

Package ‘koopman.dmd’

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Type Package

Title Koopman Operator and Dynamic Mode Decomposition for Dynamical Systems

Version 0.2.2

Description Dynamic Mode Decomposition (DMD) with Koopman operator theory extensions, powered by a Rust backend via 'extendr'. Provides standard DMD as described in Schmid (2010) <[doi:10.1017/S0022112010001217](https://doi.org/10.1017/S0022112010001217)>, DMD with control for forced linear systems following Proctor, Brunton, and Kutz (2016) <[doi:10.1137/15M1013857](https://doi.org/10.1137/15M1013857)>, Extended DMD with lifting functions, Hankel-DMD via time-delay embedding, Generalized Laplace Analysis for direct eigenfunction computation, and harmonic time averages and mesochronic harmonic plots for phase space analysis as developed in Mezic (2020) <[doi:10.48550/arXiv.2009.05883](https://doi.org/10.48550/arXiv.2009.05883)>. Includes built-in area-preserving and chaotic maps for experimentation.

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Suggests testthat (>= 3.0.0), knitr, rmarkdown

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URL <https://github.com/jimeharrisjr/rust-dmd>,
<https://jimeharrisjr.github.io/rust-dmd/>

BugReports <https://github.com/jimeharrisjr/rust-dmd/issues>

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Value

Integer vector (1=resonating, 2=chaotic, 3=non-resonating).

Examples

```
# Classify orbits from their HTA magnitudes
mags <- c(0.5, 0.001, 1e-06, 0.1)
classify_phase_space(mags)

# Thresholds are adjustable
classify_phase_space(mags, resonating_threshold = 0.1, chaotic_threshold = 0.001)
```

dmd

Dynamic Mode Decomposition

Description

Perform Dynamic Mode Decomposition on time-series data.

Usage

```
dmd(
  X,
  rank = NULL,
  center = FALSE,
  dt = 1,
  lifting = NULL,
  lifting_param = NULL
)
```

Arguments

X	Numeric matrix (n_vars x n_time).
rank	Integer truncation rank, or NULL for automatic.
center	Logical, center data by subtracting row means.
dt	Numeric time step.
lifting	Character lifting type or NULL.
lifting_param	Integer lifting parameter or NULL.

Value

An S3 object of class "dmd".

Examples

```
# One row per variable, one column per time step
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))

d <- dmd(X, rank = 2, dt = t[2] - t[1])
print(d)
summary(d)

# Extended DMD: lift into a higher-dimensional space where nonlinear
# dynamics become approximately linear
d2 <- dmd(X, lifting = "polynomial", lifting_param = 2)
```

dmdc

Dynamic Mode Decomposition with Control (DMDc)

Description

Identify the forced linear system $x_{t+1} = Ax_t + Bu_t$ from snapshot pairs and control inputs, following Proctor, Brunton and Kutz (2016). Unlike [dmd](#), which takes one contiguous trajectory, `dmdc` takes explicit pair matrices: `X1` holds states at time t , `X2` the states one step later, and `U` the control input applied during each transition. Columns may therefore come from many concatenated trajectories.

Usage

```
dmdc(
  X1,
  X2,
  U = NULL,
  rank_input = NULL,
  rank_output = NULL,
  dt = 1,
  known_B = NULL
)
```

Arguments

<code>X1</code>	Numeric matrix of states at time t ($n_states \times n_pairs$).
<code>X2</code>	Numeric matrix of states at time $t+1$ ($n_states \times n_pairs$).
<code>U</code>	Numeric matrix of control inputs during each transition ($n_inputs \times n_pairs$), or <code>NULL</code> for an autonomous multi-trajectory fit. A vector is taken as a single input row.
<code>rank_input</code>	Integer truncation rank for the regression-input SVD, or <code>NULL</code> for automatic (99 percent cumulative variance).
<code>rank_output</code>	Integer rank of the output basis (SVD of <code>X2</code>), or <code>NULL</code> to keep everything full-order.

dt	Numeric time step between snapshot pairs.
known_B	Known input matrix B (n_states x n_inputs), or NULL to estimate B jointly with A.

