

Statistical Learning for Process Data

Session 3: Introduction to ProcData Package

Xueying Tang

Department of Mathematics
University of Arizona

July 19, 2021

Agenda

- ▶ Overview of ProcData package

Agenda

- ▶ Overview of ProcData package
- ▶ Installation

Agenda

- ▶ Overview of ProcData package
- ▶ Installation
- ▶ Data input/output, structure, and processing
- ▶ Feature extraction
- ▶ Sequence models

ProcData Overview

- ▶ ProcData provides tools for processing and analyzing process data.

ProcData Overview

- ▶ ProcData provides tools for processing and analyzing process data. It includes
 - ▶ an S3 class "proc" for response processes

ProcData Overview

- ▶ ProcData provides tools for processing and analyzing process data. It includes
 - ▶ an S3 class "proc" for response processes
 - ▶ functions for data processing

ProcData Overview

- ▶ ProcData provides tools for processing and analyzing process data. It includes
 - ▶ an S3 class "proc" for response processes
 - ▶ functions for data processing
 - ▶ state-of-the-art methods for process data analysis

ProcData Overview

- ▶ ProcData provides tools for processing and analyzing process data. It includes
 - ▶ an S3 class "proc" for response processes
 - ▶ functions for data processing
 - ▶ state-of-the-art methods for process data analysis
 - ▶ feature extraction methods
 - ▶ sequence models

ProcData Overview

Category	Functions
Data input/output	<code>read.seqs</code> , <code>write.seqs</code>
Data structure	<code>proc</code> , <code>print.proc</code> , <code>summary.proc</code> , <code>cc_data</code>
Data processing	<code>remove_repeat</code> , <code>remove_action</code> , <code>replace_action</code> , <code>combine_actions</code>
Feature extraction	<code>seq2feature_mds</code> , <code>chooseK_mds</code> , <code>seq2feature_seq2seq</code> [†] , <code>chooseK_seq2seq</code> [†]
Sequence model	<code>seqm</code> [†] , <code>predict.seqm</code> [†]

Table: Summary of ProcData. Functions marked by [†] depend on keras.

Installation

1. Install ProcData and dependent R packages from CRAN

```
install.packages("ProcData", dependencies=T)
```

or from GitHub:

```
devtools::install_github("xytangtang/ProcData",  
                           dependencies=T)
```

Installation

1. Install ProcData and dependent R packages from CRAN

```
install.packages("ProcData", dependencies=T)
```

or from GitHub:

```
devtools::install_github("xytangtang/ProcData",  
                           dependencies=T)
```

2. Install dependent Python libraries

```
library(keras)  
install_keras()
```

Installation

1. Install ProcData and dependent R packages from CRAN

```
install.packages("ProcData", dependencies=T)
```

or from GitHub:

```
devtools::install_github("xytangtang/ProcData",  
                           dependencies=T)
```

2. Install dependent Python libraries

```
library(keras)  
install_keras()
```

- ▶ If this step is skipped or fails, calling functions marked by † in the table on Page 4 will lead to an error.
- ▶ Getting help for installation:
<https://github.com/xytangtang/ProcData/issues>

Data Input/Output

File Style "single"

	A	B	C	D	E	F
1	Country	Gender	Age	ID	Action	Time
2	US	F	18	101	Start,CHECK_A,End	0,6803,8774
3	US	M	35	102	Start,OPT1_1,OPT2_1,RUN,CHECK_D,End	0,18263,21010,22034,42015,44132
4	US	F	31	103	Start,OPT1_3,OPT2_2,RUN,CHECK_A,End	0,71910,75087,81733,105105,129596
5	JP	M	22	104	Start,OPT1_1,OPT2_2,RUN,OPT1_3,OPT2_2	0,81733,95466,98860,105105,106560,1110
6	JP	F	40	105	Start,CHECK_C,End	0,3325,3568
7						

File Style "multiple"

	A	B	C	D	E	F
1	Country	Gender	Age	ID	Action	Time
2	US	F	18	101	Start	0
3	US	F	18	101	CHECK_A	6803
4	US	F	18	101	End	8774
5	US	M	35	102	Start	0
6	US	M	35	102	OPT1_1	18263
7	US	M	35	102	OPT2_1	21010
8	US	M	35	102	RUN	22034
9	US	M	35	102	CHECK_D	42015
10	US	M	35	102	End	44132

Data Input/Output

- ▶ Two file styles: single or multiple

Data Input/Output

- ▶ Two file styles: single or multiple
- ▶ `read.seqs`: read process data from a CSV file as a `proc` object

Data Input/Output

- ▶ Two file styles: single or multiple
- ▶ `read.seqs`: read process data from a CSV file as a `proc` object
- ▶ `write.seqs`: write a `proc` object to a CSV file.

