Package: politeness (via r-universe)

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Type Package

Title Detecting Politeness Features in Text

Version 0.9.3

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Description Detecting markers of politeness in English natural language. This package allows researchers to easily visualize and quantify politeness between groups of documents. This package combines prior research on the linguistic markers of politeness. We thank the Spencer Foundation, the Hewlett Foundation, and Harvard's Institute for Quantitative Social Science for support.

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Encoding UTF-8

LazyData true

Depends R (>= 3.5.0)

Imports tm, quanteda, ggplot2, parallel, spacyr, textir, glmnet, data.table, stringr, stringi

RoxygenNote 7.2.3

Suggests knitr, rmarkdown, testthat

VignetteBuilder knitr

Repository https://myeomans.r-universe.dev

RemoteUrl https://github.com/myeomans/politeness

RemoteRef HEAD

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Contents

`bowl_offers` *Purchase offers for bowl*

Description

A dataset containing the purchase offer message and a label indicating if the writer was assigned to be warm (1) or tough (0)

Usage

```
bowl_offers
```

Format

A data frame with 70 rows and 2 variables:

message character of purchase offer message

condition binary label indicating if message is warm or tough

Source

Jeong, M., Minson, J., Yeomans, M. & Gino, F. (2019).

"Communicating Warmth in Distributed Negotiations is Surprisingly Ineffective." Study 3.

Study 3. https://osf.io/t7sd6/

feature_table *Table of Politeness Features*

Description

This table describes all the text features extracted in this package. See vignette for details.

Usage

```
feature_table
```

Format

A data.frame with information about the politeness features.

findPoliteTexts *Find polite text*

Description

Finds examples of most or least polite text in a corpus

Usage

```
findPoliteTexts(
  text,
  df_polite,
  covar,
  type = c("most", "least", "both"),
  num_docs = 5L,
  ...
)
```

Arguments

text	a character vector of texts.
df_polite	a data.frame with politeness features, as outputed by politeness, used to train model.
covar	a vector of politeness labels, or other covariate.
type	a string indicating if function should return the most or least polite texts or both. If length > 1 only first value is used.
num_docs	integer of number of documents to be returned. Default is 5.
...	additional parameters to be passed to politenessProjection.

Details

Function returns a data.frame ranked by (more or least) politeness. If `type == 'most'`, the `num_docs` most polite texts will be returned. If `type == 'least'`, the `num_docs` least polite texts will be returned. If `type == 'both'`, both most and least polite text will be returned. if `num_docs` is even, half will be most and half least polite else half + 1 will be most polite.

`df_polite` must have the same number of rows as the `length(text)` and `length(covar)`.

Value

data.frame with texts ranked by (more or least) politeness. See details for more information.

Examples

```
data("phone_offers")
polite.data<-politeness(phone_offers$message, parser="none",drop_blank=FALSE)

findPoliteTexts(phone_offers$message,
                polite.data,
                phone_offers$condition,
                type = "most",
                num_docs = 5)

findPoliteTexts(phone_offers$message,
                polite.data,
                phone_offers$condition,
                type = "least",
                num_docs = 10)
```

---

phone_offers *#' Positive Emotions List #' #' Positive words. #' #' @format A list of 2006 positively-valenced words #' "positive_list"*

Description

#' Negative Emotions List #' #' Negative words. #' #' @format A list of 4783 negatively-valenced words #' "negative_list"

Usage

```
phone_offers
```

Format

A data frame with 355 rows and 2 variables:

message character of purchase offer message

condition binary label indicating if message is warm or tough

Details

#' Hedge Words List #' #' Hedges #' #' @format A list of 72 hedging words. #' "hedge_list"

#' Feature Dictionaries #' #' Six dictionary-like features for the detector: Negations; Pauses; Swearing; Pronouns; Formal Titles; and Informal Titles. #' #' @format A list of six `quanteda::dictionary` objects "polite_dicts" Purchase offers for phone

A dataset containing the purchase offer message and a label indicating if the writer was assigned to be warm (1) or tough (0)

Source

Jeong, M., Minson, J., Yeomans, M. & Gino, F. (2019).

"Communicating Warmth in Distributed Negotiations is Surprisingly Ineffective."

