

# Package ‘GPLSVCM’

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**Type** Package

**Title** Generalized Partial Linear Spatially Varying Coefficient Model

**Version** 0.1.0

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**Depends** R (>= 2.10),  
BPST

**Description** Identify the linear and nonlinear components of the model and fit the corresponding Generalized Partial Linear Spatially Varying Coefficient Model.

**License** GPL (>= 2)

**Encoding** UTF-8

**LazyData** true

**Imports** mgcv,  
MGLM,  
MASS,  
Triangulation,  
plyr,  
graphics,  
stats,  
plot3D,  
boot,  
Matrix

**RoxygenNote** 7.1.1

## R topics documented:

compute_PIs . . . . .	2
Crash_Texas . . . . .	4
CV_fit . . . . .	4
cv_gplsvcm . . . . .	5
Datagenerator . . . . .	6
dev_est . . . . .	7
gplsvcm_est . . . . .	8
gplsvcm_fit . . . . .	9
gplsvcm_fitwMIDF . . . . .	12
gplsvcm_plot . . . . .	14

gplsvcn_predict . . . . .	15
loo_fit . . . . .	16
nb_bps . . . . .	16

<b>Index</b>	<b>17</b>
--------------	-----------

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compute_PIs	<i>Compute the prediction intervals for responses of new test points from a fitted generalized partial linear spatially varying coefficient model.</i>
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## Description

compute\_PIs compute the prediction intervals for responses of new test points from a fitted gplsvcn object based on a selected prediction method among Jackknife, Jackknife+ and K-fold cross validation (CV+), and return the prediction intervals for responses of predicted points.

## Usage

```
compute_PIs(
  Y_train,
  X_train,
  ind_l,
  ind_nl,
  U_train,
  X_pred,
  U_pred,
  V,
  Tr,
  d = 2,
  r = 1,
  lambda = 10^seq(-6, 6, by = 0.5),
  family,
  off = 0,
  r_theta = c(2, 8),
  eps = 0.01,
  method = "CV+",
  cp = 0.95,
  nfold = 10
)
```

## Arguments

Y_train	The response variable, a n by one matrix where n is the number of observations in the training data set .
X_train	The design matrix of n by np where np is the number of covariates. Each row is a vector of the covariates for an observation in the training data set.
ind_l	The vector of the indexes which indicate the columns of linear covariates in X_train.
ind_nl	The vector of the indexes which indicate the columns of nonlinear covariates in X_train.

U_train	A n by two matrix where each row is the coordinates of an observation in the training data set.
X_pred	The design matrix for prediction.
U_pred	The matrix of coordinates for prediction.
V	A N by two matrix of vertices of a triangulation, where N is the number of vertices and each row is the coordinates for a vertex.
Tr	A n_Tr by three triangulation matrix, where n_Tr is the number of triangles in the triangulation and each row is the indices of vertices in V.
d	The degree of piecewise polynomials – default is 2.
r	The smoothness parameter and $r < d$ – default is 1.
lambda	The vector of the candidates of penalty parameter – default is grid points of 10 to the power of a sequence from -6 to 6 by 0.5.
family	The family object which specifies the distribution and link to use (see <a href="#">glm</a> and <a href="#">family</a> ).
off	The offset – default is 0.
r_theta	The vector of the upper and lower bound of an interval to search for an additional parameter theta for negative binomial scenario – default is c(2,8).
eps	The error tolerance for the Pearson estimate of the scale parameter, which is as close to 1, when estimating an additional parameter theta for negative binomial scenario – default is 0.01.
method	The prediction method used in the computation, options are "CV+", "Jackknife" and "Jackknife+" – default is "CV+".
cp	The desired coverage level for the prediction intervals – default is 0.95.
nfold	The number of folds for CV+ method – default is 10.

## Details

The construction of the polynomial spline functions is via [basis](#).

## Value

A data frame of computed prediction intervals for responses of the predicted points.

## References

Barber et al.(2021) Predictive inference with the jackknife+. Ann.Statist.49(1):486-507 <https://projecteuclid.org/journals/annals-of-statistics/volume-49/issue-1/Predictive-inference-with-the-10.1214/20-AOS1965.full>

## Examples

```
# See an example of gplsvcm_fitwMIDF.
```

---

Crash_Texas	<i>Crash data in Texas</i>
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### Description

A dataset containing the number of car crashes within each census tract in Texas of year 2107 and other variables of 4771 locations.