Data Input/Output

- ▶ Two file styles: single or multiple
- ▶ `read.seqs`: read process data from a CSV file as a proc object
- ▶ `write.seqs`: write a proc object to a CSV file.

Code demonstration: `chuck 1` in `procdemo_demo.R`

Climate Control Dataset

ProcData contains a dataset of PISA climate control item.

CLIMATE CONTROL

You have no instructions for your new air conditioner. You need to work out how to use it.

You can change the top, central and bottom controls on the left by using the sliders (→). The initial setting for each control is indicated by ▲.

By clicking APPLY, you will see any changes in the temperature and humidity of the room in the temperature and humidity graphs. The box to the left of each graph shows the current level of temperature or humidity.

The interface shows three sliders for 'Top Control', 'Central Control', and 'Bottom Control'. Each slider has a scale from -- to ++ and a blue triangle marker. The 'Top Control' marker is at the first tick, 'Central Control' is at the second, and 'Bottom Control' is at the third. To the right of the sliders are two graphs: 'Temperature' and 'Humidity'. Both graphs show a grey box with the number '25' inside. Below the sliders are two buttons: 'APPLY' and 'RESET'. At the bottom center is an illustration of a white air conditioner unit.

Question : CLIMATE CONTROL

Find whether each control influences temperature and humidity by changing the sliders. You can start again by clicking RESET.

Draw lines in the diagram on the right to show what each control influences.

To draw a line, click on a control and then click on either Temperature or Humidity. You can remove any line by clicking on it.



Climate Control Dataset

ProcData contains a dataset of PISA climate control item.

- ▶ Item interface: <https://www.oecd.org/pisa/test-2012/testquestions/question3/>

Climate Control Dataset

ProcData contains a dataset of PISA climate control item.

- ▶ Item interface: `https://www.oecd.org/pisa/test-2012/testquestions/question3/`
- ▶ Load dataset in R
`data(cc_data)`

Climate Control Dataset

ProcData contains a dataset of PISA climate control item.

- ▶ Item interface: `https://www.oecd.org/pisa/test-2012/testquestions/question3/`
- ▶ Load dataset in R
`data(cc_data)`
- ▶ `cc_data` is a list of two elements
 - ▶ `seqs`: a proc object of 16,763 processes
 - ▶ `responses`: binary item responses

Code demonstration: chunk 2 in `procdemo_demo.R`

Data Structure

Response process:

Action:	Start,	Click_CS,	...	Next,	Next_OK
Time:	0.0,	2.9,	...	60.4,	62.2

- ▶ S3 class `proc` is defined in `ProcData` for response processes.

Data Structure

Response process:

Action:	Start,	Click_CS,	...	Next,	Next_OK
Time:	0.0,	2.9,	...	60.4,	62.2

- ▶ S3 class `proc` is defined in `ProcData` for response processes.
- ▶ A `proc` object of n processes is a list of two elements.
 - ▶ `action_seqs`: a list of n action sequences (character vectors).
 - ▶ `time_seqs`: a list of n time sequences (numeric vectors).
- ▶ Names of `action_seqs` and `time_seqs` (if not `NULL`) are IDs of respondents.

Data Structure – Methods for `proc`

- ▶ `print.proc` prints response processes in an easy-to-read format

Data Structure – Methods for proc

- ▶ `print.proc` prints response processes in an easy-to-read format
- ▶ `summary.proc` gives a list of quantities summarized from a set of response processes.
 - ▶ number of processes
 - ▶ number of distinct actions
 - ▶ action set
 - ▶ total response time
 - ▶ ...

Data Structure – Methods for `proc`

- ▶ `print.proc` prints response processes in an easy-to-read format
- ▶ `summary.proc` gives a list of quantities summarized from a set of response processes.
 - ▶ number of processes
 - ▶ number of distinct actions
 - ▶ action set
 - ▶ total response time
 - ▶ ...

