

# Package ‘bldR’

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**Type** Package

**Title** Bilingual-Led Diversity Library

**Version** 0.1.0

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**Description** This library contains functions for simulating and testing models of Bilingual-Led Divergence, such as described in the paper: Ellison, TM & L Miceli (2017 - 93(2):255-287) Language Monitoring in Bilinguals as a Mechanism for Rapid Lexical Divergence. Language. The (R6) classes include TensorModel implementing the cognitive models in the paper, PopulationSimulation providing agent-based simulations where each agent implements the cognitive model, and L2017 which constructs the graphs and statistics presented in the paper.

**License** GNU Public License Version 3 - GPLv3

**LazyData** FALSE

**RoxygenNote** 6.0.1

## R topics documented:

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## Description

bldR is a package for modelling bilingual-led divergence of lexica.

## Details

bldR provides three R6 classes: [TensorAgent](#), [PopulationSimulation](#), and [L2017](#).

Each are described on their own pages.

## See Also

Most uptodate version at <https://github.com/tyrannomark/bldR>

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| EM2017_ParameterFit | <i>EM2017_ParameterFit</i> |
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## Description

This data set contains the fit of the cognitive model described in Ellison & Miceli (2017 - Language 93(2):255-287) - hereafter EM - to the data returned by the experiment also described in the paper.

## Format

A data frame with 35968726 rows and 9 columns. Of these, 3 columns are for internal use and can be ignored. The others fields are the following:

**Form** The form returned used (one doppel per stimulus, and abstract values indicating non-doppels used in English and Dutch).

**Meaning** Names for th 41 distinct meanings tested in the experiment. These take the form of "S"<stimulus-number+"\_">(allcaps)<meaning>.

**Language** The language the simulated participant is trying to express their form in: either English or Dutch. In the experiment reported in EM, articipants were only aiming to produce forms in English, so the values for Language="Dutch" are not used in parameter-fitting.

**lm** Language mode varying from 0.00 to 1.00 in increments of 0.01.

**ml** Monitoring level varying from 0.00 to 1.00 in increments of 0.01.

**p** The conditional probability of the form given the parameter values - basically, how well did the model do at predicting the frequency with which this form would be used by bilinguals.

## Author(s)

T. Mark Ellison <m.ellison@anu.edu.au>

## References

<https://github.com/tyrannomark/bldR>

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| EM2017_Responses | <i>EM2017_Responses</i> |
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**Description**

This data set contains the responses returned in the experiment described in Ellison & Miceli (2017 - Language 93(2):255-287).

**Format**

A data frame with 2049 rows and 5 columns. The columns are as follows:

**Condition** Either "Monolingual" or "Bilingual" indicating the linguistic background of the participant.

**ParticipantId** The identifying code for the participant.

**StimulusId** The identifying code for the stimulus.

**Response** The response given by this participant to this stimulus.

**IsDoppel** A numeric value with 1 if the response is a doppel, and 0 if not.

**Author(s)**

T. Mark Ellison <m.ellison@anu.edu.au>

**References**

<https://github.com/tyrannomark/bldR>

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| EM2017_Stimuli | <i>EM2017_Stimuli</i> |
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**Description**

This data set contains the stimuli used in the experiment described in Ellison & Miceli (2017 - Language 93(2):255-287).

**Format**

A data frame with 41 rows and 7 columns. The columns are as follows:

**StimulusId** An identifying number associated with this particular stimulus.

**DutchDoppel** The Dutch half of the doppel which is a potential response to the stimulus.

**EnglishDoppel** The English half of the doppel which is a potential response to the stimulus. This is a potential response to the stimulus.

**EnglishNonDoppelExample** An example word which could also be a response to the stimulus, but which has no doppel in Dutch.

**DutchContext** This is the paragraph immediately preceding the sentence frame, tightening the semantic context and (when used with EnglishFrame) pushing the participants into bilingual mode.