### Usage

```
data(Crash_Texas)
```

### Format

A data frame with 4771 rows and 7 variables:

**count** off-roadway crash frequencies  
**vmt** log of vehicle miles traveled  
**pop** log of total population  
**old** proportion of people age 65 and older  
**hispanics** proportion of Hispanics  
**lon** longitude of a location  
**lat** latitude of a location

### Examples

```
data(Crash_Texas)
count <- Crash_Texas$count
hist(count)
summary(count)
```

---

CV_fit	<i>Fit the model with K fold cross validation and compute the CV residuals</i>
--------	--

---

### Description

This is an internal function of package GPLSVCM.

### Usage

```
CV_fit(
  Y_train,
  X_l_train,
  X_nl_train,
  ind_l,
  ind_nl,
  U_train,
  X_l_pred,
```

```

X_nl_pred,
U_pred,
B_pred,
V,
Tr,
d,
r,
lambda,
family,
off,
r_theta,
eps,
nfold
)

```

cv\_gplsvcm

*Calculate the K-fold Cross Validation Mean Square Prediction Error from a fitted generalized partial linear spatially varying coefficient model.*

## Description

cv\_gplsvcm implements K-fold cross-validation from a fitted gplsvcm object, and returns the mean squared prediction error (MSPE).

## Usage

```

cv_gplsvcm(
  Y,
  X,
  ind_l,
  ind_nl,
  U,
  V,
  Tr,
  d = 2,
  r = 1,
  lambda = 10^seq(-6, 6, by = 0.5),
  family,
  off = 0,
  r_theta = c(2, 8),
  eps = 0.01,
  nfold = 10
)

```

## Arguments

Y	The response variable, a n by one matrix where n is the number of observations.
X	The design matrix of n by np where np is the number of covariates. Each row is a vector of the covariates for an observation.
ind_l	The vector of the indexes which indicate the columns of linear covariates in X.

ind_nl	The vector of the indexes which indicate the columns of nonlinear covariates in $X$ .
U	A $n$ by two matrix where each row is the coordinates of an observation.
V	A $N$ by two matrix of vertices of a triangulation, where $N$ is the number of vertices and each row is the coordinates for a vertex.
Tr	A $n_{Tr}$ by three triangulation matrix, where $n_{Tr}$ is the number of triangles in the triangulation and each row is the indices of vertices in $V$ .
d	The degree of piecewise polynomials – default is 2.
r	The smoothness parameter and $r < d$ – default is 1.
lambda	The vector of the candidates of penalty parameter – default is grid points of 10 to the power of a sequence from -6 to 6 by 0.5.
family	The family object which specifies the distribution and link to use (see <a href="#">glm</a> and <a href="#">family</a> ).
off	The offset – default is 0.
r_theta	The vector of the upper and lower bound of an interval to search for an additional parameter $\theta$ for negative binomial scenario – default is $c(2,8)$ .
eps	The error tolerance for the Pearson estimate of the scale parameter, which is as close to 1, when estimating an additional parameter $\theta$ for negative binomial scenario – default is 0.01.
nfold	The number of folds for cross validation – default is 10.

### Details

The construction of the polynomial spline functions is via [basis](#).

### Value

The mean square prediction error (MSPE).

### Examples

```
# See an example of gplsvcm_fitwMIDF.
```

---

Datagenerator

*Generating populations for simulation.*

---

### Description

Datagenerator is used to generate samples on horseshoe domain for Scenario 1 (Gaussian), Scenario 2 (Poisson) and Scenario 3 (negative binomial).

### Usage

```
Datagenerator(family, ngrid)
```

**Arguments**

family	The family object, specifying the distribution and link to use. Choose "gaussian" for Gaussian distribution, "poisson" for poisson distribution, and "nb_bps" for negative binomial distribution.
ngrid	The distance between grid points – default is set to 0.02.

**Details**

This function used the package mgcv, see [fs.boundary](#) and [fs.test](#)

**Value**

A data matrix with a response ('y'), true coefficient functions ('m1' and 'm2'), nonlinear covariates ('x1' and 'x2'), linear covariates ('x3' and 'x4') and locations ('u' and 'v').

**Examples**

```
family=nb_bps()
ngrid = 0.02
pop = Datagenerator(family, ngrid)
```

---

dev_est	<i>Calculate the generalized deviance of a fitted model</i>
---------	---

---

**Description**

dev\_est is an internal function of gplsvcm\_fitwMIDF.