Code demonstration: `chuck 3` in `procdemo_demo.R`

Data Processing

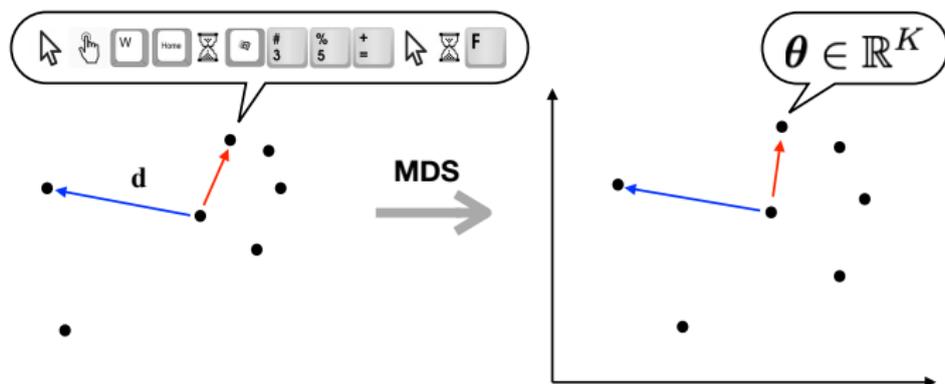
- ▶ `sub_seqs`: subset proc objects
- ▶ `remove_repeat`: remove consecutive repeated actions
- ▶ `remove_action`: remove a set of actions
- ▶ `replace_action`: rename an action
- ▶ `combine_actions`: combine several consecutive actions into a single action

Code demonstration: chunk 4 in `procdemo.R`

Feature Extraction

- ▶ Goal: transform a response process s into $\theta \in \mathbb{R}^K$ without losing too much information
- ▶ Two methods:
 - ▶ Multidimensional Scaling
 - ▶ Sequence-to-sequence autoencoder

Feature Extraction – Multidimensional Scaling



Feature Extraction – Multidimensional Scaling

- ▶ $d_{ij} = d(\mathbf{s}_i, \mathbf{s}_j)$: dissimilarity between action sequences \mathbf{s}_i and \mathbf{s}_j .

Feature Extraction – Multidimensional Scaling

▶ $d_{ij} = d(\mathbf{s}_i, \mathbf{s}_j)$: dissimilarity between action sequences \mathbf{s}_i and \mathbf{s}_j .

▶ Minimize

$$F(\boldsymbol{\theta}_1, \dots, \boldsymbol{\theta}_n) = \sum_{i < j} (d_{ij} - \|\boldsymbol{\theta}_i - \boldsymbol{\theta}_j\|)^2$$

▶ $\|\boldsymbol{\theta}_i - \boldsymbol{\theta}_j\|$: distance between latent features $\boldsymbol{\theta}_i$ and $\boldsymbol{\theta}_j$.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
               dist_type = "oss_action", pca = TRUE,  
               return_dist = FALSE, ...)
```

- ▶ `seq2feature_mds` extracts MDS features from response processes.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
               dist_type = "oss_action", pca = TRUE,  
               return_dist = FALSE, ...)
```

- ▶ `seq2feature_mds` extracts MDS features from response processes.
- ▶ `seqs` can be a proc object or a dissimilarity matrix.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ `seq2feature_mds` extracts MDS features from response processes.
- ▶ `seqs` can be a proc object or a dissimilarity matrix.
- ▶ `K` specifies feature dimension.
 - ▶ use `chooseK_mds` to select the dimension by cross-validation.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ `seq2feature_mds` extracts MDS features from response processes.
- ▶ `seqs` can be a `proc` object or a dissimilarity matrix.
- ▶ `K` specifies feature dimension.
 - ▶ use `chooseK_mds` to select the dimension by cross-validation.
- ▶ If `pca=TRUE`, principal components of features are returned.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ `seq2feature_mds` extracts MDS features from response processes.
- ▶ `seqs` can be a `proc` object or a dissimilarity matrix.
- ▶ `K` specifies feature dimension.
 - ▶ use `chooseK_mds` to select the dimension by cross-validation.
- ▶ If `pca=TRUE`, principal components of features are returned.
- ▶ If `return_dist=TRUE`, the full dissimilarity matrix is returned.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ Dissimilarity measure: Order-based Symbol Similarity (OSS; Gómez-Alonso and Valls, 2008)
- ▶ Customized dissimilarity measures can be used by computing the dissimilarity matrix outside `seq2feature_mds` and then passing the matrix to the function through `seqs`.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ Optimization method is specified via `method`.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ Optimization method is specified via `method`.
- ▶ `method="small"`: classical MDS via `cmdscale`.
 - ▶ Requires large memory if sample size is large.

