

## Title

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**lpr4ytz** — Estimate the local persuasion rate

## Syntax

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```
lpr4ytz depvar treatrvar instrvar [covariates] [if] [in] [,  
model(string) title(string)]
```

## Options

<i>option</i>	<i>Description</i>
<b>model</b> ( <i>string</i> )	Regression model when <i>covariates</i> are present
<b>title</b> ( <i>string</i> )	Title

## Description

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**lpr4ytz** estimates the local persuasion rate (LPR). *varlist* should include *depvar treatrvar instrvar covariates* in order. Here, *depvar* is binary outcomes (*y*), *treatrvar* is binary treatments (*t*), *instrvar* is binary instruments (*z*), and *covariates* (*x*) are optional.

There are two cases: (i) *covariates* are absent and (ii) *covariates* are present.

- Without *x*, the LPR is defined by

$$\mathbf{LPR} = \{\Pr(y=1|z=1)-\Pr(y=1|z=0)\}/\{\Pr[y=0,t=0|z=0]-\Pr[y=0,t=0|z=1]\}.$$

The estimate and its standard error are obtained by the following procedure:

1. The numerator of the LPR is estimated by regressing *y* on *z*.
2. The denominator is estimated by regressing  $(1-y)*(1-t)$  on *z*.
3. The LPR is obtained as the ratio.
4. The standard error is computed via STATA command **nlcom**.

- With *x*, the LPR is defined by

$$\mathbf{LPR} = E[\mathbf{LPR\_num}(x)]/E[\mathbf{LPR\_den}(x)]$$

where

$$\mathbf{LPR\_num}(x) = \Pr(y=1|z=1,x) - \Pr(y=1|z=0,x)$$

and

$$\mathbf{LPR\_den}(x) = \Pr[y=0,t=0|z=0,x] - \Pr[y=0,t=0|z=1,x].$$

The estimate is obtained by the following procedure.

If **model**("no\_interaction") is selected (default choice),

1. The numerator of the LPR is estimated by regressing  $y$  on  $z$  and  $x$ .
2. The denominator is estimated by regressing  $(1-y)*(1-t)$  on  $z$  and  $x$ .
3. The LPR is obtained as the ratio.
4. The standard error is computed via STATA command **nlcom**.

Note that in this case, **LPR**( $x$ ) does not depend on  $x$ , because of the linear regression model specification.

Alternatively, if **model**("interaction") is selected,

1.  $\Pr(y=1|z,x)$  is estimated by regressing  $y$  on  $x$  given  $z = 0,1$ .
2.  $\Pr[y=0,t=0|z,x]$  is estimated by regressing  $(1-y)*(1-t)$  on  $x$  given  $z = 0,1$ .
3.  $\Pr(t=1|z,x)$  is estimated by regressing  $t$  on  $x$  given  $z = 0,1$ .
4. For each  $x$  in the estimation sample, both **LPR\_num**( $x$ ) and **LPR\_den**( $x$ ) are evaluated.
5. Then, the sample analog of **LPR** is constructed.

When *covariates* are present, the standard error is missing because an analytic formula for the standard error is complex. Bootstrap inference is implemented when this package's command **persuasio** is called to conduct inference.

## **Options**

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**model**(*string*) specifies a regression model.

This option is only relevant when *x* is present. The default option is "no\_interaction" between *z* and *x*. When "interaction" is selected, full interactions between *z* and *x* are allowed.

**title**(*string*) specifies a title.

### **Remarks**

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It is recommended to use this package's command **persuasio** instead of calling **lpr4ytz** directly.

### **Examples**

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We first call the dataset included in the package.

```
. use GKB_persuasio, clear
```

The first example estimates the LPR without covariates.

```
. lpr4ytz voteddem_all readsome post
```

The second example adds a covariate.

```
. lpr4ytz voteddem_all readsome post MZwave2
```

The third example allows for interactions between *x* and *z*.

```
. lpr4ytz voteddem_all readsome post MZwave2, model("interaction")
```

### **Stored results**

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#### **Scalars**

**e(N)**: sample size

**e(lpr\_coef)**: estimate of the local persuasion rate

**e(lpr\_se)**: standard error of the estimate of the local persuasion rate

#### **Macros**

**e(outcome):** variable name of the binary outcome variable

**e(treatment):** variable name of the binary treatment variable

**e(instrument):** variable name of the binary instrumental variable

**e(covariates):** variable name(s) of the covariates if they exist

**e(model):** regression model specification ("no\_interaction" or "interaction")

**Functions:**

**e(sample):** 1 if the observations are used for estimation, and 0 otherwise.

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**References**

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Sung Jae Jun and Sokbae Lee (2022), Identifying the Effect of Persuasion, [arXiv:1812.02276](https://arxiv.org/abs/1812.02276) [econ.EM].

**Version**

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