Package: maptpx (via r-universe)

November 4, 2024

Title MAP Estimation of Topic Models

Version 1.9-7

Author Matt Taddy <mataddy@gmail.com>

Depends R (>= 2.10), slam

Suggests MASS

Description Maximum a posteriori (MAP) estimation for topic models
(i.e., Latent Dirichlet Allocation) in text analysis, as
described in Taddy (2012) 'On estimation and selection for
topic models'. Previous versions of this code were included as
part of the 'textir' package. If you want to take advantage of
openmp parallelization, uncomment the relevant flags in
src/MAKEVARS before compiling.

Maintainer Matt Taddy <mataddy@gmail.com>

License GPL-3

URL <http://taddylab.com>

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

RemoteUrl https://github.com/taddylab/maptpx

RemoteRef HEAD

RemoteSha a6639c25ff0d133feeced4ba9b50345c6ddd191c

Contents

1

---

counts   *Utilities for count matrices*

Description

Tools for manipulating (sparse) count matrices.

Usage

```
normalize(x,byrow=TRUE)
stm_tfidf(x)
```

Arguments

x	A `simple_triplet_matrix` or `matrix` of counts.
byrow	Whether to normalize by row or column totals.

Value

`normalize` divides the counts by row or column totals, and `stm_tfidf` returns a matrix with entries $x_{ij} \log[n/(d_j+1)]$, where x_{ij} is term-j frequency in document-i, and d_j is the number of documents containing term-j.

Author(s)

Matt Taddy <mataddy@gmail.com>

Examples

```
normalize( matrix(1:9, ncol=3) )
normalize( matrix(1:9, ncol=3), byrow=FALSE )

(x <- matrix(rbinom(15,size=2,prob=.25),ncol=3))
stm_tfidf(x)
```

---

predict.topics   *topic predict*

Description

Predict function for Topic Models

Usage

```
## S3 method for class 'topics'
predict( object, newcounts, loglhd=FALSE, ... )
```

Arguments

`object`	An output object from the `topics` function, or the corresponding matrix of estimated topics.
`newcounts`	An `nrow(object$theta)`-column matrix of multinomial phrase/category counts for new documents/observations. Can be either a simple `matrix` or a `simple_triplet_matrix`.
`loglhd`	Whether or not to calculate and return `sum(x*log(p))`, the un-normalized log likelihood.
`...`	Additional arguments to the undocumented internal `tpx*` functions.

Details

Under the default mixed-membership topic model, this function uses sequential quadratic programming to fit topic weights Ω for new documents. Estimates for each new ω_i are, conditional on `object$theta`, MAP in the (K-1)-dimensional logit transformed parameter space.

Value

The output is an `nrow(newcounts)` by `object$K` matrix of document topic weights, or a list with including these weights as `W` and the log likelihood as `L`.

Author(s)

Matt Taddy <mataddy@gmail.com>

References

Taddy (2012), *On Estimation and Selection for Topic Models*. http://arxiv.org/abs/1109.4518

See Also

topics, plot.topics, summary.topics, congress109

Examples

```
## Simulate some data
omega <- t(rdir(500, rep(1/10,10)))
theta <- rdir(10, rep(1/1000,1000))
Q <- omega%*%t(theta)
counts <- matrix(ncol=1000, nrow=500)
totals <- rpois(500, 200)
for(i in 1:500){ counts[i,] <- rmultinom(1, size=totals[i], prob=Q[i,]) }

## predict omega given theta
W <- predict.topics( theta, counts )
plot(W, omega, pch=21, bg=8)
```

---

rdir   *Dirichlet RNG*

Description

Generate random draws from a Dirichlet distribution

Usage

```
rdir(n, alpha)
```

Arguments

n	The number of observations.
alpha	A vector of scale parameters, such that $E[p_j] = \alpha_j / \sum_i \alpha_i$.

Value

An n column matrix containing the observations.

Author(s)

Matt Taddy <mataddy@gmail.com>

Examples

```
rdir(3,rep(1,6))
```

---

topics   *Estimation for Topic Models*

Description

MAP estimation of Topic models

Usage

```
topics(counts, K, shape=NULL, initopics=NULL,
  tol=0.1, bf=FALSE, kill=2, ord=TRUE, verb=1, ...)
```

Arguments

counts	A matrix of multinomial response counts in `ncol(counts)` phrases/categories for `nrow(counts)` documents/observations. Can be either a simple `matrix` or a `simple_triplet_matrix`.
K	The number of latent topics. If `length(K)>1`, `topics` will find the Bayes factor (vs a null single topic model) for each element and return parameter estimates for the highest probability K.
shape	Optional argument to specify the Dirichlet prior concentration parameter as `shape` for topic-phrase probabilities. Defaults to `1/(K*ncol(counts))`. For fixed single K, this can also be a `ncol(counts)` by K matrix of unique shapes for each topic element.
initopics	Optional start-location for $[\theta_1...\theta_K]$, the topic-phrase probabilities. Dimensions must accord with the smallest element of K. If `NULL`, the initial estimates are built by incrementally adding topics.
tol	Convergence tolerance: optimization stops, conditional on some extra checks, when the *absolute* posterior increase over a full paramater set update is less than `tol`.
bf	An indicator for whether or not to calculate the Bayes factor for univariate K. If `length(K)>1`, this is ignored and Bayes factors are always calculated.
kill	For choosing from multiple K numbers of topics (evaluated in increasing order), the search will stop after `kill` consecutive drops in the corresponding Bayes factor. Specify `kill=0` if you want Bayes factors for all elements of K.
ord	If `TRUE`, the returned topics (columns of `theta`) will be ordered by decreasing usage (i.e., by decreasing `colSums(omega)`).
verb	A switch for controlling printed output. `verb > 0` will print something, with the level of detail increasing with `verb`.
...	Additional arguments to the undocumented internal `tpx*` functions.