Feature Extraction – Multidimensional Scaling

```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ Optimization method is specified via `method`.
- ▶ `method="small"`: classical MDS via `cmdscale`.
 - ▶ Requires large memory if sample size is large.
- ▶ `method="large"`: an algorithm for large datasets (Paradis 2018)

Feature Extraction – Multidimensional Scaling

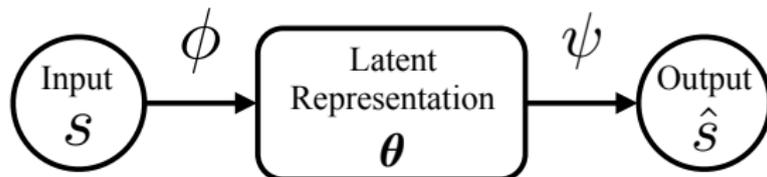
```
seq2feature_mds(seqs = NULL, K = 2, method = "auto",  
                dist_type = "oss_action", pca = TRUE,  
                return_dist = FALSE, ...)
```

- ▶ Optimization method is specified via `method`.
- ▶ `method="small"`: classical MDS via `cmdscale`.
 - ▶ Requires large memory if sample size is large.
- ▶ `method="large"`: an algorithm for large datasets (Paradis 2018)
- ▶ `method="auto"`: use `method="small"` if sample size < 5000 and `method="large"` otherwise.

Code demonstration: chunk 5 in `procdemo_demo.R`

Feature Extraction – Seq2Seq Autoencoder

Autoencoder: neural network that reproduces input in its output

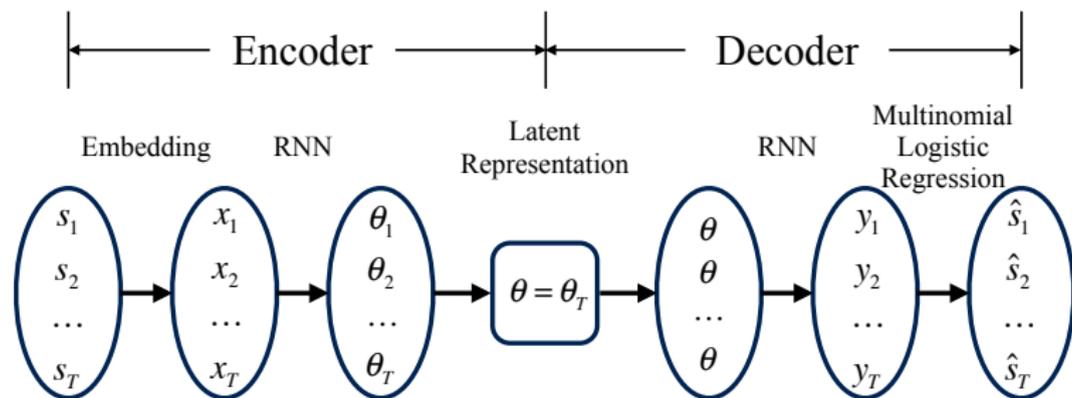


- ▶ $\phi(\cdot; \eta)$: encoder; $\psi(\cdot; \xi)$: decoder; features: $\theta = \phi(s; \eta)$
- ▶ Parameters are estimated by minimizing the average difference between \hat{s} and s

$$F(\eta, \xi) = \frac{1}{n} \sum_{i=1}^n L(s_i, \hat{s}_i)$$

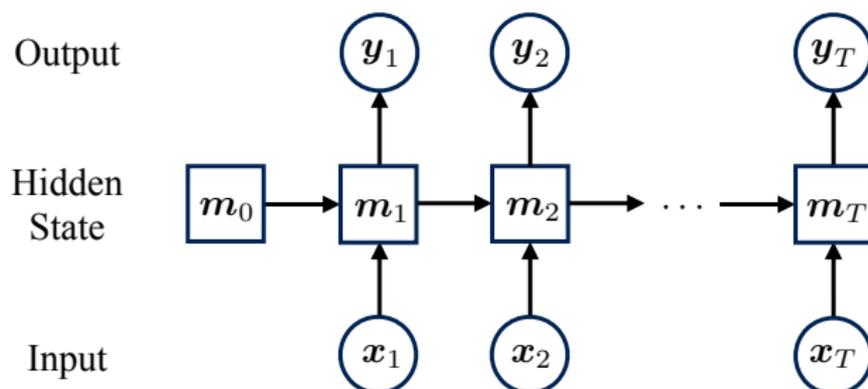
Feature Extraction – Seq2Seq Autoencoder

Action sequence autoencoder



Feature Extraction – Seq2Seq Autoencoder

Recurrent neural network



$$m_t = f(x_t, m_{t-1})$$

$$y_t = g(m_t)$$

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
  rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
  step_size = 1e-04, optimizer_name = "adam",  
  samples_train, samples_valid, ...)
```