Details

Two identification modes are available. With `known_B = NULL` (the default), A and B are solved jointly from the stacked regression $[A \ B] = X_2[X_1; U]^+$; this requires the input to be persistently exciting and exogenous (not state feedback). When the input coupling is known by construction, pass it as `known_B` and only A is estimated.

Value

An S3 object of class "dmdc" with components `a`, `b`, `a_tilde`, `b_tilde`, `basis`, `eigenvalues_re`, `eigenvalues_im`, `singular_values`, `rank_input`, `rank_output`, `dt`, `n_states`, and `n_inputs`.

References

Proctor, J. L., Brunton, S. L., and Kutz, J. N. (2016). Dynamic Mode Decomposition with Control. *SIAM Journal on Applied Dynamical Systems*, 15(1), 142-161. doi:10.1137/15M1013857

See Also

`dmd` for autonomous systems, `predict.dmdc` to simulate the identified system.

Examples

```
# Simulate  $x_{t+1} = A_0 x_t + B_0 u_t$  with a persistently exciting input
A0 <- matrix(c(0.9, 0, 0.1, 0.8), 2, 2)
B0 <- matrix(c(0.5, 1), 2, 1)
m <- 120
X1 <- matrix(0, 2, m)
X2 <- matrix(0, 2, m)
U <- matrix(0, 1, m)
x <- c(1, -0.5)
for (t in seq_len(m)) {
  u_t <- sin(0.7 * (t - 1)) + 0.5 * cos(2.3 * (t - 1) + 1)
  X1[, t] <- x
  U[, t] <- u_t
  x <- as.numeric(A0 %*% x + B0 * u_t)
  X2[, t] <- x
}

# Identify A and B jointly; both are recovered to machine precision
fit <- dmdc(X1, X2, U, rank_input = 3)
round(fit$a, 6)
round(fit$b, 6)

# Known B: pin the input matrix and estimate only A
fit2 <- dmdc(X1, X2, U, rank_input = 2, known_B = B0)
round(fit2$a, 6)
```

dmdc_spectrum	<i>DMDc spectrum analysis</i>
---------------	-------------------------------

Description

Per-mode frequency, growth rate and stability for the operator identified by `dmdc`. DMDc has no mode amplitudes, so the amplitude column is reported as 0.

Usage

```
dmdc_spectrum(object, dt = NULL)
```

Arguments

<code>object</code>	A dmdc object.
<code>dt</code>	Time step. Uses the stored dt by default.

Value

Data frame with mode information.

Examples

```
A0 <- matrix(c(0.9, 0, 0.1, 0.8), 2, 2)
B0 <- matrix(c(0.5, 1), 2, 1)
m <- 60
X1 <- matrix(0, 2, m); X2 <- matrix(0, 2, m); U <- matrix(0, 1, m)
x <- c(1, -0.5)
for (t in seq_len(m)) {
  u_t <- sin(0.7 * (t - 1)) + 0.5 * cos(2.3 * (t - 1) + 1)
  X1[, t] <- x
  U[, t] <- u_t
  x <- as.numeric(A0 %*% x + B0 * u_t)
  X2[, t] <- x
}
fit <- dmdc(X1, X2, U, rank_input = 3)

dmdc_spectrum(fit)
```

dmdc_stability	<i>DMDc stability analysis</i>
----------------	--------------------------------

Description

Classify the stability of the operator identified by `dmdc` from the eigenvalues of $\tilde{A}$.

Usage

```
dmdc_stability(object, tol = 1e-06)
```

Arguments

<code>object</code>	A <code>dmdc</code> object.
<code>tol</code>	Tolerance for marginal classification.

Value

List with stability information (`is_stable`, `is_unstable`, `is_marginal`, `spectral_radius`).

