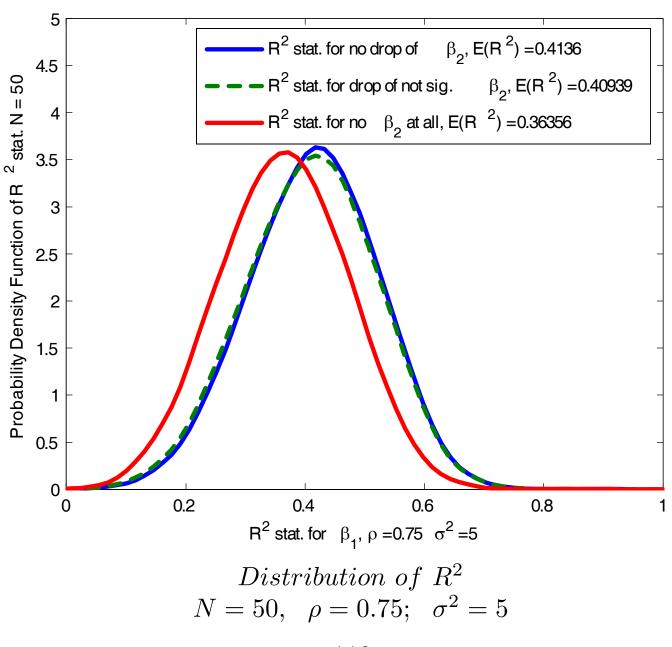


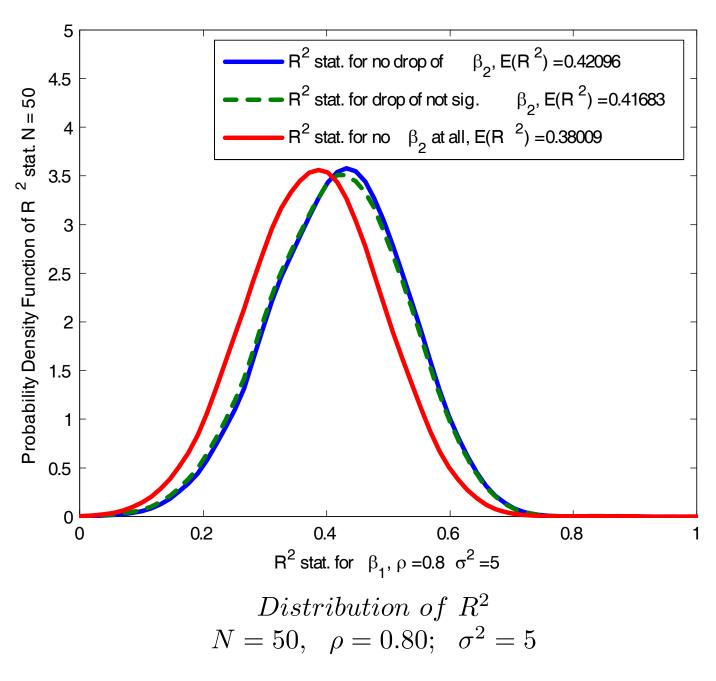


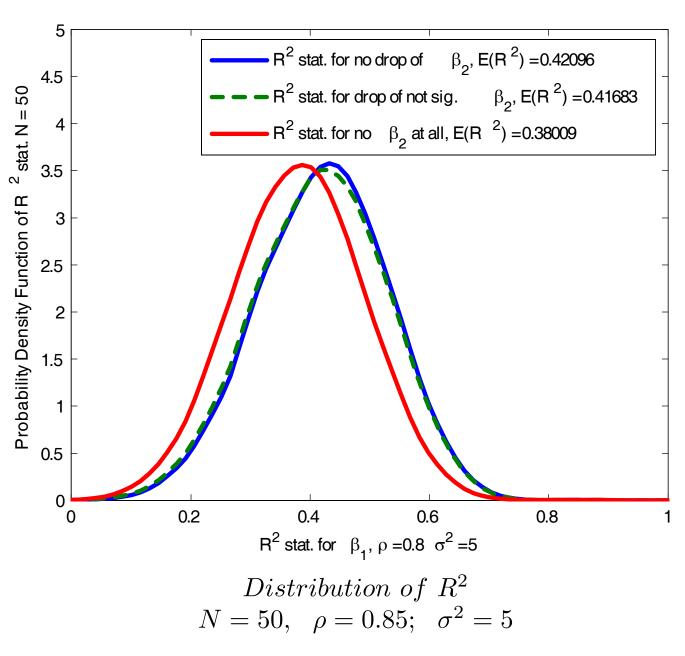


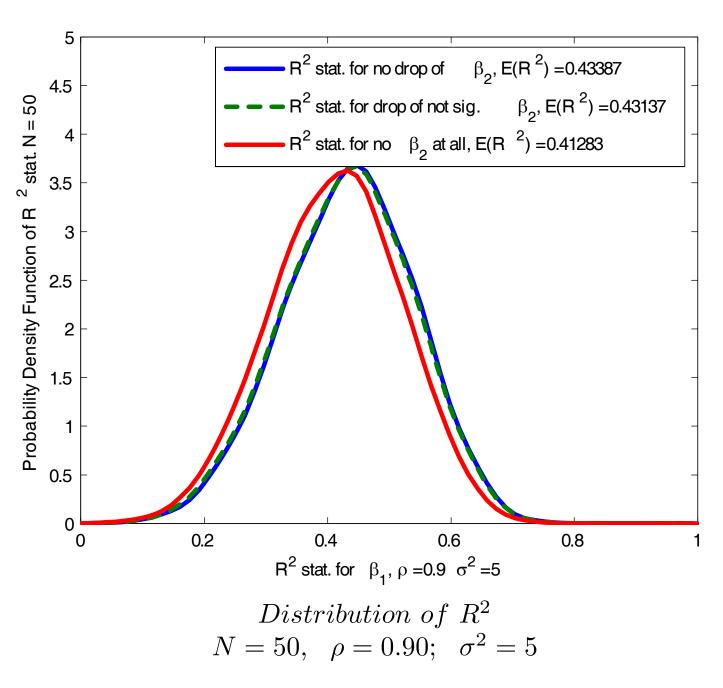


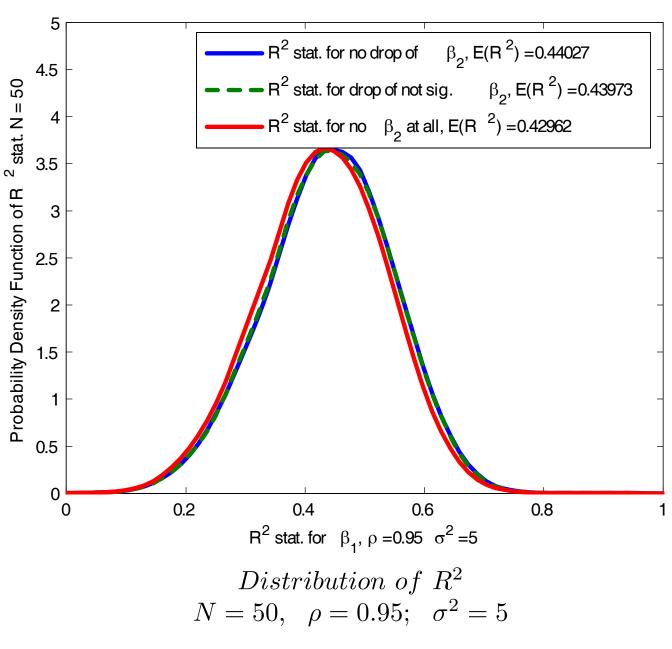




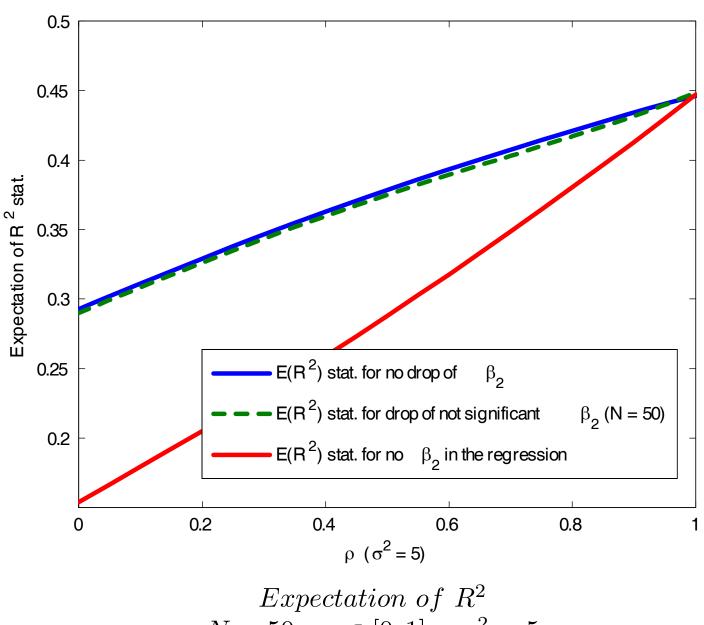








\mathbb{R}^2 Expectation



Expectation of R^2 $N = 50, \ \rho \in [0, 1]; \ \sigma^2 = 5$