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
  rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
  step_size = 1e-04, optimizer_name = "adam",  
  samples_train, samples_valid, ...)
```

- ▶ seq2feature_seq2seq extracts **K** features from response processes given in the proc object **seqs**.
 - ▶ K can be selected by chooseK_seq2seq.

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
  rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
  step_size = 1e-04, optimizer_name = "adam",  
  samples_train, samples_valid, ...)
```

- ▶ `seq2feature_seq2seq` extracts K features from response processes given in the proc object `seqs`.
 - ▶ K can be selected by `chooseK_seq2seq`.
- ▶ `rnn_type` specifies the structure of the encoder RNN and the decoder RNN ("lstm" or "gru").

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
  rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
  step_size = 1e-04, optimizer_name = "adam",  
  samples_train, samples_valid, ...)
```

- ▶ `seq2feature_seq2seq` extracts K features from response processes given in the proc object `seqs`.
 - ▶ K can be selected by `chooseK_seq2seq`.
- ▶ `rnn_type` specifies the structure of the encoder RNN and the decoder RNN ("lstm" or "gru").
- ▶ If `pca=TRUE`, principal components of extracted features are returned.

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
    rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
    step_size = 1e-04, optimizer_name = "adam",  
    samples_train, samples_valid, ...)
```

- ▶ The minimizer of the objective function $F(\eta, \xi)$ is obtained by stochastic approximation.

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
    rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
    step_size = 1e-04, optimizer_name = "adam",  
    samples_train, samples_valid, ...)
```

- ▶ The minimizer of the objective function $F(\eta, \xi)$ is obtained by stochastic approximation.
- ▶ `n_epoch` specifies the number of epochs to run.

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
    rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
    step_size = 1e-04, optimizer_name = "adam",  
    samples_train, samples_valid, ...)
```

- ▶ The minimizer of the objective function $F(\eta, \xi)$ is obtained by stochastic approximation.
- ▶ `n_epoch` specifies the number of epochs to run.
- ▶ `step_size` specifies the (baseline) step size.

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
    rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
    step_size = 1e-04, optimizer_name = "adam",  
    samples_train, samples_valid, ...)
```

- ▶ The minimizer of the objective function $F(\eta, \xi)$ is obtained by stochastic approximation.
- ▶ `n_epoch` specifies the number of epochs to run.
- ▶ `step_size` specifies the (baseline) step size.
- ▶ `optimizer_name` specifies the algorithms.
 - ▶ Available options are "sgd", "rmsprop", "adadelata", and "adam".

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
                    rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
                    step_size = 1e-04, optimizer_name = "adam",  
                    samples_train, samples_valid, ...)
```

- ▶ Validation-based early stopping is used to prevent overfitting.

