

# Package: conduits (via r-universe)

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## Contents

augment.conditional_acf . . . . .	2
augment.conditional_ccf . . . . .	3
augment.conditional_moment . . . . .	5
calc_dt_CI . . . . .	6
conditional_acf . . . . .	7

conditional_ccf . . . . .	9
conditional_mean . . . . .	11
conditional_var . . . . .	12
conduits . . . . .	14
estimate_dt . . . . .	14
NEON_PRIN_5min_cleaned . . . . .	16
normalize . . . . .	16
unnormalize . . . . .	17

## Index 19

augment.conditional\_acf

*Augment data with information from a conditional auto-correlation fit*

### Description

This function produces estimated conditional autocorrelation between  $x_t$  and  $y_t$  at lag  $k$ , i.e.  $r_k = E(x_{t+k}y_t)$ .

### Usage

```
## S3 method for class 'conditional_acf'
augment(x, ...)
```

### Arguments

`x` Model object of class "conditional\_acf" returned from `conditional_acf` with information to append to observations.

`...` Additional arguments, not currently used.

### Value

A `tibble` with information about data points.

### Examples

```
old_ts <- NEON_PRIN_5min_cleaned |>
  dplyr::select(
    Timestamp, site, turbidity, level,
    conductance, temperature
  ) |>
  tidyr::pivot_wider(
    names_from = site,
    values_from = turbidity:temperature
  )

fit_mean <- old_ts |>
  conditional_mean(turbidity_downstream ~
```

```

s(level_upstream, k = 8) +
s(conductance_upstream, k = 8) +
s(temperature_upstream, k = 8))

fit_var <- old_ts |>
conditional_var(
  turbidity_downstream ~
    s(level_upstream, k = 7) +
    s(conductance_upstream, k = 7) +
    s(temperature_upstream, k = 7),
  family = "Gamma",
  fit_mean = fit_mean
)
fit_c_acf <- old_ts |>
tidyr::drop_na() |>
conditional_acf(
  turbidity_upstream ~ splines::ns(level_upstream, df = 5) +
    splines::ns(conductance_upstream, df = 5),
  lag_max = 10, fit_mean = fit_mean, fit_var = fit_var,
  df_correlation = c(5, 5)
)

data_inf <- fit_c_acf |> augment()

```

---

```
augment.conditional_ccf
```

*Augment data with information from a conditional cross-correlation fit*

---

## Description

This function produces estimated conditional cross-correlation between  $x_t$  and  $y_t$  at lag  $k$ , i.e.  $r_k = E(x_{t+k}y_t)$ .

## Usage

```
## S3 method for class 'conditional_ccf'
augment(x, ...)
```

## Arguments

<code>x</code>	Model object of class "conditional_ccf" returned from <code>conditional_ccf</code> with information to append to observations.
<code>...</code>	Additional arguments, not currently used.

## Value

A `tibble` with information about data points.

**Examples**

```

old_ts <- NEON_PRIN_5min_cleaned |>
  dplyr::select(
    Timestamp, site, turbidity, level,
    conductance, temperature
  ) |>
  tidyr::pivot_wider(
    names_from = site,
    values_from = turbidity:temperature
  )

fit_mean_y <- old_ts |>
  conditional_mean(turbidity_downstream ~
    s(level_upstream, k = 8) +
    s(conductance_upstream, k = 8) +
    s(temperature_upstream, k = 8))

fit_var_y <- old_ts |>
  conditional_var(
    turbidity_downstream ~
    s(level_upstream, k = 7) +
    s(conductance_upstream, k = 7) +
    s(temperature_upstream, k = 7),
    family = "Gamma",
    fit_mean = fit_mean_y
  )

fit_mean_x <- old_ts |>
  conditional_mean(turbidity_upstream ~
    s(level_upstream, k = 8) +
    s(conductance_upstream, k = 8) +
    s(temperature_upstream, k = 8))

fit_var_x <- old_ts |>
  conditional_var(
    turbidity_upstream ~
    s(level_upstream, k = 7) +
    s(conductance_upstream, k = 7) +
    s(temperature_upstream, k = 7),
    family = "Gamma",
    fit_mean = fit_mean_x
  )

fit_c_ccf <- old_ts |>
  tidyr::drop_na() |>
  conditional_ccf(
    I(turbidity_upstream * turbidity_downstream) ~ splines::ns(
      level_upstream,
      df = 5
    ) +
    splines::ns(conductance_upstream, df = 5),
    lag_max = 10,
  )

```

```

    fit_mean_x = fit_mean_x, fit_var_x = fit_var_x,
    fit_mean_y = fit_mean_y, fit_var_y = fit_var_y,
    df_correlation = c(5, 5)
  )

data_inf <- fit_c_ccf |> augment()

```

---

```
augment.conditional_moment
```

*Augment data with information from a conditional mean fit or conditional variance fit*

---

## Description

This function produces partial residuals for each predictor, and the estimated conditional means, standard error and confidence limits.

