

# Package: **factiv** (via r-universe)

November 4, 2024

**Title** Instrumental Variables Estimation for  $2^k$  Factorial Experiments

**Version** 0.1.0

**Description** Implements instrumental variable estimators for  $2^K$  factorial experiments with noncompliance.

**Depends** R ( $\geq 2.10$ )

**Imports** stats, generics, Formula

**Suggests** tibble, dplyr, rmarkdown, knitr

**VignetteBuilder** knitr

**License** GPL ( $\geq 2$ )

**URL** <https://github.com/mattblackwell/factiv>

**BugReports** <https://github.com/mattblackwell/factiv/issues>

**Encoding** UTF-8

**LazyData** true

**Roxygen** list(markdown = TRUE)

**RoxygenNote** 7.1.1

**Repository** <https://mattblackwell.r-universe.dev>

**RemoteUrl** <https://github.com/mattblackwell/factiv>

**RemoteRef** HEAD

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compliance\_profile      *Complier covariate profiles*

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### Description

Calculates averages of covariates by compliance group in a  $2^K$  factorial setting.

### Usage

```
compliance_profile(formula, data, subset)
```

### Arguments

formula	one-sided formula to indicate the treatment assignment, treatment uptake, and covariates. The right-hand side of the formula should have three components separated by the   symbol, with the first component containing the K binary treatment variables (treatment assignment), the second component containing the K binary instruments associated with each treatment variable (treatment uptake), and the third giving the covariates to be included in the profile. The order of the variables in the first two parts of the formula must match.
data	a data.frame on which to apply the formula.
subset	subset of the data to pass to estimation.

### Value

A list with two objects:

raw_table	a data.frame whose rows represent the covariates and whose columns represent the different compliance groups. Each entry is the estimated mean of the covariate for that compliance group.
std_table	a data.frame similarly structured to raw_table but with the standardized difference between the compliance group means and the overall means in place of the raw means.

### Author(s)

Matthew Blackwell

### References

Matthew Blackwell and Nicole Pashley (2021) "Noncompliance in Factorial Experiments." Working paper.

**Examples**

```
data(newhaven)

cov_prof <- compliance_profile(~ inperson + phone | inperson_rand
  + phone_rand | age + maj_party + turnout_96, data = newhaven)

cov_prof
```

iv\_factorial

*IV Estimation of 2^K Factorial Design***Description**

Estimates principal stratum-specific effects and interactions in a 2^K factorial experiment

**Usage**

```
iv_factorial(formula, data, subset, method = "lm", level = 0.95)
```

**Arguments**

formula	formula specification of the factorial design with noncompliance. The right-hand side of the formula should have two components separated by the   symbol, with the first component containing the K binary treatment variables and the second component containing the K binary instruments associated with each treatment variable. The order of the variables in the formula must match.
data	A data.frame on which to apply the formula.
subset	subset of the data to pass to estimation.
method	character indicating if the estimator should be "lm" using the least squares approach (default) or "cmd" to estimate via efficient minimum distance estimator.
level	the confidence level required.

**Details**

This function estimates treatment effects for 2^K factorial experiments in the face of noncompliance on all factors. A monotonicity assumption is assumed for both treatment-instrument pairs, along with treatment exclusion. See Blackwell (2017) for more details on those assumptions.

The procedure uses iterative generalized method of moments (GMM) to estimate both the proportions of each compliance class (also known as principal strata) and the average potential outcomes within those classes. It also provides estimates of several one-way, joint, and interactive treatment effects within these classes.

Under the above assumptions, the compliance classes are the product of the compliance classes for each treatment-instrument pair. For instance, "cc" is the class that would comply with both treatments, "ca" is the class that would comply with the first treatment and always take the second treatment, and "cn" is the class that would comply with the first treatment and never take the second treatment. Finally, note that treatment effects are only well-defined for compliance classes for which there is compliance on at least one treatment.

**Value**

A list of class `iv_factorial` that contains the following components:

<code>rho</code>	vector of estimated compliance class probabilities.
<code>psi</code>	vector of the estimated conditional mean of the outcome within the compliance classes.
<code>vcov</code>	estimated asymptotic variance matrix of the combined <code>rho</code> and <code>psi</code> parameters.
<code>pcafe_est</code>	vector of estimated main effects of each factor among perfect compliers.
<code>pcafe_se</code>	vector of estimated standard errors for the estimated effects in <code>tau</code> .
<code>pcafe_cis</code>	a matrix of confidence intervals for the PCAFE estimates.
<code>level</code>	the confidence level of the returned confidence intervals.

**Author(s)**

Matt Blackwell

**References**

Matthew Blackwell (2017) Instrumental Variable Methods for Conditional Effects and Causal Interaction in Voter Mobilization Experiments, *Journal of the American Statistical Association*, 112:518, 590-599, doi:[10.1080/01621459.2016.1246363](https://doi.org/10.1080/01621459.2016.1246363)

Matthew Blackwell and Nicole Pashley (2020) "Noncompliance in Factorial Experiments." Working paper.

**Examples**

```
data(newhaven)

out <- iv_factorial(turnout_98 ~ inperson + phone | inperson_rand
  + phone_rand, data = newhaven)

summary(out)
```

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`iv_finite_factorial` *Finite-Sample IV Estimation of 2^K Factorial Design*

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**Description**

Estimate main effect IV ratios for 2^K factorial experiments

**Usage**

```
iv_finite_factorial(formula, data, subset, level = 0.95)
```

**Arguments**

formula	formula specification of the factorial design with noncompliance. The right-hand side of the formula should have two components separated by the   symbol, with the first component containing the K binary treatment variables and the second component containing the K binary instruments associated with each treatment variable. The order of the variables in the formula must match.
data	a data.frame on which to apply the formula.
subset	subset of the data to pass to estimation.
level	the confidence level required.