**EnglishContext** This is the paragraph immediately preceding the sentence frame used with monolingual English speakers.

**EnglishFrame** This is the sentence frame (with a non-initial gap) used to elicit a response from the participants.

#### Author(s)

T. Mark Ellison <m.ellison@anu.edu.au>

#### References

<https://github.com/tyrannomark/bldR>

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| L2017 | <i>L2017</i> |
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#### Description

This class contains methods for creating the graphs in Ellison & Miceli (2017 - Language 93(2):255-287). These graphs are based on simulations from the classes `TensorAgent` and `PopulationSimulation`. There is a method corresponding to each graph, which creates the graph in PNG format and writes it to the current working directory.

#### Usage

L2017

#### Format

`R6Class` object.

#### Details

A further method uses the Student t test to evaluate the significance of the difference in distribution between the monolingual and bilingual results from the experiment.

#### Value

An object of class L2017.

Object of [L2017](#) with methods for creating graphs for Ellison & Miceli (2017) published in Language 93(2):255-287.

#### Methods

`$new()` Creates a new, empty lexicon object.

`$draw_1ab(graph_number, language, side)` Draws graphs 1a and 1b, depending on arguments (see example).

`$t_test()` Performs the Student t test comparing doppel rates between monolinguals and bilinguals in the experimental results.

`$draw_2()` Draws graph 2.

`$draw_3a()` Draws graph 3a.

`$draw_3b()` Draws graph 3b.  
`$draw_4a()` Draws graph 4a.  
`$draw_4b(graphnum="4b")` Draws graph 4b.  
`$draw_4c()` Draws graph 4c.  
`$draw_6()` Draws graph 6.  
`$draw_7a()` Draws graph 7a.  
`$draw_7b()` Draws graph 7b.  
`$draw_8()` Draws graph 8.  
`$draw_9()` Draws graph 9.  
`$draw_10a(rx=0.2, ry=0.2, xc=0.5, yc=0.5, A=100, B=100, AB=100, linetypes=c("solid", "dashed"), linecolour=)`  
 Draws graph 10a - graphs 10b, 10c build on this function, changing the default parameter values.  
`$draw_10b()` Draws graph 10b.  
`$draw_10c()` Draws graph 10c.  
`$draw_all()` Draws graphs all graphs from the paper.

### Examples

```

library(bldR)
self$t_test();
self$draw_1ab("1a", "SG", "left");
self$draw_1ab("1b", "E", "right");
self$draw_2();
self$draw_3a();
self$draw_3b();
self$draw_4a();
self$draw_4b();
self$draw_4c();
self$draw_6();
self$draw_7a();
self$draw_7b();
self$draw_9();
self$draw_10a();
self$draw_10b();
self$draw_10c();
  
```

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PopulationSimulation    *PopulationSimulation: simulating populations of tensor agents.*

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### Description

This R6 class defines functions for running agent-based simulations described in Ellison & Miceli (2017 - Language 93(2):255-287) - hereafter EM. These simulations use the cognitive model defined by the class `TensorAgent` in this package. The modelling works by setting the parameters using auxiliary methods, then calling a `$simulate()` method to execute the simulation. The simulation models a two-language community, with 2 monolingual populations and a bilingual community. The sizes of each of these communities can be set by parameter. The language mode and monitoring level are set globally in the simulation, applying to all individuals in the bilingual community (they have no effect on the monolingual communities).

**Usage**

PopulationSimulation

**Format**

R6Class object.

**Details**

Graphs 10a,b,c in EM are based on simulations run using this class.

**Value**

An object of class PopulationSimulation.

Object of `PopulationSimulation` with methods for drawing graphs according suitable for final version of the paper.

**Fields**

`LanguageMode` The parameter defining how likely agents are to generate potential (even if unrealised) intrusions from non-target languages. Parameter takes values  $[0,1]$ , 0 if no chance of intrusions, 1 if 50% chance of intrusion.