## Usage

```

## S3 method for class 'conditional_moment'
augment(x, level = 0.95, ...)

```

## Arguments

x	Model object of class "conditional_moment" returned from <a href="#">conditional_mean</a> or <a href="#">conditional_var</a> with information to append to observations.
level	Confidence level. Default is set to 0.95.
...	Additional arguments, not currently used

## Value

A [tibble](#) with information about data points.

## See Also

[gam](#)

## Examples

```

data <- NEON_PRIN_5min_cleaned |>
  dplyr::filter(site == "upstream") |>
  dplyr::select(Timestamp, turbidity, level, conductance, temperature)

fit_mean <- data |>
  conditional_mean(turbidity ~ s(level, k = 8) +
    s(conductance, k = 8) + s(temperature, k = 8))

data_inf <- fit_mean |> augment()

```

---

`calc_dt_CI`*Computing bootstrapped confidence intervals for dt*

---

### Description

This function computes the bootstrapped confidence intervals for `dt`. It resamples the residuals from the various models used in the conditional cross-correlation calculation to generate new data. As the residuals are serially correlated, a sieve bootstrap approach is used to capture the autocorrelation structure in the data.

### Usage

```
calc_dt_CI(x, m, new_data = NULL)
```

### Arguments

<code>x</code>	Model object of class "conditional_ccf" returned from <code>conditional_ccf</code>
<code>m</code>	number of replications for bootstrap confidence intervals
<code>new_data</code>	the dataset with the some predictors that are set to the median value (if required). Default is set to NULL.

### Value

A `tibble` with estimated time lag "dt"

### Author(s)

Priyanga Dilini Talagala & Puwasala Gamakumara

### Examples

```
## Not run:
old_ts <- NEON_PRIN_5min_cleaned |>
  dplyr::select(
    Timestamp, site, turbidity, level, temperature
  ) |>
  tidyr::pivot_wider(
    names_from = site,
    values_from = turbidity:temperature
  )
fit_mean_y <- old_ts |>
  conditional_mean(turbidity_downstream ~
    s(level_upstream, k = 5) +
    s(temperature_upstream, k = 5)
  )
fit_var_y <- old_ts |>
  conditional_var(
    turbidity_downstream ~
```

```

      s(level_upstream, k = 4) +
      s(temperature_upstream, k = 4),
      family = "Gamma",
      fit_mean = fit_mean_y
    )
fit_mean_x <- old_ts |>
  conditional_mean(turbidity_upstream ~
    s(level_upstream, k = 5) +
    s(temperature_upstream, k = 5)
  )
fit_var_x <- old_ts |>
  conditional_var(
    turbidity_upstream ~
    s(level_upstream, k = 4) +
    s(temperature_upstream, k = 4),
    family = "Gamma",
    fit_mean = fit_mean_x
  )
fit_c_ccf <- old_ts |>
  tidyr::drop_na() |>
  conditional_ccf(
    I(turbidity_upstream * turbidity_downstream) ~
    splines::ns(level_upstream, df = 3) +
    splines::ns(temperature_upstream, df = 3),
    lag_max = 10,
    fit_mean_x = fit_mean_x, fit_var_x = fit_var_x,
    fit_mean_y = fit_mean_y, fit_var_y = fit_var_y,
    df_correlation = c(3, 3)
  )
df_dt <- fit_c_ccf |> calc_dt_CI(100)

# Calculate dt vs an upstream covariate while holding the
# remaining upstream covariates at their medians
new_data <- fit_c_ccf$data
new_data <- new_data |>
  dplyr::mutate(temperature_upstream = median(temperature_upstream))
df_dt2 <- fit_c_ccf |> calc_dt_CI(100, new_data)

## End(Not run)