**Details**

This function estimates the ratio of the effect of treatment assignment on the outcome to the effect of treatment assignment on treatment uptake in  $2^K$  factorial experiments. The approach uses finite sample asymptotic inference to generate confidence intervals.

**Value**

A list of class `iv_finite_factorial` that contains the following components:

tau	a vector of estimated effect ratios for each factor.
tau_cis	a matrix of confidence intervals for each effect ratio. This matrix has 4 columns because it is possible to have disjoint confidence intervals in this method.
tau_y	a vector of the estimated effects on the outcome.
v_tau_y	the estimated sample variances of the effects on the outcome.
tau_d	a vector of the estimated effects on treatment uptake.
v_tau_y	the estimated sample variances of the effects on treatment uptake.
level	the confidence level of tau_cis.

**Author(s)**

Matt Blackwell

**References**

Matthew Blackwell and Nicole Pashley (2021) "Noncompliance in Factorial Experiments." Working paper.

**Examples**

```
data(newhaven)

out <- iv_finite_factorial(turnout_98 ~ inperson + phone | inperson_rand
  + phone_rand, data = newhaven)

out
```

```
joint <- iv_finite_factorial(turnout_98 ~ inperson + phone |
  inperson_rand + phone_rand, data = newhaven)

joint
```

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newhaven

*New Haven Voter Modelization Field Experiment*

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### Description

A data set that contains a subset of the voter mobilization field experiment analyzed in Gerber and Green (2000).

### Format

A data frame with 7865 observations and 6 variables:

**ward** ward of the residence.

**turnout\_98** indicator variable for voting in the 1998 general election.

**inperson** indicator variable for if the respondent received in-person canvassing.

**phone** indicator variable for if the respondent received phone canvassing.

**inperson\_rand** indicator variable for if the respondent was randomized to receive in-person canvassing.

**phone\_rand** indicator variable for if the respondent was randomized to receive phone canvassing.

**age** respondent age.

**maj\_party** indicator variable for if the respondent is a registered Democrat or Republican.

**turnout\_96** indicator variable for whether or not the respondent voted in the 1996 general election.

### Details

Data was cleaned and used in Bowers and Hansen (2009). It was a 2x2 factorial design with non-compliance on both factors. This is the subset of the subjects who lived in a single-person household and were not randomized to receive get-out-the-vote mailers. Blackwell (2017) analyzed the full data adjusting for noncompliance.

### Source

<https://doi.org/10.7910/DVN/Q6CID7>

## References

Gerber, Alan S., and Donald P. Green. “The Effects of Canvassing, Telephone Calls, and Direct Mail on Voter Turnout: A Field Experiment.” *The American Political Science Review* 94, no. 3 (2000): 653. <https://doi.org/10.2307/2585837>.

Hansen, Ben B., and Jake Bowers. “Attributing Effects to a Cluster-Randomized Get-Out-the-Vote Campaign.” *Journal of the American Statistical Association* 104, no. 487 (2009): 873–85. <https://doi.org/10.1198/jasa.2009.ap06589>.

Blackwell, Matthew. “Instrumental Variable Methods for Conditional Effects and Causal Interaction in Voter Mobilization Experiments.” *Journal of the American Statistical Association* 112, no. 518 (2017): 590–99. <https://doi.org/10.1080/01621459.2016.1246363>.

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tidy.iv\_factorial      *Tidy an iv\_factorial object*

---

## Description

Tidy summarizes information about the components of a model.

## Usage

```
## S3 method for class 'iv_factorial'
tidy(x, conf.int = FALSE, conf.level = 0.95, ...)
```

## Arguments

x	An <code>iv_factorial</code> object produced by a call to <code>iv_factorial()</code>
conf.int	Logical indicating whether or not to include a confidence interval in the tidied output. Defaults to FALSE
conf.level	The confidence level to use for the confidence interval if <code>conf.int = TRUE</code> . Must be strictly greater than 0 and less than 1. Defaults to 0.95, which corresponds to a 95 percent confidence interval.
...	Additional arguments. Not used. Needed to match generic signature only.

## Value

A `tibble::tibble()` with columns:

term	The name of the effect term.
estimand	Which complier effect being estimated.
estimate	The estimated value of the effect.
std.error	The estimated standard error of the effect.
conf.low	Lower bound of the confidence interval for the estimate.
conf.high	Upper bound of the confidence interval for the estimate.

## Author(s)

Matt Blackwell

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 tidy.iv\_finite\_factorial

*Tidy an iv\_finite\_factorial object*


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## Description

Tidy summarizes information about the components of a model.

## Usage

```
## S3 method for class 'iv_finite_factorial'
tidy(x, conf.level = 0.95, ...)
```

## Arguments

x	An iv_factorial object produced by a call to <code>iv_finite_factorial()</code>
conf.level	The confidence level to use for the confidence interval. Must be strictly greater than 0 and less than 1. Defaults to 0.95, which corresponds to a 95 percent confidence interval.
...	Additional arguments. Not used. Needed to match generic signature only.

## Value

A `tibble::tibble()` with columns:

term	The name of the effect term.
estimand	Which complier effect being estimated.
estimate	The estimated value of the effect.
ci_1_lower	Lower bound for the first interval of the Fieller confidence regression for the estimate.
ci_1_upper	Upper bound for the first interval of the Fieller confidence regression for the estimate.
ci_2_lower	Lower bound for the second interval of the Fieller confidence regression for the estimate. Only non-NA when the confidence region is disjoint.
ci_2_upper	Upper bound for the second interval of the Fieller confidence regression for the estimate. Only non-NA when the confidence region is disjoint.

## Author(s)

Matt Blackwell



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