Study 1. https://osf.io/t7sd6/

`politeness` *Politeness Features*

Description

Detects linguistic markers of politeness in natural language. This function is the workhorse of the `politeness` package, taking an N-length vector of text documents and returning an N-row data.frame of feature counts.

Usage

```
politeness(
  text,
  parser = c("none", "spacy"),
  metric = c("count", "binary", "average"),
  drop_blank = FALSE,
  uk_english = FALSE,
  num_mc_cores = 1
)
```

Arguments

`text`	character A vector of texts, each of which will be tallied for politeness features.
`parser`	character Name of dependency parser to use (see details). Without a dependency parser, some features will be approximated, while others cannot be calculated at all.
`metric`	character What metric to return? Raw feature count totals, Binary presence/absence of features, or feature counts per 100 words. Default is "count".
`drop_blank`	logical Should features that were not found in any text be removed from the data.frame? Default is FALSE

uk_english	logical Does the text contain any British English spelling? Including variants (e.g. Canadian). Default is FALSE
num_mc_cores	integer Number of cores for parallelization. Default is 1, but we encourage users to try parallel::detectCores() if possible.

Details

Some politeness features depend on part-of-speech tagged sentences (e.g. "bare commands" are a particular verb class). To include these features in the analysis, a POS tagger must be initialized beforehand - we currently support SpaCy which must be installed separately in Python (see example for implementation).

Value

a data.frame of politeness features, with one row for every item in 'text'. Possible politeness features are listed in feature_table

References

Brown, P., & Levinson, S. C. (1987). Politeness: Some universals in language usage (Vol. 4). Cambridge university press.

Danescu-Niculescu-Mizil, C., Sudhof, M., Jurafsky, D., Leskovec, J., & Potts, C. (2013). A computational approach to politeness with application to social factors. arXiv preprint arXiv:1306.6078.

Voigt, R., Camp, N. P., Prabhakaran, V., Hamilton, W. L., ... & Eberhardt, J. L. (2017). Language from police body camera footage shows racial disparities in officer respect. Proceedings of the National Academy of Sciences, 201702413.

Examples

```
data("phone_offers")

politeness(phone_offers$message, parser="none",drop_blank=FALSE)

colMeans(politeness(phone_offers$message, parser="none", metric="binary", drop_blank=FALSE))
colMeans(politeness(phone_offers$message, parser="none", metric="count", drop_blank=FALSE))

dim(politeness(phone_offers$message, parser="none",drop_blank=FALSE))
dim(politeness(phone_offers$message, parser="none",drop_blank=TRUE))

## Not run:
# Detect multiple cores automatically for parallel processing
politeness(phone_offers$message, num_mc_cores=parallel::detectCores())

# Connect to SpaCy installation for part-of-speech features
install.packages("spacyr")
spacyr::spacy_initialize(python_executable = PYTHON_PATH)
politeness(phone_offers$message, parser="spacy",drop_blank=FALSE)


## End(Not run)
```

politenessDNM *Politeness Features*

Description

Detects linguistic markers of politeness in natural language. This function emulates the original features of the Danescu-Niculescu-Mizil Politeness paper. This primarily exists to contrast with the full feature set in the main package, and is not recommended otherwise.

Usage

```
politenessDNM(text, uk_english = FALSE)
```

Arguments

text	character A vector of texts, each of which will be tallied for politeness features.
uk_english	logical Does the text contain any British English spelling? Including variants (e.g. Canadian). Default is FALSE

Value

a data.frame of politeness features, with one row for every item in ‘text‘. The original names are used where possible.

References

Danescu-Niculescu-Mizil, C., Sudhof, M., Jurafsky, D., Leskovec, J., & Potts, C. (2013). A computational approach to politeness with application to social factors. arXiv preprint arXiv:1306.6078.

Examples

```
## Not run:
# Connect to SpaCy installation for part-of-speech features
install.packages("spacyr")
spacyr::spacy_initialize(python_executable = PYTHON_PATH)
data("phone_offers")

politeness(phone_offers$message)


## End(Not run)
```

politenessModel *Pre-Trained Politeness Classifier*

Description

Pre-trained model to detect politeness based on data from Danescu-Niculescu-Mizil et al. (2013)

Usage

```
politenessModel(texts, num_mc_cores = 1)
```

Arguments

texts	character A vector of texts, each of which will be given a politeness score.
num_mc_cores	integer Number of cores for parallelization.