Examples

```
A0 <- matrix(c(0.9, 0, 0.1, 0.8), 2, 2)
B0 <- matrix(c(0.5, 1), 2, 1)
m <- 60
X1 <- matrix(0, 2, m); X2 <- matrix(0, 2, m); U <- matrix(0, 1, m)
x <- c(1, -0.5)
for (t in seq_len(m)) {
  u_t <- sin(0.7 * (t - 1)) + 0.5 * cos(2.3 * (t - 1) + 1)
  X1[, t] <- x
  U[, t] <- u_t
  x <- as.numeric(A0 %*% x + B0 * u_t)
  X2[, t] <- x
}
fit <- dmdc(X1, X2, U, rank_input = 3)

dmdc_stability(fit)
```

dmd_dominant_modes	<i>DMD dominant modes</i>
--------------------	---------------------------

Description

DMD dominant modes

Usage

```
dmd_dominant_modes(
  object,
  n = 3,
  criterion = c("amplitude", "energy", "stability")
)
```

Arguments

object	A dmd object.
n	Number of modes.
criterion	"amplitude", "energy", or "stability".

Value

Integer vector of 1-based mode indices.

Examples

```
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))
d <- dmd(X, rank = 2, dt = t[2] - t[1])

# Indices of the most significant modes
dmd_dominant_modes(d, n = 1)
```

dmd_error	<i>DMD reconstruction error</i>
-----------	---------------------------------

Description

DMD reconstruction error

Usage

```
dmd_error(object)
```

Arguments

object	A dmd object.
--------	---------------

Value

List with error metrics (rmse, mae, mape, relative_error).

Examples

```
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))
d <- dmd(X, rank = 2, dt = t[2] - t[1])

# Reconstruction error metrics
dmd_error(d)
```

dmd_reconstruct

Reconstruct data from DMD modes

Description

Reconstruct data from DMD modes

Usage

```
dmd_reconstruct(object, n_steps = NULL, modes = NULL)
```

Arguments

object	A dmd object.
n_steps	Number of time steps. Defaults to original length.
modes	Integer vector of mode indices (1-based), or NULL for all.

Value

Numeric matrix.

Examples

```
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))
d <- dmd(X, rank = 2, dt = t[2] - t[1])

recon <- dmd_reconstruct(d, n_steps = ncol(X))
dim(recon)
```

dmd_residual	<i>DMD residual analysis</i>
--------------	------------------------------

Description

DMD residual analysis

Usage

```
dmd_residual(object)
```

Arguments

object A dmd object.

Value

List with residual_norm and residual_relative.

Examples

```
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))
d <- dmd(X, rank = 2, dt = t[2] - t[1])

dmd_residual(d)
```

dmd_spectrum	<i>DMD spectrum analysis</i>
--------------	------------------------------

Description

DMD spectrum analysis

Usage

```
dmd_spectrum(object, dt = NULL)
```

Arguments

object A dmd object.
dt Time step. Uses stored dt by default.

Value

Data frame with mode information.

Examples

```
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))
d <- dmd(X, rank = 2, dt = t[2] - t[1])

# Frequency, growth rate, amplitude and stability of each mode
dmd_spectrum(d)
```

dmd_stability	<i>DMD stability analysis</i>
---------------	-------------------------------

Description

DMD stability analysis

Usage

```
dmd_stability(object, tol = 1e-06)
```

Arguments

object	A dmd object.
tol	Tolerance for marginal classification.

Value

List with stability information.

Examples

```
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))
d <- dmd(X, rank = 2, dt = t[2] - t[1])

dmd_stability(d)
```

extended_standard_map *Extended standard map (3D)*

Description

Extended standard map (3D)

Usage

```
extended_standard_map(state, epsilon = 0.01, delta = 0.001)
```

Arguments

state	Numeric vector of length 3.
epsilon	Perturbation parameter.
delta	Coupling parameter.