**Usage**

```
dev_est(X_l, X_nl, mfit, family)
```

**Arguments**

X_l	The matrix of linear covariates for observations.
X_nl	The matrix of nonlinear covariates for observations.
mfit	A list of information returned by function gplsvcm_est
family	The family object which specifies the distribution and link to use.

gplsvcm\_est

*Estimation for GPLSCVMs***Description**

This is an internal function of GPLSVCM which is used in function `gplsvcm_fit` and `gplsvcm_fitwMIDF`.

**Usage**

```
gplsvcm_est(Y, X_l, X_nl, U, V, Tr, d, r, lambda, family, off, r_theta, eps)
```

**Arguments**

<code>Y</code>	The response variable, a $n$ by one matrix where $n$ is the number of observations.
<code>X_l</code>	The matrix of linear covariates for observations, of dimension $n$ by $np\_l$ where $n$ is number of observations and $np\_l$ is the number of linear covariates.
<code>X_nl</code>	The matrix of nonlinear covariates for observations, of dimension $n$ by $np\_nl$ where $n$ is number of observations and $np\_nl$ is the number of nonlinear covariates.
<code>U</code>	A $n$ by two matrix where each row is the coordinates of an observation.
<code>V</code>	A $N$ by two matrix of vertices of a triangulation, where $N$ is the number of vertices and each row is the coordinates for a vertex.
<code>Tr</code>	A $n\_Tr$ by three triangulation matrix, where $n\_Tr$ is the number of triangles in the triangulation and each row is the indices of vertices in $V$ .
<code>d</code>	The degree of piecewise polynomials.
<code>r</code>	The smoothness parameter and $r < d$ .
<code>lambda</code>	The vector of the candidates of penalty parameter.
<code>family</code>	The family object which specifies the distribution and link to use (see <a href="#">glm</a> and <a href="#">family</a> ).
<code>off</code>	The offset.
<code>r_theta</code>	The vector of the upper and lower bound of an interval to search for an additional parameter $\theta$ for negative binomial scenario.
<code>eps</code>	The error tolerance for the Pearson estimate of the scale parameter, which is as close as possible to 1, when estimating an additional parameter $\theta$ for negative binomial scenario.

**Details**

The construction of the polynomial spline functions is via [basis](#).



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gplsvcm_fit	<i>Fitting generalized partial linear spatially varying coefficient regression models</i>
-------------	---

---

## Description

`gplsvcm_fit` fits the generalized partial linear spatially varying coefficient models.

## Usage

```
gplsvcm_fit(
  Y,
  X,
  ind_l,
  ind_nl,
  U,
  V,
  Tr,
  d = 2,
  r = 1,
  lambda = 10^seq(-6, 6, by = 0.5),
  family,
  off = 0,
  r_theta = c(2, 8),
  eps = 0.01
)
```

## Arguments

<code>Y</code>	The response variable, a $n$ by one matrix where $n$ is the number of observations.
<code>X</code>	The design matrix of $n$ by $n_p$ where $n_p$ is the number of covariates. Each row is a vector of the covariates for an observation.
<code>ind_l</code>	The vector of the indexes which indicate the columns of linear covariates in <code>X</code> .
<code>ind_nl</code>	The vector of the indexes which indicate the columns of nonlinear covariates in <code>X</code> .
<code>U</code>	A $n$ by two matrix where each row is the coordinates of an observation.
<code>V</code>	A $N$ by two matrix of vertices of a triangulation, where $N$ is the number of vertices and each row is the coordinates for a vertex.
<code>Tr</code>	A $n_{Tr}$ by three triangulation matrix, where $n_{Tr}$ is the number of triangles in the triangulation and each row is the indices of vertices in <code>V</code> .
<code>d</code>	The degree of piecewise polynomials – default is 2.
<code>r</code>	The smoothness parameter and $r < d$ – default is 1.
<code>lambda</code>	The vector of the candidates of penalty parameter – default is grid points of 10 to the power of a sequence from -6 to 6 by 0.5.
<code>family</code>	The family object which specifies the distribution and link to use (see <a href="#">glm</a> and <a href="#">family</a> ).
<code>off</code>	The offset – default is 0.

r_theta	The vector of the upper and lower bound of an interval to search for an additional parameter theta for negative binomial scenario – default is c(2,8).
eps	The error tolerance for the Pearson estimate of the scale parameter, which is as close to 1, when estimating an additional parameter theta for negative binomial scenario – default is 0.01.

