## Package 'ZIQSIR'

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Title ZIQ-SIR package	ZIQ-SIR package on 0.1.0 or Zirui Wang, Tianying Wang tainer Zirui Wang <wzr23@mails.tsinghua.edu.cn>, Tianying Wang <tianying.wang@colostate.edu> ription We provide a novel semiparametric single-index method to test the relationship between the co- variate(s) of interest and zero-inflated data. By inputting the response variable, explanatory vari- able(s), the indices of the variables to be tested, and the method to be used, we can test the rela- tionship between the variables of interest and the zero-inflated response variable.  see GPL-2 ding UTF-8 Data true nds R (&gt;= 3.5.0), orts MASS, lme4, splines2, PearsonDS, quantreg genNote 7.3.2  optics documented:  ZIQSIR</tianying.wang@colostate.edu></wzr23@mails.tsinghua.edu.cn>	
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Depends R (>= 3.5.0),		
	splines2, PearsonDS,	
RoxygenNote 7.3.2		
R topics documen	topics documented:	
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ZIQSIR	Provides a novel tool to obtain the p-value for testing associations for zero-inflated response.	

### Description

Provides a novel tool to obtain the p-value for testing associations for zero-inflated response.

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#### Usage

```
ZIQSIR(
   y,
   X,
   taus = c(0.1, 0.25, 0.5, 0.75, 0.9),
   m = 3,
   test_num,
   method = "Chi"
)
```

#### **Arguments**

y n\*1 vector, the observed outcome for n samples.

X n\*p matrix, the observed p covariates for n samples.

taus k\*1 vector, a grid of quantile levels; e.g., 0.5 for the median, 0.75 for the 3rd

quartile; default is c(0.1, 0.25, 0.5, 0.75, 0.9).

m numeric variable, the order of B-spline function; default is 3.

test\_num a vector, representing the test corresponds to which covariate(s) in X.

method different method for calculating p-value: 'large' for large sample cases, where

we use the asymptotic distribution for hypothesis testing; 'small' for small sample cases, where we use the Pearson Type III distribution to approximate the null

distribution.

#### **Details**

- Please choose 'large' or 'small' for method, no other options.
- taus must be a subset or equal to the grid used to produce input.

#### Value

A p-value for testing the association between the covariate(s) of interest and the zero-inflated response.

#### **Examples**

```
# example code
#### demo 1:
# small sample size
# using Pearson Type III method
# alternative distribution
# sample size
set.seed(10001)
n = 500
# the probability of Y>0 given covariate x
p = function(x1,x2,x3,x4,x5,gam0=-0.4,gam1=-0.480,gam2=-0.022,gam3=0.021,gam4=0.015,gam5=-0.009) \\ \{ (x1,x2,x3,x4,x5,gam0=-0.4,gam1=-0.480,gam2=-0.022,gam3=0.021,gam4=0.015,gam5=-0.009) \\ \{ (x1,x2,x3,x4,x5,gam0=-0.4,gam1=-0.480,gam2=-0.022,gam3=0.021,gam4=0.015,gam5=-0.009) \\ \{ (x1,x2,x3,x4,x5,gam0=-0.480,gam1=-0.480,gam2=-0.022,gam3=0.021,gam4=0.015,gam5=-0.009) \\ \{ (x1,x2,x3,x4,x5,gam0=-0.480,gam2=-0.022,gam3=0.021,gam4=0.015,gam5=-0.009) \\ \{ (x1,x2,x3,x4,x5,gam0=-0.480,gam2=-0.022,gam3=0.021,gam4=0.001,gam2=-0.002,gam3=0.002) \\ \{ (x1,x2,x3,x4,x5,gam0=-0.480,gam2=-0.002,gam3=0.002,gam3=0.002,gam3=0.002,gam3=0.002) \\ \{ (x1,x2,x3,x4,x5,gam0=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3=-0.002,gam3
         1c = gam0 + gam1*x1 + gam2*x2 + gam3*x3 + gam4*x4 + gam5*x5
           exp(lc)/(1+exp(lc))
}
# beta_tau
bet1 = function(x)\{(0.3*sqrt(x)-x)*2\}
bet2 = function(x)\{x^2*2.2\}
bet3 = function(x){
```