`MonitoringLevel` The parameter defining how intensely agents monitor to avoid non-target language intrusions. Parameter takes values  $[0,1]$ , 0 if no monitoring takes place, 1 if all potential intrusions are blocked.

`Population_A` The number of agents in population speaking only language A.

`Population_B` The number of agents in population speaking only language B.

`Population_AB` The number of agents in population speaking both languages: A and B.

`Population` The total number of agents.

`SamplesPerAgent` The amount of data collected from each agent to serve as inputs to other agents' lexical memory.

`NumberOfGenerations` The number of generations. In this simulation process, one generation involves two steps: talking on inputs, and generating data according to the cognitive model which will be input for the next generation.

**Methods**

`new()` Creates a new, empty PopulationSimulation object.

`$setPopulationStructure(A=100,B=100,AB=0)` Sets the population levels for the 2 monolingual communities and the bilingual community.

`$setLanguageMode(languageMode)` Sets the language mode field.

`$setMonitoringLevel(monitoredLevel)` Sets the monitoring level field.

`$setSamplesPerAgent(samplesPerAgent)` Sets the samples-per-agent field.

`$setNumberOfGenerations(numberOfGenerations)` Sets the number of generations for the simulation.

`$clearLexicon()` Empty the lexicons of all agents.

`$constructDataTensor()` Initialises the tensor structures of each agent.

`$make_p_f_st__bm()` Builds the probabilistic mapping from meanings and target languages to forms in each agent. This is a complete run of the TensorAgent simulation for each agent, based on the input distribution of language-meaning-form combinations.

`$setLexicon(A_d=0.5, A_nA=0.5, A_nB=0.0, B_d=0.5, B_nA=0.0, B_nB=0.5)` Set the initial frequency associated with each form in each language. `A_d` is the doppel as it is in language A, `A_nA` is the non-doppel native to language A as it appears in A, `A_nB` is the non-doppel native to language B should it appear in A, `B_d` is the doppel in language B, `B_nA` is the non-doppel native to A if it appears in language B, `B_nB` is the non-doppel native to B as it appears in that language. Values for each of these can be any positive floating-point number.

`$productionDistribution()` Aggregates the distribution over forms in each language.

`$sampleFromDistribution(distribution, exact=FALSE)` Takes a sample from the given distribution. If `exact` is TRUE then the distribution extracted is exact distributional copy as opposed to a random sample.

`$setDistribution(samples)` Sets the distribution over lexical items in each agent (call `$clearLexicon()` first).

`$simulate(exact=FALSE)` Simulate for the specified number of generations, with the given parameter values, from a starting distribution of `A_d=0.55, A_nA=0.45, A_nB=0.0, B_d=0.5, B_nA=0.0, B_nB=0.5`.

## Examples

```
library(bldR)
ps <- PopulationSimulation$new();
ps$setPopulationStructure(20,20,20);
ps$setNumberOfGenerations(10);
ps$setMonitoringLevel(1.0);
ps$setLanguageMode(0.54);
ps$setSamplesPerAgent(100);
ps$setLexicon();
ps$simulate(exact=FALSE);
```

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TensorAgent

*TensorAgent: a model of bilingual lexical selection.*

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## Description

This class implements the model of individual lexical selection described in Ellison & Miceli (2017 - Language 93(2):255-287) - hereafter EM. It is implemented using the TensorA, because that class facilitates linear calculations parameterised over a variable number of parameter dimensions.

## Usage

TensorAgent

## Format

R6Class object.

## Value

Object of class `TensorAgent`.

## Fields

**Documentation** Here are TensorModel's fields. EM abbreviates Ellison & Miceli (2017).

`$LanguageMode` This is the language mode parameter on a scale  $[0,1]$ .

`$MonitoringLevel` The is the effort put into monitoring on a scale  $[0,1]$ .

`$Meanings` The meanings currently in the lexicon (a list of strings).