```

---

conditional\_acf

*Computing conditional autocorrelations at given lags*


---

### Description

This function computes autocorrelation between  $x_{t+k}$  and  $y_{t+k}$  at  $k = 1, 2, \dots$  conditional on a set of time series  $z_t$

**Usage**

```
conditional_acf(data, formula, lag_max, fit_mean, fit_var, df_correlation)
```

**Arguments**

data	a tibble containing all the time series including $y_{t-k}$ which are uniquely identified by the corresponding Timestamp.
formula	A GAM formula. See <a href="#">formula.gam</a> .
lag_max	Maximum lag at which to calculate the conditional acf
fit_mean	Model object of class "conditional_moment" returned from <a href="#">conditional_mean</a>
fit_var	Model object of class "conditional_moment" returned from <a href="#">conditional_var</a>
df_correlation	a vector specifying the degrees of freedom to be considered for each numerical predictor when fitting additive models for conditional auto-correlations. Each component of the vector should correspond to each predictor specified in "z_numeric".

**Details**

Suppose  $x_t$  and  $y_t$  are conditionally normalised with respect to  $z_t$  using [conditional\\_mean](#) and [conditional\\_var](#). Then we can estimate the conditional cross-correlation between  $x_t$  and  $y_t$  at lag  $k$ , i.e.  $r_k = E(x_{t+k}y_t | z_t)$  via generalised additive models (GAM). [conditional\\_ccf](#) uses natural splines implemented in [splines](#) package to estimate the conditional cross-correlations between two time series given a set of time series predictors. Users first need to normalise  $x_t$  and  $y_t$  at lag  $k$  using [conditional\\_mean](#) and [conditional\\_var](#)

**Value**

The function returns a list of objects of class "glm" as described in [glm](#).

**See Also**

[glm](#)

**Examples**

```
old_ts <- NEON_PRIN_5min_cleaned |>
  dplyr::select(
    Timestamp, site, turbidity, level,
    conductance, temperature
  ) |>
  tidyr::pivot_wider(
    names_from = site,
    values_from = turbidity:temperature
  )

fit_mean <- old_ts |>
  conditional_mean(turbidity_downstream ~
    s(level_upstream, k = 8) +
    s(conductance_upstream, k = 8) +
```



```

s(temperature_upstream, k = 8))

fit_var <- old_ts |>
conditional_var(
  turbidity_downstream ~
    s(level_upstream, k = 7) +
    s(conductance_upstream, k = 7) +
    s(temperature_upstream, k = 7),
  family = "Gamma",
  fit_mean = fit_mean
)
fit_c_acf <- old_ts |>
tidyr::drop_na() |>
conditional_acf(
  turbidity_upstream ~ splines::ns(level_upstream, df = 5) +
    splines::ns(conductance_upstream, df = 5),
  lag_max = 10, fit_mean = fit_mean, fit_var = fit_var,
  df_correlation = c(5, 5)
)

```

---

conditional\_ccf

*Computing conditional cross-correlations at given lags*


---

### Description

This function computes cross correlation between  $x_t$  and  $y_{t+k}$  at  $k = 1, 2, \dots$  conditional on a set of time series  $z_t$

### Usage

```

conditional_ccf(
  data,
  formula,
  lag_max = 10,
  fit_mean_x,
  fit_var_x,
  fit_mean_y,
  fit_var_y,
  df_correlation
)

```

### Arguments

data	a tibble containing all the time series including ystar*xstar which are uniquely identified by the corresponding Timestamp.
formula	A GAM formula. The response variable should be in the format of $I(x*y) \sim .$ See <a href="#">formula.gam</a> .
lag_max	Maximum lag at which to calculate the conditional ccf

fit_mean_x	Model object of class "conditional_moment" returned from <code>conditional_mean</code> for series x
fit_var_x	Model object of class "conditional_moment" returned from <code>conditional_var</code> for series x
fit_mean_y	Model object of class "conditional_moment" returned from <code>conditional_mean</code> for series y
fit_var_y	Model object of class "conditional_moment" returned from <code>conditional_var</code> for series y
df_correlation	a vector specifying the degrees of freedom to be considered for each numerical predictor when fitting additive models for conditional cross-correlations. Each component of the vector should correspond to the degrees of freedom each predictor.