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```
(x^2-0.5*x+0.6)*2/3
bet5 = function(x)\{-(0.3*x^2-x)*2\}
bet4 = function(x){-sin(x*2*pi)*0.1}
bet0 = function(x)\{-147.7*x-50*x^2-20\}
bet = function(x,u)\{x^4*u*10^(-5)/6+x^2*u*0.2/3\}
# G_tau function
func <- function(x, tau)</pre>
 return(bet(x %*% rbind(bet1(tau),bet2(tau),bet3(tau),bet4(tau),bet5(tau))+bet0(tau),tau))
}
# the given covariate x
X1 = c(0,1) # sex
X2 = qnorm(0.5, 28, 2) # bmi
X3 = qnorm(0.5, 92.5, 13) # waist
X4 = qnorm(0.5, 80, 12) # diastolic_bp
X5 = qnorm(0.5, 124, 18.5) # systolic_bp
X0 = cbind(c(rep(X1[1], 1), rep(X1[2], 1)), rep(X2, 2), rep(X3, 2), rep(X4, 2), rep(X5, 2))
# given samples
x1 = rbinom(n, 1, 0.5) \# Medicament use
x2 = rnorm(n, 28, 2) # bmi
x3 = rnorm(n, 92.5, 13) # waist
x4 = rnorm(n,80,12) # diastolic_bp
x5 = rnorm(n, 124, 18.5) # systolic_bp
x0 = rep(1,n)
X = cbind(x0,x1,x2,x3,x4,x5)
u = runif(n)
b = rbinom(n, 1, p(x1, x2, x3, x4, x5))
w = bet(bet1(u)*x1+bet2(u)*x2+bet3(u)*x3+bet4(u)*x4+bet5(u)*x5+bet0(u),u)
y = b*w
ZIQSIR(y,X,m = 3,test_num = 4,method = "small")
### demo 2
# simulation results under large sample size
# using Chi-square method
# under null hypothesis
# sample size
n = 2000
# the probability of Y>0 given covariate x
p = function(x1,x2,x3,x4,x5,x6,
         gam0=2.32, gam1=-0.06, gam2=-0.03*0, gam3=-0.010*0, gam4=-0.005, gam5=0.0005, gam6=-0.030)
  1c = gam0 + gam1*x1 + gam2*x2 + gam3*x3 + gam4*x4 + gam5*x5 + gam6*x6
  exp(lc)/(1+exp(lc))
# beta_tau
bet1 = function(x){5*x^2+1}
bet2 = function(x)\{(0.1*sin(x*2*pi)+0.05)*0\}
bet3 = function(x){
  (0.05*(x-0.5)^2+0.04)*0
bet4 = function(x)\{(0.1*sqrt(x)+2.9*x)*0.05\}
bet5 = function(x)\{(0.4*(x-1)^2)*0.1\}
bet6 = function(x){((x-0.6)*(x-1.1))*1.1}
bet0 = function(x)\{62.9*x^2+33.4*x\}
```

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```
bet = function(x,u)\{x^4*u*10^(-4)*0.4+x^3*u*10^{-3}*0.1\}
# G_tau function
func <- function(x, tau)</pre>
 return(bet(x %*% rbind(bet1(tau),bet2(tau),bet3(tau),bet4(tau),bet5(tau),bet6(u))+bet0(tau),tau))
}
# given samples
x1 = rbinom(n, 1, 0.5) # Gender
x2 = rnorm(n, 28, 2) \# bmi
x3 = 2*x2+rnorm(n, 36.5, 9) # waist
x4 = rnorm(n, 80, 12) # diastolic_bp
x5 = x4*1.3 + rnorm(n, 20, 7.75) # systolic_bp
x6 = sample(1:4, n, replace = TRUE, prob = c(0.25, 0.025, 0.07, 0.625))
x0 = rep(1,n)
u = runif(n)
b = rbinom(n,1,p(x1,x2,x3,x4,x5,x6))
w = bet(bet1(u)*x1+bet2(u)*x2+bet3(u)*x3+bet4(u)*x4+bet5(u)*x5+bet6(u)*x6+bet0(u),u)
y = b*w
X = cbind(x1, x2, x3, x4, x5, x6)
ZIQSIR(y,X,m = 3,test_num = c(2,3),method = "large")
```

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