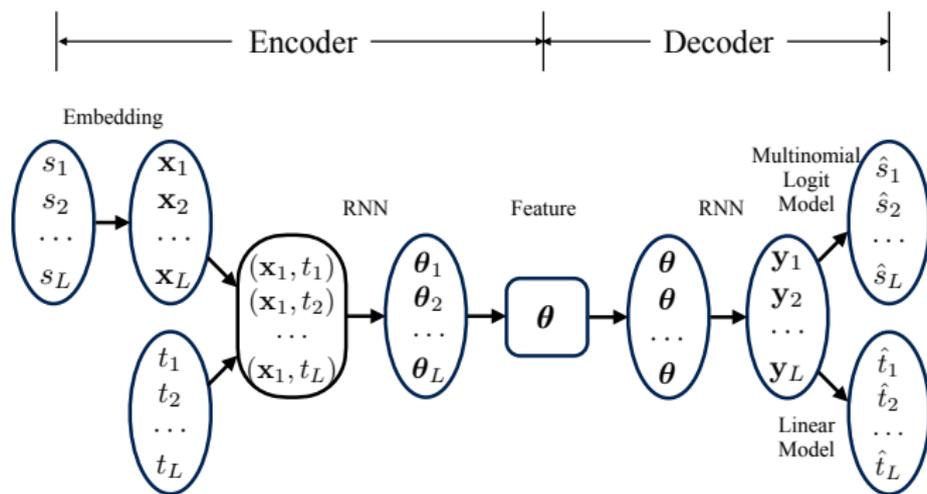
Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type = "action",  
                    rnn_type = "lstm", pca = TRUE, n_epoch = 50,  
                    step_size = 1e-04, optimizer_name = "adam",  
                    samples_train, samples_valid, ...)
```

- ▶ Validation-based early stopping is used to prevent overfitting.
- ▶ `samples_train` and `samples_valid` specifies the indices of observations in the training and validation set.

Feature Extraction – Seq2Seq Autoencoder

Action-time sequence autoencoder



Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type, rnn_type,  
    n_epoch, step_size, optimizer_name,  
    cumulative=FALSE, log=TRUE, weights=c(1,0.5),  
    samples_train, samples_valid, ...)
```

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type, rnn_type,  
    n_epoch, step_size, optimizer_name,  
    cumulative=FALSE, log=TRUE, weights=c(1,0.5),  
    samples_train, samples_valid, ...)
```

▶ `ae_type="both"`

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type, rnn_type,  
                    n_epoch, step_size, optimizer_name,  
                    cumulative=FALSE, log=TRUE, weights=c(1,0.5),  
                    samples_train, samples_valid, ...)
```

- ▶ `ae_type="both"`
- ▶ If `cumulative=FALSE`, time between two consecutive actions will be used instead of the timestamp sequence.

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type, rnn_type,  
    n_epoch, step_size, optimizer_name,  
    cumulative=FALSE, log=TRUE, weights=c(1,0.5),  
    samples_train, samples_valid, ...)
```

- ▶ `ae_type="both"`
- ▶ If `cumulative=FALSE`, time between two consecutive actions will be used instead of the timestamp sequence.
- ▶ If `log=TRUE`, the logarithm of the time sequence will be used.

Feature Extraction – Seq2Seq Autoencoder

```
seq2feature_seq2seq(seqs, K, ae_type, rnn_type,  
    n_epoch, step_size, optimizer_name,  
    cumulative=FALSE, log=TRUE, weights=c(1,0.5),  
    samples_train, samples_valid, ...)
```

- ▶ `ae_type="both"`
- ▶ If `cumulative=FALSE`, time between two consecutive actions will be used instead of the timestamp sequence.
- ▶ If `log=TRUE`, the logarithm of the time sequence will be used.
- ▶ `weights` specifies the weights of action sequence difference and time sequence difference in the loss function.

Code demonstration: chunk 6 in `procdemo_demo.R`

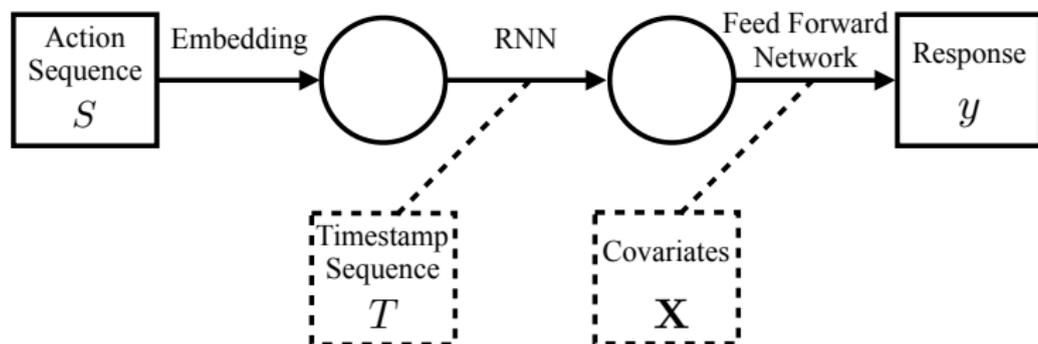
Sequence Model

Processes	Covariates	Response variable
	31	✓
	24	✗
	45	✓
	37	✗
	29	?

Sequence models can be used for

- ▶ predicting response variables based on processes and covariates
- ▶ extracting features related to the response variable

Sequence Model



Sequence Model