### Details

Suppose  $x_t$  and  $y_t$  are conditionally normalised with respect to  $z_t$  using `conditional_mean` and `conditional_var`. Then we can estimate the conditional cross-correlation between  $x_t$  and  $y_t$  at lag  $k$ , i.e.  $r_k = E(x_{t+k}y_t)$  via generalised additive models (GAM). `conditional_ccf` uses natural splines implemented in `splines` package to estimate the conditional cross-correlations between two time series given a set of time series predictors. Users first need to normalise  $x_t$  and  $y_t$  at lag  $k$  using `conditional_mean` and `conditional_var`

### Value

The function returns a list of objects of class "glm" as described in [glm](#). the length of the list is equal to `lag_max`

### See Also

[glm](#)

### Examples

```
old_ts <- NEON_PRIN_5min_cleaned |>
  dplyr::select(
    Timestamp, site, turbidity, level,
    conductance, temperature
  ) |>
  tidyr::pivot_wider(
    names_from = site,
    values_from = turbidity:temperature
  )

fit_mean_y <- old_ts |>
  conditional_mean(turbidity_downstream ~
    s(level_upstream, k = 8) +
    s(conductance_upstream, k = 8) +
    s(temperature_upstream, k = 8))

fit_var_y <- old_ts |>
```

```

conditional_var(
  turbidity_downstream ~
    s(level_upstream, k = 7) +
    s(conductance_upstream, k = 7) +
    s(temperature_upstream, k = 7),
  family = "Gamma",
  fit_mean = fit_mean_y
)

fit_mean_x <- old_ts |>
conditional_mean(turbidity_upstream ~
  s(level_upstream, k = 8) +
  s(conductance_upstream, k = 8) +
  s(temperature_upstream, k = 8))

fit_var_x <- old_ts |>
conditional_var(
  turbidity_upstream ~
    s(level_upstream, k = 7) +
    s(conductance_upstream, k = 7) +
    s(temperature_upstream, k = 7),
  family = "Gamma",
  fit_mean = fit_mean_x
)

fit_c_ccf <- old_ts |>
tidyr::drop_na() |>
conditional_ccf(
  I(turbidity_upstream * turbidity_downstream) ~ splines::ns(
    level_upstream,
    df = 5
  ) +
  splines::ns(temperature_upstream, df = 5),
  lag_max = 10,
  fit_mean_x = fit_mean_x, fit_var_x = fit_var_x,
  fit_mean_y = fit_mean_y, fit_var_y = fit_var_y,
  df_correlation = c(5, 5)
)

```

---

conditional\_mean

*Estimating conditional mean of a time series*


---

### Description

This function estimates the means of a time series conditional on a set of other times series via additive models.

### Usage

```
conditional_mean(data, formula)
```

**Arguments**

data	a tibble containing all the time series which are uniquely identified by the corresponding Timestamp.
formula	A GAM formula. See <a href="#">formula.gam</a> . The details of model specification are given under 'Details'.

**Details**

Suppose  $x_t$  is a time series where its mean is a function of  $z_t$ . i.e.  $E(x_t|z_t) = m_x(z_t)$ . Then  $m_x(z_t)$  can be estimated via generalised additive models (GAM). This function uses GAMs implemented in `mgcv` package to estimate the conditional means of a time series given a set of time series predictors.

**Value**

The function returns an object of class "gam" as described in [gamObject](#).

**See Also**

[gam](#)

**Examples**

```
data <- NEON_PRIN_5min_cleaned |>
  dplyr::filter(site == "upstream") |>
  dplyr::select(Timestamp, turbidity, level, conductance, temperature)

fit_mean <- data |>
  conditional_mean(turbidity ~ s(level, k = 8) +
    s(conductance, k = 8) + s(temperature, k = 8))
```

---

conditional\_var

*Estimating conditional variance of a time series*

---

**Description**

This function estimates the variance of a time series conditional on a set of other times series via additive models.

**Usage**

```
conditional_var(data, formula, family = c("Gamma", "lognormal"), fit_mean)
```

**Arguments**

data	A tibble containing all the time series which are uniquely identified by the corresponding Timestamp.
formula	An object of class "formula": a symbolic description of the model to be fitted. The details of model specification are given under 'Details'.
family	the family to be used in conditional variance model. Currently this can take either "Gamma" or "lognormal".
fit_mean	A GAM object return from <a href="#">conditional_mean</a>

**Details**

Suppose  $x_t$  is a time series where its variance is a function of  $z_t$ . i.e.  $\text{Var}(x_t|z_t) = v_x(z_t)$ . Then  $v_x(z_t)$  can be estimated via generalised additive models (GAM). This function uses GAMs implemented in `mgcv` package to estimate the conditional variance of a time series given a set of time series predictors.