## Details

The `gplsvcm_fit` function is for fitting the Generalized Partial Linear Spatially Varying Coefficient Models (GPLSVCM) when the model structure is specified before analysis, that is, the parameters `ind_l` and `ind_nl` are specified before fitting the model. The construction of the polynomial spline functions is via [basis](#). If the true model structure is not known before model fitting, we recommend using another function `gplsvcm_fitwMIDF` in this package. Note, if `ind_l` is specified as a null vector, `gplsvcm_fit` will fit a glm model, and if `ind_nl` is specified as a null vector, `gplsvcm_fit` will fit a gsvcm model.

## Value

The function returns a list of fitted object information from S3 class "gplsvcm" with the following items:

alpha_hat	The estimated coefficients for the nonlinear component of the model.
beta_hat	The estimated coefficients for the linear component of the model.
Qtheta	The estimated spline coefficients.
lambda_sel	The selected penalty parameter through generalized cross-validation (GCV) for bivariate penalized spline over triangulation estimation.
gcv	The GCV statistics for <code>lambda_sel</code> .
df	The effective degree of freedom for the model.
theta	The estimated additional parameter theta for negative binomial scenario.
Y	The matrix of responses, of dimension n by one where n is number of observations inside the triangulation.
X_nl	The matrix of nonlinear covariates for observations inside the triangulation, of dimension n by np_l where n is number of observations inside the triangulation and np_l is the number of linear covarites.
X_l	The matrix of linear covariates for observations inside the triangulation, n by np_l where n is number of observations inside the triangulation and np_nl is the number of nonlinear covarites.
U	The matrix of coordinates for observations inside the triangulation, of dimension n by 2 where n is number of observations inside the triangulation and each row is the coordinates of an observation.
ind_l	The vector of the indexes which indicate the columns of linear covariates in X.
ind_nl	The vector of the indexes which indicate the columns of nonlinear covariates in X.
family	The family object.
V	The matrix of vertices of the triangulation, with dimension N by two where N is the number of vertices of the triangulation and each row is the coordinates for a vertex

Tr	The triangulation matrix of of the triangulation, with dimention $n_{Tr}$ by three, where $n_{Tr}$ is the number of triangles in the triangulation and each row is the indices of vertices in V.
d	The degree of piecewise polynomials.
r	The smoothness parameter.
B	The spline basis function of dimension $n$ by $n_{Tr} * \{(d+1)(d+2)/2\}$ , where $n$ and $n_{Tr}$ are the number of observations and the number of triangles inside the given triangulation respectively, $d$ is the degree of the spline. If some points do not fall in the triangulation, the generation of the spline basis will not take those points into consideration.
Q2	The Q2 matrix after QR decomposition of the smoothness matrix H.
K	The thin-plate energy function.
ind_inside	A vector contains the indexes of all the points which are inside the triangulation.
tria_all	The area of each triangle within the given triangulation.
lambda	The vector of the candidates of penalty parameter used in fitting the model.
r_theta	The vector of the upper and lower bound of an interval to search for an additional parameter theta used in negative binomial scenario.
off	The offset.
eps	The error tolerance used for the Pearson estimate of the scale parameter for negative binomial scenario.

## Examples

```
# Population:
family=poisson()
ngrid = 0.02

# Data generation:
pop = Datagenerator(family, ngrid)
N=nrow(pop)

# Triangulations and setup:
Tr = Tr0; V = V0; n = 1000; d = 2; r = 1;

# set up for smoothing parameters in the penalty term:
lambda_start=0.0001; lambda_end=10; nlambdas=10
lambda=exp(seq(log(lambda_start),log(lambda_end),length.out=nlambdas))

# Generate Sample:
ind_s=sample(N,n,replace=FALSE)
data=as.matrix(pop[ind_s,])
Y=data[,1]; alpha=data[,c(2:3)]; beta=data[,c(4:5)]; X=data[,c(6:9)];
ind_l=c(3,4); ind_nl=c(1,2); U=data[,c(10:11)];

# Fit the model:
mfit = gplsvcm_fit(Y, X,ind_l,ind_nl,U, V, Tr, d , r , lambda,family,off = 0,
r_theta = c(2, 8), eps= 0.01)
```

---

gplsvcm_fitwMIDF	<i>Fitting the generalized partial linear spatially varying coefficient model with Model Selection</i>
------------------	--

---

## Description

gplsvcm\_fitwMIDF perform a model selection procedure to identify the linear and nonlinear components first and then fit the corresponding generalized partial linear spatially varying coefficient model.

