



## Title

**tot\_tut** — Implementation for the estimation of treatment on the treated (ToT), treatment on the untreated (TuT) and the average treatment effect (ATE) jointly using the design introduced in "The controlled choice design and private paternalism in pawnshop borrowing". Other causal statistics of interest like the average selection bias and the average selection on the level are also provided.

## Syntax

```
tot_tut depvar treatvar choicevar [if] [in] [, vce(robust | cluster clustvar)
pvals b_rep1(#) b_rep2(#) ]
```

## Description

**tot\_tut** estimates jointly the treatment on the treated (ToT), treatment on the untreated (TuT), the average treatment effect, selection on gains, selection bias, and selection on the level, exploiting a design with three arms: a control arm, a forced arm and a choice arm. The specification strategy involves estimating two iv regressions per each selection estimand, and jointly obtaining errors. Moreover, the command displays bounds for the ToT and TuT without imposing exclusion restriction, and where indicated, also p-values for the violation of the IV validity. Details on the implementation can be found in the paper "The controlled choice design and private paternalism in pawnshop borrowing" and its accompanying appendix.

Note : Jointly inference for selection on gains, selection bias, and selection on the > level is not provided.

## Arguments

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**Arguments**

*depvar*, this is the outcome of interest.

*treatvar*, categorical variable indicating treatment status: control arm (0), forced arm (1), choice arm (2).

*choicevar*, binary variable indicating choice.

## Options

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**Options**

**vce(robust | cluster *clustvar*)** specifies the type of standard error reported, which includes types that are robust to some kinds of misspecification (robust - the default), and that allow for intragroup correlation (cluster *clustvar*).

**pvals**, if specified presents p-values for the violation of the IV validity. For details see Huber & Mellace (2015).

**b\_rep1(#)** Indicates the number of bootstrap replications for the first bootstrap in the computation of the p-value for the IV validity. Default is **b\_rep1(50)**

**b\_rep2(#)** Indicates the number of bootstrap replications for the second bootstrap in the computation of the p-value for the IV validity. Default is **b\_rep1(50)**

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## Examples

**"The controlled choice design and private paternalism in pawnshop borrowing"**

**Setup**

```

use tot_tut_commitment.dta, clear
gen x0 = -(Z==2)*(choose==0)
gen x1 = (Z==2)*(choose==1)
gen z0_ = -(Z==0)
gen z0_ = (Z==0)
gen z1_ = (Z==1)
gen z2_ = (Z==2)

```

**ToT & ATE using ivregress**

```
ivregress 2sls apr z1_ (x1 = z2), vce(cluster clustvar)
```

**TuT & ATE using ivregress**

```
ivregress 2sls apr z0_ (x0 = z2), vce(cluster clustvar)
```

**Simultaneous inference for ToT & TuT - selection on gains**

```
tot_tut apr Z choose, vce(cluster clustvar)
```

**Simultaneous inference for ToT & TuT, and p-values for IV validity**

```
tot_tut apr Z choose, vce(cluster clustvar) pvals b_rep1(50) b_rep2(50)
```

**Stored results**

**tot\_tut** stores the following in **e()**:

**Scalars**

<b>e(N)</b>	number of observations.
<b>e(df_r)</b>	residual degrees of freedom.
<b>e(pval_tot)</b>	p-value for the violation of the exclusion restriction for the treated.
<b>e(pval_tut)</b>	p-value for the violation of the exclusion restriction for the un-treated.
<b>e(pval_tot_tut)</b>	p-value for the simultaneous violation of the exclusion restriction.

**Matrices**

<b>e(b)</b>	coefficient vector.
<b>e(V)</b>	variance-covariance matrix of the estimators.

**tot\_tut** stores the following in **r()**:

**Scalars**

<b>r(ub_tut)</b>	Upper bound for the TuT without exclusion.
<b>r(lb_tut)</b>	Lower bound for the TuT without exclusion.
<b>r(ub_tot)</b>	Upper bound for the ToT without exclusion.
<b>r(lb_tot)</b>	Lower bound for the ToT without exclusion.

**References**

DiTraglia, McIntosh, Meza, Seira, Sadka. "The controlled choice design and private paternalism in pawnshop borrowing". Working paper.

Huber, M., & Mellace, G. (2015). TESTING INSTRUMENT VALIDITY FOR LATE IDENTIFICATION BASED ON INEQUALITY MOMENT CONSTRAINTS. The Review of Economics and Statistics, 97(2), 398-411.

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