Details

This is a wrapper around a pre-trained model of "politeness" for all the data from the 2013 DNM et al paper. This model requires grammar parsing via SpaCy. Please see spacyr for details on installation.

Value

a vector with receptiveness scores

References

Danescu-Niculescu-Mizil, C., Sudhof, M., Jurafsky, D., Leskovec, J. & Potts, C. (2013). A computational approach to politeness with application to social factors. Proc. 51st ACL, 250-259.

Examples

```
## Not run:
data("phone_offers")

politenessModel(phone_offers$message)


## End(Not run)
```

politenessPlot *Politeness plot*

Description

Plots the prevalence of politeness features in documents, divided by a binary covariate.

Usage

```
politenessPlot(
  df_polite,
  split = NULL,
  split_levels = NULL,
  split_name = NULL,
  split_cols = c("firebrick", "navy"),
  top_title = "",
  drop_blank = 0.05,
  middle_out = 0.5,
  features = NULL,
  ordered = FALSE,
  CI = 0.68
)
```

Arguments

df_polite	a data.frame with politeness features calculated from a document set, as output by politeness.
split	a vector of covariate values. must have a length equal to the number of documents included in df_polite. No NA values allowed.
split_levels	character vector of length 2 default NULL. Labels for covariate levels for legend. If NULL, this will be inferred from split.
split_name	character default NULL. Name of the covariate for legend.
split_cols	character vector of length 2. Name of colors to use.
top_title	character default "". Title of plot.
drop_blank	Features less prevalent than this in the sample value are excluded from the plot. To include all features, set to 0
middle_out	Features less distinctive than this value (measured by p-value of t-test) are excluded. Defaults to 1 (i.e. include all).
features	character vector of feature names. If NULL all will be included.
ordered	logical should features be ordered according to features param? default is FALSE.
CI	Coverage of error bars. Defaults to 0.68 (i.e. standard error).

Details

Length of `split` must be the same as number of rows of `df_polite`. Typically `split` should be a two-category variable. However, if a continuous covariate is given, then the top and bottom terciles of that distribution are treated as the two categories (while dropping data from the middle tercile).

Value

a ggplot of the prevalence of politeness features, conditional on `split`. Features are sorted by variance-weighted log odds ratio.

Examples

```
data("phone_offers")

polite.data<-politeness(phone_offers$message, parser="none", drop_blank=FALSE)

politeness::politenessPlot(polite.data,
                           split=phone_offers$condition,
                           split_levels = c("Tough","Warm"),
                           split_name = "Condition",
                           top_title = "Average Feature Counts")


politeness::politenessPlot(polite.data,
                           split=phone_offers$condition,
                           split_levels = c("Tough","Warm"),
                           split_name = "Condition",
                           top_title = "Average Feature Counts",
                           features=c("Positive.Emotion","Hedges","Negation"))


polite.data<-politeness(phone_offers$message, parser="none", metric="binary", drop_blank=FALSE)

politeness::politenessPlot(polite.data,
                           split=phone_offers$condition,
                           split_levels = c("Tough","Warm"),
                           split_name = "Condition",
                           top_title = "Binary Feature Use")
```

---

politenessProjection *Politeness projection*

Description

Training and projecting a regression model using politeness features.

Usage

```
politenessProjection(
  df_polite_train,
  covar = NULL,
  df_polite_test = NULL,
  classifier = c("glmnet", "mnir"),
  cv_folds = NULL,
  ...
)
```

Arguments

df_polite_train	a data.frame with politeness features as outputed by politeness used to train model.
covar	a vector of politeness labels, or other covariate.
df_polite_test	optional data.frame with politeness features as outputed by politeness used for out-of-sample fitting. Must have same feature set as polite_train (most easily achieved by setting dropblank=FALSE in both calls to politeness).
classifier	name of classification algorithm. Defaults to "glmnet" (see glmnet) but "mnir" (see mnlm) is also available.
cv_folds	Number of outer folds for projection of training data. Default is NULL (i.e. no nested cross-validation). However, positive values are highly recommended (e.g. 10) for in-sample accuracy estimation.
...	additional parameters to be passed to the classification algorithm.

Details

List:

- train_proj projection of politeness model within training set.
- test_proj projection of politeness model onto test set (i.e. out-of-sample).
- train_coef coefficients from the trained model.