## Usage

```
gplsvcm_fitwMIDF(
  Y,
  X,
  U,
  V,
  Tr,
  d = 2,
  r = 1,
  lambda = 10^seq(-6, 6, by = 0.5),
  family,
  k_n = NULL,
  method = "BIC",
  off = 0,
  r_theta = c(2, 8),
  eps = 0.01
)
```

## Arguments

Y	The response variable, a $n$ by one matrix where $n$ is the number of observations.
X	The design matrix of $n$ by $n_p$ where $n_p$ is the number of covariates. Each row is a vector of the covariates for an observation.
U	A $n$ by two matrix where each row is the coordinates of an observation.
V	A $N$ by two matrix of vertices of a triangulation, where $N$ is the number of vertices and each row is the coordinates for a vertex.
Tr	A $n_{Tr}$ by three triangulation matrix, where $n_{Tr}$ is the number of triangles in the triangulation and each row is the indices of vertices in $V$ .
d	The degree of piecewise polynomials – default is 2.
r	The smoothness parameter and $r < d$ – default is 1.
lambda	The vector of the candidates of penalty parameter – default is grid points of 10 to the power of a sequence from -6 to 6 by 0.5.
family	The family object which specifies the distribution and link to use (see <a href="#">glm</a> and <a href="#">family</a> ).
k_n	The penalty parameter used in the model selection criteria. It need to be supplied only when the argument method is set to NULL –default is NULL.

method	The type of model selection criteria, options are "AIC", "BIC" and NULL which correspond to $k_n=2$ , $k_n=\log(n)$ and $k_n=k_n$ respectively – default is "BIC".
off	The offset – default is 0.
r_theta	The vector of the upper and lower bound of an interval to search for an additional parameter theta for negative binomial scenario – default is c(2,8).
eps	The error tolerance for the Pearson estimate of the scale parameter, which is as close as possible to 1, when estimating an additional parameter theta for negative binomial scenario – default is 0.01.

## Details

The `gplsvcm_fitwMIDF` function is used to fit a generalized partial linear spatially varying coefficient model when the linear and nonlinear parts of the design matrix  $X$  are not known before analysis. The construction of the polynomial spline functions is via `basis`. It first perform a model selection based on Generalized Information Criterion (GIC) and output the selected model by specifying the parameters `ind_l` and `ind_nl` of the function `gplsvcm_fit`. Then the selected model is fitted by the function `gplsvcm_fit`.

## Value

The function returns a list of fitted object information from S3 class "gplsvcm", see the items of the list from `gplsvcm_fit`.

## References

Zhang et al.(2010) Regularization Parameter Selections via Generalized Information Criterion.<https://www.tandfonline.com/doi/abs/10.1198/jasa.2009.tm08013>

## Examples

```
# Population:
family=poisson()
ngrid = 0.02

# Data generation:
pop = Datagenerator(family, ngrid)
N=nrow(pop)

# Triangulations and setup:
Tr = Tr0; V = V0; n = 1000; d = 2; r = 1;

# set up for smoothing parameters in the penalty term:
lambda_start=0.0001; lambda_end=10; nlambdas=10;
lambda=exp(seq(log(lambda_start),log(lambda_end),length.out=nlambdas))

# Generate Sample:
ind_s=sample(N,n,replace=FALSE)
data=as.matrix(pop[ind_s,])
Y=data[,1]; X=data[,c(6:9)]; U=data[,c(10:11)];

# True coefficients
alpha=data[,c(2:3)]; beta=data[,c(4:5)];

# Fit the model with model selection based on AIC:
mfit1 = gplsvcm_fitwMIDF(Y, X, U, V, Tr, d, r, lambda,family,k_n=NULL,
```

```

method="AIC",off = 0,r_theta = c(2, 8), eps= 0.01)

# Fit the model with model selection based on BIC:
mfit2 = gplsvcm_fitwMIDF(Y, X, U, V, Tr, d , r , lambda,family,k_n=NULL,
method="BIC",off = 0,r_theta = c(2, 8), eps= 0.01)

# prediction intervals:
ind_l=mfit2$ind_l; ind_nl=mfit2$ind_nl;
set.seed(123)
PIs=compute_PIs(Y,X,ind_l,ind_nl,U,X,U,V,Tr,d,r,lambda,family,off = 0,
r_theta = c(2, 8), eps= 0.01,method="CV+", cp=0.95, nfold = 10)

# prediction:
Y_hat = gplsvcm_predict(mfit2, X, U)

# k-fold cross-validation:
set.seed(123)
MSPE = cv_gplsvcm(Y,X,ind_l,ind_nl,U,V,Tr,d,r,lambda,family,off = 0,r_theta =
c(2, 8), eps= 0.01,nfold=10)

# plot the estimated coefficients
gplsvcm_plot(mfit2,gridnumber=100,display=c(1,1),xlab=c("u1","u1"),
ylab=c("u2","u2"),main=c(expression(paste("The Estimated Surface for",
",hat(alpha)[1])),expression(paste("The Estimated Surface for",
",hat(alpha)[2]"))))