Value

Updated state vector.

Examples

```
extended_standard_map(c(0.1, 0.2, 0.3))
```

froeschle_map *Froeschle map (4D coupled standard maps)*

Description

Froeschle map (4D coupled standard maps)

Usage

```
froeschle_map(state, epsilon1 = 0.02, epsilon2 = 0.02, eta = 0.01)
```

Arguments

state	Numeric vector of length 4.
epsilon1	First perturbation.
epsilon2	Second perturbation.
eta	Coupling parameter.

Value

Updated state vector.

Examples

```
froeschle_map(c(0.1, 0.2, 0.3, 0.4))
```

generate_trajectory	<i>Generate a trajectory from a built-in map</i>
---------------------	--

Description

Generate a trajectory from a built-in map

Usage

```
generate_trajectory(map_name, initial_condition, n_iter, ...)
```

Arguments

map_name	Character: "standard", "froeschle", "extended_standard", "henon", or "logistic".
initial_condition	Numeric vector.
n_iter	Integer number of iterations.
...	Map parameters passed as named arguments.

Value

Numeric matrix (n_dim x n_iter+1).

Examples

```
# Chirikov standard map with default parameters
traj <- generate_trajectory("standard", c(0.1, 0.2), 100)
dim(traj)

# Map parameters are passed through `...`
henon <- generate_trajectory("henon", c(0, 0), 100, a = 1.4, b = 0.3)
logistic <- generate_trajectory("logistic", 0.5, 100, r = 3.9)
```

gla	<i>Generalized Laplace Analysis</i>
-----	-------------------------------------

Description

Generalized Laplace Analysis

Usage

```
gla(y, eigenvalues = NULL, n_eigenvalues = 5, tol = 1e-06, max_iter = NULL)
```

Arguments

y	Numeric matrix (n_obs x n_time).
eigenvalues	Complex vector of known eigenvalues, or NULL.
n_eigenvalues	Number of eigenvalues to estimate.
tol	Convergence tolerance.
max_iter	Maximum iterations, or NULL.

Value

An S3 object of class "gla".

Examples

```
t <- seq(0, 10, length.out = 200)
y <- rbind(sin(t), cos(t))

g <- gla(y, n_eigenvalues = 2)
print(g)
```

gla_reconstruct	<i>Reconstruct from GLA</i>
-----------------	-----------------------------

Description

Reconstruct from GLA

Usage

```
gla_reconstruct(object, modes_to_use = NULL)
```

Arguments

object	A gla object.
modes_to_use	Integer vector of mode indices (1-based), or NULL.

Value

Numeric matrix.

Examples

```
t <- seq(0, 10, length.out = 200)
y <- rbind(sin(t), cos(t))
g <- gla(y, n_eigenvalues = 2)

recon <- gla_reconstruct(g)
dim(recon)
```

hankel_dmd

Hankel-DMD (Time-Delay Embedding DMD)

Description

Hankel-DMD (Time-Delay Embedding DMD)

Usage

```
hankel_dmd(y, delays = NULL, rank = NULL, dt = 1)
```

Arguments

y	Numeric matrix (n_obs x n_time).
delays	Integer number of delays, or NULL for automatic.
rank	Integer truncation rank, or NULL for automatic.
dt	Numeric time step.

Value

An S3 object of class "hankel_dmd".

Examples

```
# A scalar signal, as a 1-row matrix
t <- seq(0, 4 * pi, length.out = 200)
y <- matrix(sin(t), nrow = 1)

h <- hankel_dmd(y, delays = 10)
print(h)
```

hankel_reconstruct	<i>Reconstruct from Hankel-DMD</i>
--------------------	------------------------------------

Description

Reconstruct from Hankel-DMD

Usage

```
hankel_reconstruct(object, n_steps)
```

Arguments

object	A hankel_dmd object.
n_steps	Number of time steps.

Value

Numeric matrix.