Details

A latent topic model represents each i'th document's term-count vector X_i (with $\sum_j x_{ij} = m_i$ total phrase count) as having been drawn from a mixture of K multinomials, each parameterized by topic-phrase probabilities θ_i, such that

$$X_i \sim MN(m_i, \omega_1\theta_1 + ... + \omega_K\theta_K).$$

We assign a K-dimensional Dirichlet(1/K) prior to each document's topic weights $[\omega_{i1}...\omega_{iK}]$, and the prior on each θ_k is Dirichlet with concentration α. The `topics` function uses quasi-newton accelerated EM, augmented with sequential quadratic programming for conditional $\Omega|\Theta$ updates, to obtain MAP estimates for the topic model parameters. We also provide Bayes factor estimation, from marginal likelihood calculations based on a Laplace approximation around the converged MAP parameter estimates. If input `length(K)>1`, these Bayes factors are used for model selection. Full details are in Taddy (2011).

Value

An `topics` object list with entries

K	The number of latent topics estimated. If input `length(K)>1`, on output this is a single value corresponding to the model with the highest Bayes factor.
theta	The `ncol{counts}` by K matrix of estimated topic-phrase probabilities.
omega	The `nrow{counts}` by K matrix of estimated document-topic weights.
BF	The log Bayes factor for each number of topics in the input K, against a null single topic model.
D	Residual dispersion: for each element of K, estimated dispersion parameter (which should be near one for the multinomial), degrees of freedom, and p-value for a test of whether the true dispersion is > 1.
X	The input count matrix, in `dgTMatrix` format.

Note

Estimates are actually functions of the MAP (K-1 or p-1)-dimensional logit transformed natural exponential family parameters.

Author(s)

Matt Taddy <mataddy@gmail.com>

References

Taddy (2012), *On Estimation and Selection for Topic Models*. http://arxiv.org/abs/1109.4518

See Also

plot.topics, summary.topics, predict.topics, wsjibm, congress109, we8there

Examples

```
## Simulation Parameters
K <- 10
n <- 100
p <- 100
omega <- t(rdir(n, rep(1/K,K)))
theta <- rdir(K, rep(1/p,p))

## Simulated counts
Q <- omega%*%t(theta)
counts <- matrix(ncol=p, nrow=n)
totals <- rpois(n, 100)
for(i in 1:n){ counts[i,] <- rmultinom(1, size=totals[i], prob=Q[i,]) }

## Bayes Factor model selection (should choose K or nearby)
summary(simselect <- topics(counts, K=K+c(-5:5)), nwrd=0)
```

```
## MAP fit for given K
summary( simfit <- topics(counts,  K=K, verb=2), n=0 )

## Adjust for label switching and plot the fit (color by topic)
toplab <- rep(0,K)
for(k in 1:K){ toplab[k] <- which.min(colSums(abs(simfit$theta-theta[,k]))) }
par(mfrow=c(1,2))
tpxcols <- matrix(rainbow(K), ncol=ncol(theta), byrow=TRUE)
plot(theta,simfit$theta[,toplab], ylab="fitted values", pch=21, bg=tpxcols)
plot(omega,simfit$omega[,toplab], ylab="fitted values", pch=21, bg=tpxcols)
title("True vs Fitted Values (color by topic)", outer=TRUE, line=-2)

## The S3 method plot functions
par(mfrow=c(1,2))
plot(simfit, lgd.K=2)
plot(simfit, type="resid")
```

---

topicVar	*topic variance*

Description

Tools for looking at the variance of document-topic weights.

Usage

```
topicVar(counts, theta, omega)
logit(prob)
expit(eta)
```

Arguments

counts	A matrix of multinomial response counts, as inputed to the `topics` or `predict.topics` functions.
theta	A fitted topic matrix, as ouput from the `topics` or `predict.topics` functions.
omega	A fitted document topic-weight matrix, as ouput from the `topics` or `predict.topics` functions.
prob	A probability vector (positive and sums to one) or a matrix with probability vector rows.
eta	A vector of the natural exponential family parameterization for a probability vector (with first category taken as null) or a matrix with each row the NEF parameters for a single observation.

Details

These function use the natural exponential family (NEF) parametrization of a probability vector $q_0...q_{K-1}$ with the first element corresponding to a 'null' category; that is, with $NEF(q) = e_1...e_{K-1}$ and setting $e_0 = 0$, the probabilities are

$$q_k = \frac{exp[e_k]}{1 + \sum exp[e_j]}.$$

Refer to Taddy (2012) for details.

Value

`topicVar` returns an array with dimensions $(K-1, K-1, n)$, where K=ncol(omega)=ncol(theta) and n = nrow(counts) = nrow(omega), filled with the posterior covariance matrix for the NEF parametrization of each row of `omega`. Utility `logit` performs the NEF transformation and `expit` reverses it.

Author(s)

Matt Taddy <mataddy@gmail.com>

References

Taddy (2012), *On Estimation and Selection for Topic Models*. http://arxiv.org/abs/1109.4518

See Also

topics, predict.topics

Index