```

---

gplsvcm\_plot

---

*Produces coefficient function plots for a fitted generalized partial linear spatially varying coefficient model.*


---

## Description

gplsvcm\_plot produces the plots of the estimated coefficient functions from a fitted gplsvcm object.

## Usage

```

gplsvcm_plot(
  mfit,
  gridnumber = 100,
  display = NULL,
  xlab = NULL,
  ylab = NULL,
  main = NULL,
  ...
)

```

## Arguments

mfit	A fitted gplsvcm object returned from function gplsvcm_fit or gplsvcm_fitwMIDF.
gridnumber	The number of grid points on one range for plots – default is 100.

display	If supplied then it is the vector for specifying how to display the estimated surfaces for the coefficient functions, used in <code>par(mfrow=)</code> .
xlab	If supplied then is the vector of characters where each element is the x label for the estimated surface of one coefficient function.
ylab	If supplied then is the vector of characters where each element is the y label for the estimated surface of one coefficient function.
main	If supplied then is the vector of characters where each element is the title for the estimated surface of one coefficient function.
...	other graphics parameters to pass on to plotting commands. See details in <a href="#">image2D</a> .

### Details

This function used package `Triangulation` and `plot3D`, see [TriPlot](#) and [image2D](#).

### Value

None

### Examples

```
# See an example of gplsvcm_fitwMIDF.
```

---

<code>gplsvcm_predict</code>	<i>Predictions for responses of new test points from a fitted generalized partial linear spatially varying coefficient model.</i>
------------------------------	---

---

### Description

`gplsvcm_predict` is used to make predictions for the responses of predicted points from a fitted `gplsvcm` object.

### Usage

```
gplsvcm_predict(mfit, Xpred, Upred)
```

### Arguments

mfit	A fitted <code>gplsvcm</code> object returned from function <code>gplsvcm_fit</code> or <code>gplsvcm_fitwMIDF</code> .
Xpred	The design matrix for prediction.
Upred	The matrix of coordinates for prediction.

### Details

The construction of the polynomial spline functions is via [basis](#)

### Value

A vector of predicted response.

**Examples**

```
# See an example of gplsvcm_fitwMIDF.
```

---

loo_fit	<i>Fit the model with the i-th training data point removed and compute the leave one out residuals</i>
---------	--

---

**Description**

This is an internal function of package GPLSVCM.

**Usage**

```
loo_fit(
  Y_train,
  X_l_train,
  X_nl_train,
  ind_l,
  ind_nl,
  U_train,
  X_l_pred,
  X_nl_pred,
  U_pred,
  B_pred,
  V,
  Tr,
  d,
  r,
  lambda,
  family,
  off,
  r_theta,
  eps
)
```

---

nb_bps	<i>Negative Binomial Family</i>
--------	---------------------------------

---

**Description**

Negative Binomial Family

**Usage**

```
nb_bps(link = "log", theta)
```

**Details**

This is a built in function in GgAM.



# Index

## \* datasets

Crash\_Texas, [4](#)

basis, [3](#), [6](#), [8](#), [10](#), [13](#), [15](#)

compute\_PIs, [2](#)

Crash\_Texas, [4](#)

CV\_fit, [4](#)

cv\_gplsvcm, [5](#)

Datagenerator, [6](#)

dev\_est, [7](#)

family, [3](#), [6](#), [8](#), [9](#), [12](#)

fs.boundary, [7](#)

fs.test, [7](#)

glm, [3](#), [6](#), [8](#), [9](#), [12](#)

gplsvcm\_est, [8](#)

gplsvcm\_fit, [9](#), [13](#)

gplsvcm\_fitwMIDF, [12](#)

gplsvcm\_plot, [14](#)

gplsvcm\_predict, [15](#)

image2D, [15](#)

loo\_fit, [16](#)

nb\_bps, [16](#)

TriPlot, [15](#)