**Value**

The function returns an object of class "gam" as described in [gamObject](#).

**See Also**

[gam](#) and [ns](#).

**Examples**

```
data <- NEON_PRIN_5min_cleaned |>
  dplyr::filter(site == "upstream") |>
  dplyr::select(Timestamp, turbidity, level, conductance, temperature)

fit_mean <- data |>
  conditional_mean(turbidity ~ s(level, k = 8) +
    s(conductance, k = 8) + s(temperature, k = 8))
## Not run:
fit_var <- data |>
  conditional_var(
    turbidity ~ s(level, k = 7) + s(conductance, k = 7) + s(temperature, k = 7),
    family = "Gamma",
    fit_mean = fit_mean
  )

## End(Not run)
```

---

`conduits`*conduits: CONDitional User Interface for Time Series normalisation*

---

**Description**

Methods and tools for conditional normalisation of time series using additive models. This includes functions to estimate conditional means, conditional variances, conditional autocorrelation functions and conditional cross-correlation functions. Examples show these functions being used to estimate river flow time between two sensor locations in a river system.

**Author(s)**

Puwasala Gamakumara, Priyanga Dilini Talagala, Rob J Hyndman

---

`estimate_dt`*Estimating time delay between two sensors in a river system*

---

**Description**

This function estimates the time that takes water to flow from an upstream location to a downstream location conditional on the observed water-quality variables from the upstream sensor. That time lag is defined as the lag that gives maximum cross-correlation conditional on upstream water-quality variables.

**Usage**

```
estimate_dt(x)
```

**Arguments**

`x` Model object of class "conditional\_ccf" returned from `conditional_ccf`

**Value**

A `tibble` with estimated time lag "dt" and corresponding maximum cross-correlation

**Author(s)**

Puwasala Gamakumara & Priyanga Dilini Talagala

**Examples**

```

old_ts <- NEON_PRIN_5min_cleaned |>
  dplyr::select(
    Timestamp, site, turbidity, level, temperature
  ) |>
  tidyr::pivot_wider(
    names_from = site,
    values_from = turbidity:temperature
  )

fit_mean_y <- old_ts |>
  conditional_mean(turbidity_downstream ~
    s(level_upstream, k = 5) +
    s(temperature_upstream, k = 5))

fit_var_y <- old_ts |>
  conditional_var(
    turbidity_downstream ~
      s(level_upstream, k = 4) +
      s(temperature_upstream, k = 4),
    family = "Gamma",
    fit_mean = fit_mean_y
  )

fit_mean_x <- old_ts |>
  conditional_mean(turbidity_upstream ~
    s(level_upstream, k = 5) +
    s(temperature_upstream, k = 5))

fit_var_x <- old_ts |>
  conditional_var(
    turbidity_upstream ~
      s(level_upstream, k = 4) +
      s(temperature_upstream, k = 4),
    family = "Gamma",
    fit_mean = fit_mean_x
  )

fit_c_ccf <- old_ts |>
  tidyr::drop_na() |>
  conditional_ccf(
    I(turbidity_upstream * turbidity_downstream) ~
      splines::ns(level_upstream, df = 3) +
      splines::ns(temperature_upstream, df = 3),
    lag_max = 10,
    fit_mean_x = fit_mean_x, fit_var_x = fit_var_x,
    fit_mean_y = fit_mean_y, fit_var_y = fit_var_y,
    df_correlation = c(3, 3)
  )

new_data <- fit_c_ccf |> estimate_dt()

```

---

NEON\_PRIN\_5min\_cleaned

*Anomaly removed data for water quality variables aggregated at 5-minute intervals from Pringle Creek, Texas.*

---

### Description

NEON\_PRIN\_5min\_cleaned consists anomaly removed data for water quality variables from upstream and downstream sensors in Pringle Creek in Texas for the period spanning from 2019-07-01 to 2019-12-31 aggregated at 5-minute intervals.