```
seqm(seqs, response, covariates, response_type,  
      actions, max_len, rnn_type, include_time,  
      time_interval, log_time, K_emb, K_rnn,  
      n_hidden, K_hidden, index_valid, n_epoch,  
      optimizer_name, step_size, ...)
```

Sequence Model

```
seqm(seqs, response, covariates, response_type,  
      actions, max_len, rnn_type, include_time,  
      time_interval, log_time, K_emb, K_rnn,  
      n_hidden, K_hidden, index_valid, n_epoch,  
      optimizer_name, step_size, ...)
```

- ▶ seqs: response processes
- ▶ response: response variable
- ▶ covariates: covariates
- ▶ response_type: type of response variable "binary" or "scale".

Sequence Model

```
seqm(seqs, response, covariates, response_type,  
     actions, max_len, rnn_type, include_time,  
     time_interval, log_time, K_emb, K_rnn,  
     n_hidden, K_hidden, index_valid, n_epoch,  
     optimizer_name, step_size, ...)
```

- ▶ actions: action set; it contains all actions appearing in seqs by default.
- ▶ max_len: maximum process length; it is the length of the longest process in seqs by default.
- ▶ rnn_type: structure of RNN: "lstm" or "gru".

Sequence Model

```
seqm(seqs, response, covariates, response_type,  
     actions, max_len, rnn_type, include_time,  
     time_interval, log_time, K_emb, K_rnn,  
     n_hidden, K_hidden, index_valid, n_epoch,  
     optimizer_name, step_size, ...)
```

- ▶ `include_time`: if TRUE, time sequence is included in the model.
- ▶ `time_interval`: if TRUE, time between two consecutive actions is used.
- ▶ `log_time`: if TRUE, logarithm of time sequences is used in the model.

Sequence Model

```
seqm(seqs, response, covariates, response_type,  
     actions, max_len, rnn_type, include_time,  
     time_interval, log_time, K_emb, K_rnn,  
     n_hidden, K_hidden, index_valid, n_epoch,  
     optimizer_name, step_size, ...)
```

- ▶ `K_emb`: embedding dimension
- ▶ `K_rnn`: dimension of the output of RNN
- ▶ `n_hidden`: the number of layers in the feed forward neural network.
- ▶ `K_hidden`: a numeric vector specifying the dimension of each layer in the feed forward network.

Sequence Model

```
seqm(seqs, response, covariates, response_type,  
     actions, max_len, rnn_type, include_time,  
     time_interval, log_time, K_emb, K_rnn,  
     n_hidden, K_hidden, index_valid, n_epoch,  
     optimizer_name, step_size, ...)
```

- ▶ `index_valid`: indices of processes or the proportion of processes in the validation set.
- ▶ `n_epoch`: the number of epochs to be run when optimizing the objective function.
- ▶ `optimizer_name`: name of the optimizer
- ▶ `step_size`: (baseline) step size of updates during optimization

Sequence Model

`seqm` returns an object of class `seqm`. It is a list containing

- ▶ human-readable model structure
- ▶ the fitted model to be used by `predict.seqm`
- ▶ trace of values of objective function in the training process.

Sequence Model

Predictions and features can be obtained by the `predict` method for sequence models

```
predict(object, new_seqs, new_covariates = NULL,  
        type = "response", ...)
```

- ▶ `object`: an object of class `seqm`
- ▶ `new_seqs`: an object of class `"proc"` with which to predict.
- ▶ `new_covariates`: a covariate matrix with which to predict.
- ▶ `type`: a string specifying whether to predict responses (`"response"`) or to extract features (`"feature"`) or both (`"both"`).

Code demonstration: chunk 7 in `procdemo.R`

Reference

- ▶ ProcData package: <https://cran.r-project.org/package=ProcData>
- ▶ Tang, X., Zhang, S., Wang, Z., Liu, J., and Ying, Z. (2021). ProcData: An R package for process data analysis. *Psychometrika*. Accepted. <https://arxiv.org/abs/2006.05061>
- ▶ Gómez-Alonso, C., and Valls, A. (2008). A similarity measure for sequences of categorical data based on the ordering of common elements. *Modeling decisions for artificial intelligence* pp. 134–145.
- ▶ Paradis, E. (2018) Multidimensional scaling with very large datasets. *Journal of Computational and Graphical Statistics* pp. 1–5.

Report bugs at
<https://github.com/xytangtang/ProcData/issues>.

Thank you!

xytang@math.arizona.edu