

Package: RcausalEGM (via r-universe)

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Type Package

Title A General Causal Inference Framework by Encoding Generative Modeling

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Description CausalEGM is a general causal inference framework for estimating causal effects by encoding generative modeling, which can be applied in both discrete and continuous treatment settings. A description of the methods is given in Liu (2022) <[doi:10.48550/arXiv.2212.05925](https://doi.org/10.48550/arXiv.2212.05925)>.

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Encoding UTF-8

Imports reticulate

URL <https://github.com/SUwonglab/CausalEGM>

BugReports <https://github.com/SUwonglab/CausalEGM/issues>

Depends R(>= 3.6.0)

RoxygenNote 7.2.3

Suggests rmarkdown, knitr, testthat (>= 3.0.0)

VignetteBuilder knitr

Config/testthat/edition 3

NeedsCompilation no

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causalegm	<i>Main function for estimating causal effect in either binary or continuous treatment settings.</i>
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Description

This function takes observation data (x,y,v) as input, and estimate the ATE/ITE/ADRF.

Usage

```
causalegm(
  x,
  y,
  v,
  z_dims = c(3, 3, 6, 6),
  output_dir = ".",
  dataset = "myData",
  lr = 2e-04,
  bs = 32,
  alpha = 1,
  beta = 1,
  gamma = 10,
  g_d_freq = 5,
  g_units = c(64, 64, 64, 64, 64),
  e_units = c(64, 64, 64, 64, 64),
  f_units = c(64, 32, 8),
  h_units = c(64, 32, 8),
  dv_units = c(64, 32, 8),
  dz_units = c(64, 32, 8),
  save_model = FALSE,
  save_res = FALSE,
  binary_treatment = TRUE,
  use_z_rec = TRUE,
  use_v_gan = TRUE,
  random_seed = 123,
  n_iter = 30000,
  normalize = FALSE,
  x_min = NULL,
  x_max = NULL
)
```

Arguments

<code>x</code>	is the treatment variable, one-dimensional array with size n .
<code>y</code>	is the potential outcome, one-dimensional array with size n .
<code>v</code>	is the covariates, two-dimensional array with size n by p .
<code>z_dims</code>	is the latent dimensions for z_0, z_1, z_2, z_3 respectively. Total dimension should be much smaller than the dimension of covariates v . Default: <code>c(3,3,6,6)</code>
<code>output_dir</code>	is the folder to save the results including model hyperparameters and the estimated causal effect. Default is <code>"."</code> .
<code>dataset</code>	is the name for the input data. Default: <code>"myData"</code> .
<code>lr</code>	is the learning rate. Default: <code>0.0002</code> .
<code>bs</code>	is the batch size. Default: <code>32</code> .
<code>alpha</code>	is the coefficient for the reconstruction loss. Default: <code>1</code> .
<code>beta</code>	is the coefficient for the MSE loss of x and y . Default: <code>1</code> .
<code>gamma</code>	is the coefficient for the gradient penalty loss. Default: <code>10</code> .
<code>g_d_freq</code>	is the iteration frequency between training generator and discriminator in the Roundtrip framework. Default: <code>5</code> .
<code>g_units</code>	is the list of hidden nodes in the generator/decoder network. Default: <code>c(64,64,64,64,64)</code> .
<code>e_units</code>	is the list of hidden nodes in the encoder network. Default: <code>c(64,64,64,64,64)</code> .
<code>f_units</code>	is the list of hidden nodes in the f network for predicting y . Default: <code>c(64,32,8)</code> .
<code>h_units</code>	is the list of hidden nodes in the h network for predicting x . Default: <code>c(64,32,8)</code> .
<code>dv_units</code>	is the list of hidden nodes in the discriminator for distribution match v . Default: <code>c(64,32,8)</code> .
<code>dz_units</code>	is the list of hidden nodes in the discriminator for distribution match z . Default: <code>c(64,32,8)</code> .
<code>save_model</code>	whether to save the trained model. Default: <code>FALSE</code> .
<code>save_res</code>	whether to save the results during training. Default: <code>FALSE</code> .
<code>binary_treatment</code>	whether the treatment is binary or continuous. Default: <code>TRUE</code> .
<code>use_z_rec</code>	whether to use the reconstruction loss for z . Default: <code>TRUE</code> .
<code>use_v_gan</code>	whether to use the GAN training for v . Default: <code>TRUE</code> .
<code>random_seed</code>	is the random seed to fix randomness. Default: <code>123</code> .
<code>n_iter</code>	is the training iterations. Default: <code>30000</code> .
<code>normalize</code>	whether apply normalization to covariates. Default: <code>FALSE</code> .
<code>x_min</code>	ADRF start value. Default: <code>NULL</code>
<code>x_max</code>	ADRF end value. Default: <code>NULL</code>

Value

causalegm returns an object of `class` "causalegm".

An object of class "causalegm" is a list containing the following:

causal_pre	the predicted causal effects, which are individual causal effects (ITEs) in binary treatment settings and dose-response values in continuous treatment settings.
getCATE	the method for getting the conditional average treatment effect (CATE). It takes covariates <i>v</i> as input.
predict	the method for outcome function. It takes treatment <i>x</i> and covariates <i>v</i> as inputs.

References

Qiao Liu, Zhongren Chen, Wing Hung Wong. CausalEGM: a general causal inference framework by encoding generative modeling. *arXiv preprint arXiv:2212.05925*, 2022.

Examples

```
#Generate a simple simulation data.
n <- 1000
p <- 10
v <- matrix(rnorm(n * p), n, p)
x <- rbinom(n, 1, 0.4 + 0.2 * (v[, 1] > 0))
y <- pmax(v[, 1], 0) * x + v[, 2] + pmin(v[, 3], 0) + rnorm(n)
model <- causalegm(x=x, y=y, v=v, n_iter=3000)
paste("The average treatment effect (ATE):", round(model$ATE, 2))
```

get_est

Make predictions with causalEGM model.

Description

When *x* is NULL, the conditional average treatment effect (CATE), namely $\tau(v)$, is estimated using a trained causalEGM model. When *x* is provided, estimating the potential outcome *y* given treatment *x* and covariates *v* using a trained causalEGM model.

Usage

```
get_est(object, v, x = NULL)
```

Arguments

object	An object of class "causalegm".
v	is the covariates, two-dimensional array with size <i>n</i> by <i>p</i> .
x	is the optional treatment variable, one-dimensional array with size <i>n</i> . Defaults to NULL.

Value

Vector of predictions.

Examples

```
#Generate a simple simulation data.
n <- 1000
p <- 10
v <- matrix(rnorm(n * p), n, p)
x <- rbinom(n, 1, 0.4 + 0.2 * (v[, 1] > 0))
y <- pmax(v[, 1], 0) * x + v[, 2] + pmin(v[, 3], 0) + rnorm(n)
model <- causalegm(x=x, y=y, v=v, n_iter=3000)
n_test <- 100
v_test <- matrix(rnorm(n_test * p), n_test, p)
x_test <- rbinom(n_test, 1, 0.4 + 0.2 * (v_test[, 1] > 0))
pred_cate <- get_est(model, v = v_test) # CATE estimate
pred_y <- get_est(model, v = v_test, x = x_test) # y given treatment x plus covariates v
```

install_causalegm	<i>Install the python CausalEGM package</i>
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Description

Install the python CausalEGM package

Usage

```
install_causalegm(method = "auto", pip = TRUE)
```

Arguments

method	default "auto"
pip	boolean flag, default TRUE

Value

No return value

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