Examples

```
t <- seq(0, 4 * pi, length.out = 200)
y <- matrix(sin(t), nrow = 1)
h <- hankel_dmd(y, delays = 10)

recon <- hankel_reconstruct(h, 50)
```

harmonic_time_average	<i>Harmonic Time Average</i>
-----------------------	------------------------------

Description

Harmonic Time Average

Usage

```
harmonic_time_average(
  map_name,
  initial_condition,
  observable = "sin_pi",
  omega = 0.1,
  n_iter = 10000,
  ...
)
```

Arguments

map_name	Character map name.
initial_condition	Numeric vector.
observable	Character observable name.
omega	Numeric frequency.
n_iter	Integer iterations.
...	Map parameters.

Value

List with magnitude, phase, hta_re, hta_im.

Examples

```
# Harmonic time average of an orbit of the Chirikov standard map
harmonic_time_average("standard", c(0.1, 0.2), "sin_pi", 0.1, 500)
```

henon_map	<i>Henon map (2D dissipative)</i>
-----------	-----------------------------------

Description

Henon map (2D dissipative)

Usage

```
henon_map(state, a = 1.4, b = 0.3)
```

Arguments

state	Numeric vector c(x, y).
a	Parameter a.
b	Parameter b.

Value

Updated state vector.

Examples

```
henon_map(c(0, 0))

# Classic chaotic parameters
henon_map(c(0, 0), a = 1.4, b = 0.3)
```

hta_convergence	<i>HTA convergence analysis</i>
-----------------	---------------------------------

Description

HTA convergence analysis

Usage

```
hta_convergence(  
  map_name,  
  initial_condition,  
  observable = "sin_pi",  
  omega = 0.1,  
  n_iter = 10000,  
  ...  
)
```

Arguments

map_name	Character map name.
initial_condition	Numeric vector.
observable	Character observable name.
omega	Numeric frequency.
n_iter	Integer iterations.
...	Map parameters.

Value

List with times, hta_magnitudes, convergence_rate, dynamics_type.

Examples

```
# How the time average converges along the orbit  
conv <- hta_convergence("standard", c(0.1, 0.2), "sin_pi", 0.1, 500)  
str(conv)
```

logistic_map	<i>Logistic map (1D)</i>
--------------	--------------------------

Description

Logistic map (1D)

Usage

```
logistic_map(state, r = 3.9)
```

Arguments

state	Numeric scalar.
r	Growth rate parameter.

Value

Updated state value.

Examples

```
logistic_map(0.5)

# In the chaotic regime
logistic_map(0.5, r = 3.9)
```

mesochronic_compute	<i>Mesochronic harmonic plot computation</i>
---------------------	--

Description

Mesochronic harmonic plot computation

Usage

```
mesochronic_compute(
  map_name,
  x_range = c(0, 1),
  y_range = c(0, 1),
  resolution = 100,
  observable = "sin_pi",
  omega = 0.1,
  n_iter = 30000,
  ...
)
```

Arguments

map_name	Character map name.
x_range	Numeric vector c(min, max).
y_range	Numeric vector c(min, max).
resolution	Integer grid resolution.
observable	Character observable name.
omega	Numeric frequency.
n_iter	Integer iterations.
...	Map parameters.

Value

List with hta_matrix, phase_matrix, x_coords, y_coords.

Examples

```
# Mesochronic plot over a coarse grid. Raise `resolution` and `n_iter`
# for publication-quality figures; both cost time roughly linearly.
mhp <- mesochronic_compute("standard", c(0, 1), c(0, 1), 10, "sin_pi", 0.1, 100)
str(mhp)
```

predict.dmd	<i>Predict from DMD model</i>
-------------	-------------------------------

Description

Predict from DMD model

Usage

```
## S3 method for class 'dmd'
predict(object, n_ahead = 10, x0 = NULL, method = c("modes", "matrix"), ...)
```

Arguments

object	A dmd object.
n_ahead	Number of steps to predict.
x0	Optional initial condition vector.
method	"modes" (default) or "matrix".
...	Additional arguments (ignored).