`$Languages` The languages modelled in the lexicon (a list of strings).

`$NumberOfLanguages` The number of languages modelled in the lexicon (integer).

`$LexicalTensor` The tensor with the frequency of occurrence for each language-meaning-form combination.

`$version` An integer tracking whether updates to values to avoid repeated calculation of the same values.

`$delta_lt` Kronecker delta (1 if  $l=t$  zero otherwise), see Equation 2 in EM.

`$p_l_t__b` Probability of using language  $l$  given a target language  $t$  and language mode  $b$ .

`$p_f_sl` Probability of using form  $f$  to represent meaning  $s$  in language  $l$ , see Equation 3 in EM.

`$p_f_st__b` Probability of using form  $f$  to representing meaning  $s$  when trying to use language  $t$  and the bilingual mode is  $b$ , see Equation 4 in EM.

`$p_lfst__bm` Probability of identifying language  $l$  as the source of form  $f$  when trying to express meaning  $s$  in language  $t$ , see Equation 8 in EM.

`$p_f_st__bm` Probability of using form  $f$  to express meaning  $s$  when aiming to speak language  $t$ , given bilingual mode  $b$  and monitoring level  $m$ , see Equation 11 in EM.

`$p_l_t__b_version` The version number associated with the current value of `p_l_t__b`.

`$p_f_sl_version` The version number associated with the current value of `p_f_sl`.

`$p_f_st__b_version` The version number associated with the current value of `p_f_st__b`.

`$p_lfst__bm_version` The version number associated with the current value of `p_lfst__bm`.

`$p_f_st__bm_version` The version number associated with the current value of `p_f_st__bm`.

## Methods

`$new()` Creates a new, empty TensorModel object.

`$clearLexicon()` Empties the lexical memory, removing all meaning-language-form triples.

`$setLanguageMode(languageMode)` Set the level of interaction between target and other languages.

`$setMonitoringLevel(monitoredLevel)` Set the level of monitoring exerted by the agent.

`$addExample(meaning, language, form, ct=1)` Adds meaning-language-form triple to the model if it does not exist already. It then adds `ct` to the frequency recorded for this triple.

`$normalise(t, overIndices)` Normalise a tensor `t` over some vector of indices `overIndices`.

`$constructDataTensor()` Uses the tuples entered using `$addExample(. .)` to build a tensor representing the distribution over experienced meaning-language-form combinations.

`$makeMeLanguagePairs()` Constructs , a rank-2 tensor, both of whose indices range over languages, and whose values for each dimension combination is 1.0.

`$make_p_l_t__b()` Calculate the probability of a language given a target language.

`$make_p_f_sl()` Calculate the probability of using form  $f$  to represent meaning  $s$  in language  $l$ .



`$make_p_f_st__b()` Calculate the probability of using form `$f$` to representing meaning `$s$` when trying to use language `$t$` and the bilingual mode is `$b$`.

`$make_p_lfst__bm()` Calculate the probability of identifying language `$l$` as the source of form `$f$` when trying to express meaning `$s$` in language `$t$`.

`$make_p_f_st__bm()` Calculate the probability of using form `$f$` to express meaning `$s$` when aiming to speak language `$t$`, given bilingual mode `$b$` and monitoring level `$m$`.

`$as.data.frame()` Create a data frame with columns Meaning, Language, Form and probability of form `p`, giving the distribution `p_f_st__bm`.

## Examples

```
library(bldR)
ta <- TensorAgent$new();
ta$clearLexicon();
ta$addExample("PHOTO","English","foto", ct=0.5);
ta$addExample("PHOTO","English","picture", ct=0.5);
ta$addExample("PHOTO","Dutch", "foto", ct=1.0);
ta$constructDataTensor();
ta$setLanguageMode(0.4);
ta$setMonitoringLevel(0.8);
ta$make_p_f_st__bm();
df <- ta$as.data.frame();
print(df);
```

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