Observe that for the case of $\rho = 1$, we obtain the model:

$$Y_i = X_i \begin{bmatrix} \beta_1 \\ \beta_2 \end{bmatrix} + \varepsilon_i; \ i = 1, ..., N; \quad \varepsilon_i \sim \mathcal{N}(0, \sigma^2)$$
 i.i.d.;

$$Y_{i} = X_{1,i}\beta_{1} + X_{2,i}\beta_{2} + \varepsilon_{i}$$

$$= X_{1,i}\beta_{1} + (X_{1,i}\xi)\beta_{2} + \varepsilon_{i}$$

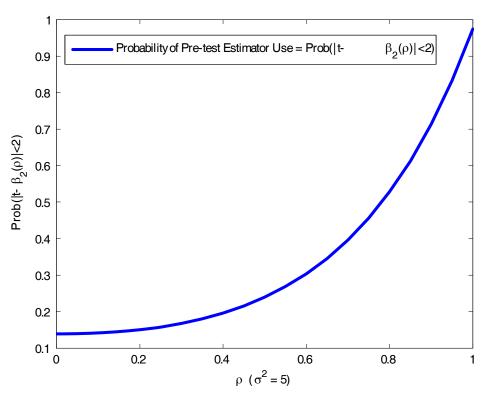
$$Y_{i} = X_{1,i}(\beta_{1} + \xi) + \varepsilon_{i}$$

No loss of information. R^2 remains the same, but the estimator is highly biased.

8 Conclusion

Suppose the variables have a reasonable correlation.

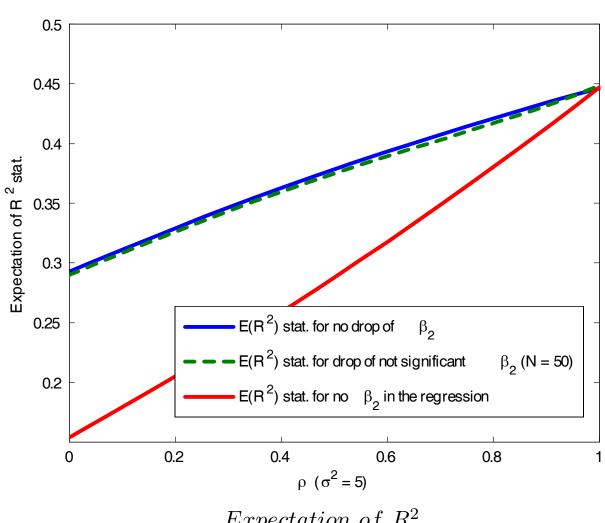
⇒ There is a high Probability to Procedure the second Estimation:



$$\begin{aligned} & \text{Probability of Non Rejection of } H_0: \widehat{\beta}_2 = 0 \\ & \text{Pr}(|t|_{\widehat{\beta}_2} < 2) \\ & N = 50, \ \sigma^2 = 5; \rho \in [0, 1] \\ & \widehat{\beta}_1 \approx N \left(\beta_1, \frac{\sigma^2}{N} \cdot \frac{1}{1 - \rho^2}\right) \\ & \overline{\beta}_1 \approx N \left(\beta_1 + \rho \beta_2, \frac{\sigma^2}{N} + \frac{\beta_2^2 \cdot (1 - \rho^2)}{N}\right) \\ & (\text{Pre-test}) \ \widetilde{\beta}_1 = \left\{ \begin{array}{c} \widehat{\beta}_1 \ \text{with Prob.} = \ \Pr\left(|\widehat{t}|_{\widehat{\beta}_2} > 2\right) \\ & \overline{\beta}_1 \ \text{with Prob.} = \ \Pr\left(|\widehat{t}|_{\widehat{\beta}_2} < 2\right) \end{array} \right. \\ & E\left(\widetilde{\beta}_1\right) = \Pr\left(|\widehat{t}|_{\widehat{\beta}_2} > 2\right) \cdot \beta_1 + \left[\Pr\left(|\widehat{t}|_{\widehat{\beta}_2} < 2\right)\right] \cdot (\beta_1 + \rho \beta_2) \end{aligned}$$

R² won't change much:

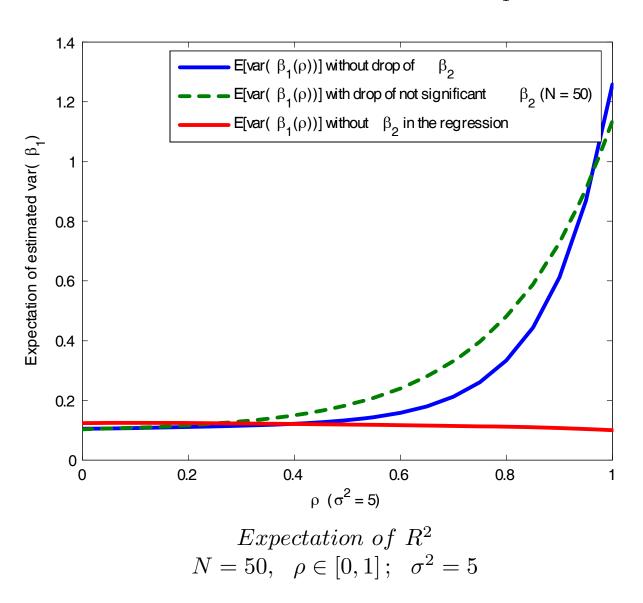
 \mathbb{R}^2 Expectation



 $Expectation of R^2$ $N = 50, \ \rho \in [0, 1]; \ \sigma^2 = 5$

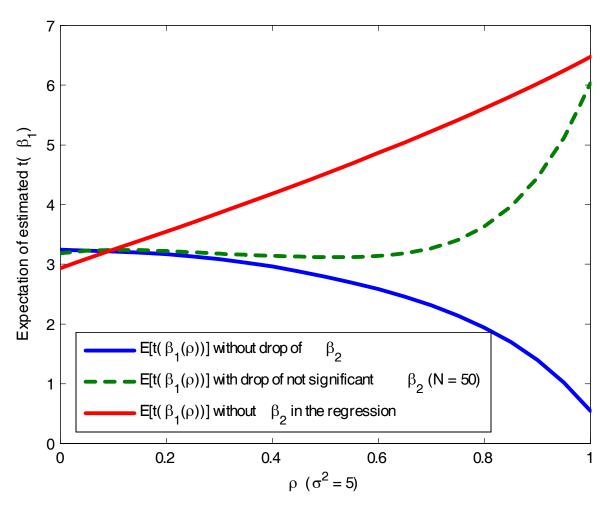
The Variance Might Decrease:

Expectation of The Variance of β_1



The Estimator $\overline{\beta}_1$ is very significant:

T-statistic Expectation for β_1



Expectation of T – statistics for β_1 $N = 50, \ \sigma^2 = 5; \rho \in [0, 1]$

And Extreme Biased:

Pre-Test Estimator Expectation