Value

Numeric matrix of predictions.

Examples

```
t <- seq(0, 10, length.out = 100)
X <- rbind(sin(t), cos(t))
d <- dmd(X, rank = 2, dt = t[2] - t[1])

# Forecast 10 steps beyond the input
pred <- predict(d, n_ahead = 10)
dim(pred)
```

predict.dmdc

Predict from a DMDc model

Description

Simulate the identified system $x_{t+1} = Ax_t + Bu_t$ forward from an initial state under a given control input sequence.

Usage

```
## S3 method for class 'dmdc'
predict(object, U = NULL, x0 = NULL, n_ahead = NULL, ...)
```

Arguments

object	A dmdc object.
U	Numeric matrix of control inputs (n_inputs x n_steps); the number of columns sets the prediction horizon. A vector is taken as a single input row. NULL applies zero input for n_ahead steps.
x0	Initial state vector. Defaults to the first stored snapshot.
n_ahead	Number of steps when U is NULL; if both are given it must match ncol(U).
...	Additional arguments (ignored).

Value

Numeric matrix of predicted states $x_1 \dots x_k$ (n_states x k).

Examples

```
A0 <- matrix(c(0.9, 0, 0.1, 0.8), 2, 2)
B0 <- matrix(c(0.5, 1), 2, 1)
m <- 120
X1 <- matrix(0, 2, m)
X2 <- matrix(0, 2, m)
U <- matrix(0, 1, m)
x <- c(1, -0.5)
for (t in seq_len(m)) {
  u_t <- sin(0.7 * (t - 1)) + 0.5 * cos(2.3 * (t - 1) + 1)
```

```

X1[, t] <- x
U[, t] <- u_t
x <- as.numeric(A0 %% x + B0 * u_t)
X2[, t] <- x
}
fit <- dmdc(X1, X2, U, rank_input = 3)

# Replaying the training inputs reproduces the observed successors
pred <- predict(fit, U = U)
max(abs(pred - X2))

# Zero-input (free) response from a chosen state
free <- predict(fit, x0 = c(1, 1), n_ahead = 10)
dim(free)

```

predict.gla

*Predict from GLA***Description**

Predict from GLA

Usage

```

## S3 method for class 'gla'
predict(object, n_ahead = 10, ...)

```

Arguments

object	A gla object.
n_ahead	Number of steps to predict.
...	Additional arguments (ignored).

Value

Numeric matrix.

Examples

```

t <- seq(0, 10, length.out = 200)
y <- rbind(sin(t), cos(t))
g <- gla(y, n_eigenvalues = 2)

pred <- predict(g, n_ahead = 5)

```

predict.hankel_dmd	<i>Predict from Hankel-DMD</i>
--------------------	--------------------------------

Description

Predict from Hankel-DMD

Usage

```
## S3 method for class 'hankel_dmd'
predict(object, n_ahead = 10, ...)
```

Arguments

object	A hankel_dmd object.
n_ahead	Number of steps to predict.
...	Additional arguments (ignored).

Value

Numeric matrix.

Examples

```
t <- seq(0, 4 * pi, length.out = 200)
y <- matrix(sin(t), nrow = 1)
h <- hankel_dmd(y, delays = 10)

pred <- predict(h, n_ahead = 10)
```

standard_map	<i>Standard map (Chirikov)</i>
--------------	--------------------------------

Description

Standard map (Chirikov)

Usage

```
standard_map(state, epsilon = 0.12)
```

Arguments

state	Numeric vector c(x, y).
epsilon	Perturbation parameter.

Value

Updated state vector.

Examples

```
# One iteration from a given state
standard_map(c(0.1, 0.2))

# Stronger nonlinearity
standard_map(c(0.1, 0.2), epsilon = 0.5)
```

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