### Usage

NEON\_PRIN\_5min\_cleaned

### Format

A data frame with water-quality variables, level and temperature data:

Timestamp Timestamp  
 site site position  
 conductance specific conductance  
 dissolvedOxygen dissolved oxygen  
 pH pH  
 chlorophyll chlorophyll  
 turbidity turbidity  
 fDOM fDOM  
 level elevation of surface water  
 temperature temperature in surface water

---

normalize

*Normalize a series using conditional moments*

---

### Description

This function produces a normalized series using conditional moments.

### Usage

normalize(data, y, fit\_mean, fit\_var)



**Arguments**

<code>data</code>	a tsibble containing all the time series which are uniquely identified by the corresponding Timestamp.
<code>y</code>	The variable name
<code>fit_mean</code>	Model object of class "conditional_moment" returned from <code>conditional_mean</code> with information to append to observations.
<code>fit_var</code>	Model object of class "conditional_moment" returned from <code>conditional_var</code> with information to append to observations.

**Value**

A vector of conditional normliased series

**Examples**

```
data <- NEON_PRIN_5min_cleaned |>
  dplyr::filter(site == "upstream") |>
  dplyr::select(Timestamp, turbidity, level, conductance, temperature) |>
  tsibble::as_tsibble(index = Timestamp)

fit_mean <- data |>
  conditional_mean(turbidity ~ s(level, k = 8) +
    s(conductance, k = 8) + s(temperature, k = 8))

fit_var <- data |>
  conditional_var(
    turbidity ~ s(level, k = 7) + s(conductance, k = 7) + s(temperature, k = 7),
    family = "Gamma",
    fit_mean = fit_mean
  )

new_ts <- data |>
  dplyr::mutate(ystar = conduits::normalize(data, turbidity, fit_mean, fit_var))
```

---

unnormalize

*Unnormalize a series using conditional moments*


---

**Description**

This function produces an unnormalized series using conditional moments.

**Usage**

```
unnormalize(data, ystar, fit_mean, fit_var)
```

**Arguments**

<code>data</code>	a tibble containing all the time series which are uniquely identified by the corresponding Timestamp.
<code>ystar</code>	The normalized variable name
<code>fit_mean</code>	Model object of class "conditional_moment" returned from <code>conditional_mean</code> with information to append to observations.
<code>fit_var</code>	Model object of class "conditional_moment" returned from <code>conditional_var</code> with information to append to observations.

**Value**

A `tibble` with the conditional normalized series

**Examples**

```
data <- NEON_PRIN_5min_cleaned |>
  dplyr::filter(site == "upstream") |>
  dplyr::select(Timestamp, turbidity, level, conductance, temperature) |>
  tibble::as_tibble(index = Timestamp)

fit_mean <- data |>
  conditional_mean(turbidity ~ s(level, k = 8) +
    s(conductance, k = 8) + s(temperature, k = 8))

fit_var <- data |>
  conditional_var(
    turbidity ~ s(level, k = 7) + s(conductance, k = 7) + s(temperature, k = 7),
    family = "Gamma",
    fit_mean = fit_mean
  )

new_ts <- data |>
  dplyr::mutate(ystar = normalize(data, turbidity, fit_mean, fit_var))

# For demonstrative purposes, declare three data points
# as missing values.
new_ts[3:5, 6] <- NA

## Not run:
library(fable)
library(dplyr)
impute_ts <- new_ts |>
  model(ARIMA(ystar)) |>
  interpolate(new_ts) |>
  rename(y_star_impt = ystar) |>
  full_join(new_ts, by = "Timestamp")
impute_ts <- impute_ts
  mutate(y = unnormalize(impute_ts, y_star_impt, fit_mean, fit_var))

## End(Not run)
```

# Index

## \* datasets

NEON\_PRIN\_5min\_cleaned, 16

augment.conditional\_acf, 2

augment.conditional\_ccf, 3

augment.conditional\_moment, 5

calc\_dt\_CI, 6

conditional\_acf, 2, 7

conditional\_ccf, 3, 6, 9, 14

conditional\_mean, 5, 8, 10, 11, 13, 17, 18

conditional\_var, 5, 8, 10, 12, 17, 18

conduits, 14

estimate\_dt, 14

formula.gam, 8, 9, 12

gam, 5, 12, 13

gamObject, 12, 13

glm, 8, 10

NEON\_PRIN\_5min\_cleaned, 16

normalize, 16

ns, 13

tibble, 2, 3, 5, 6, 14

tsibble, 18

unnormalize, 17