Value

List of df_polite_train and df_polite_test with projection. See details.

Examples

```
data("phone_offers")
data("bowl_offers")

polite.data<-politeness(phone_offers$message, parser="none",drop_blank=FALSE)

polite.holdout<-politeness(bowl_offers$message, parser="none",drop_blank=FALSE)

project<-politenessProjection(polite.data,
```

```
                                phone_offers$condition,
                                polite.holdout)

# Difference in average politeness across conditions in the new sample.

mean(project$test_proj[bowl_offers$condition==1])
mean(project$test_proj[bowl_offers$condition==0])
```

`polite_train` *Pre-Trained Politeness*

Description

A dataset to train a model for detecting politeness.

Usage

```
polite_train
```

Format

list of two objects. x contains pre-calculated politeness features for each document. y contains standardized human annotations for politeness.

Source

Danescu-Niculescu-Mizil, C., Sudhof, M., Jurafsky, D., Leskovec, J. & Potts, C. (2013). A computational approach to politeness with application to social factors. Proc. 51st ACL, 250-259.

`receptiveness` *Conversational Receptiveness*

Description

Pre-trained model to detect conversational receptiveness

Usage

```
receptiveness(texts, num_mc_cores = 1)
```

Arguments

`texts`	character A vector of texts, each of which will be tallied for politeness features.
`num_mc_cores`	integer Number of cores for parallelization.

Details

This is a wrapper around a pre-trained model of "conversational receptiveness". The model trained from Study 1 of that paper can be applied to new text with a single function. This model requires grammar parsing via SpaCy. Please see spacyr for details on installation.

Value

a vector with receptiveness scores

References

Yeomans, M., Minson, J., Collins, H., Chen, F. & Gino, F. (2020). Conversational Receptiveness: Improving Engagement with Opposing Views. OBHDP.

Examples

```
## Not run:
data("phone_offers")

receptiveness(phone_offers$message)


## End(Not run)
```

---

receptive_model

A pre-trained model for detecting conversational receptiveness. Estimated with glmnet using annotated data from a previous paper. Primarily for use within the receptiveness() function.

Description

A pre-trained model for detecting conversational receptiveness. Estimated with glmnet using annotated data from a previous paper. Primarily for use within the receptiveness() function.

Usage

```
receptive_model
```

Format

A fitted glmnet model

Source

Minson, J., Yeomans, M., Collins, H. & Dorison, C.

"Conversational Receptiveness: Improving Engagement with Opposing Views"

`receptive_names` *This is the list of variables to be extracted for the receptiveness algorithm For internal use only, within the receptiveness() function.*

Description

This is the list of variables to be extracted for the receptiveness algorithm For internal use only, within the receptiveness() function.

Usage

```
receptive_names
```

Format

Character vector containing variable names

Source

Minson, J., Yeomans, M., Collins, H. & Dorison, C.

"Conversational Receptiveness: Improving Engagement with Opposing Views"

`receptive_polite` *Pre-Trained Receptiveness Data*

Description

A dataset to train a model for detecting conversational receptiveness.

Usage

```
receptive_polite
```

Format

Pre-calculated politeness features for the receptive_train dataset

receptive_train *Pre-Trained Receptiveness Data*

Description

A dataset to train a model for detecting conversational receptiveness.

Usage

```
receptive_train
```

Format

A data frame with 2860 rows and 2 variables:

text character written response about policy disagreement

receptive numeric standardized average of annotator ratings for "receptiveness"

Primarily for use within the receptiveness() function. The data was compiled from Studies 1 and 4 of the original paper, as well as an unpublished study with a very similar design, in which text responses were rated by disagreeing others.

Source

Yeomans, M., Minson, J., Collins, H., Chen, F. & Gino, F. (2020).

"Conversational Receptiveness: Improving Engagement with Opposing Views"

https://osf.io/2n59b/

uk2us *UK to US Conversion dictionary*

Description

For internal use only. This dataset contains a quanteda dictionary for converting UK words to US words. The models in this package were all trained on US English.

Usage

```
uk2us
```

Format

A quanteda dictionary with named entries. Names are the US version, and entries are the UK version.

Source

Borrowed from the quanteda.dictionaries package on github (from user kbenoit)

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