Package 'bayesmeta'

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kshika Maduwanthi Ruberu	
e Lakshika Maduwanthi Ruberu <lakshikaruberu@gmail.com></lakshikaruberu@gmail.com>	
Description Estimate meta-analytic age-specific risk of getting cancer (penetrance) due to pathogenic variants of a given gene by integrating information from studies reporting different types of risk measures on that particular gene. These risk measures include age-specific penetrance, relative risk(RR), standard incidence ratio (SIR), and odds ratio (OR).	
License GPL-2 Imports dplyr, mvtnorm, rlist, Rcpp	
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Estimates consensus age-specific penetrance of cancer for carriers of a pathogenic gene mutation	

each specified age up to age 85.

carriers of pathogenic variants of a specific gene. Information from studies reporting different types of risk measures on that gene (e.g., age-specific penetrance, relative risk(RR), standard incidence ratio (SIR), and odds ratio (OR)) are integrated to give the age-specific penetrance for carriers at

Usage

```
bayesmeta(penet,RR_studies=T,RR,OR_studies=T,OR,ages=20:85,zero_studies=T,zero_OR,pl=F,
         ylim=c(0,1),xlim=c(20,85))
```

Arguments

penet

A data frame containing information for studies reporting age-specific penetrance with following columns:

label to identify the set of penetrance values from a single study. study label

penetrance age-specific penetrance value. lower 95% confidence limit. penet_ci_lower penet ci upper upper 95% confidence limit.

age at which penetrance is reported. ages_penet

RR studies A logical variable to indicate whether studies reporting RR or SIR are included

in the meta analysis. Default is TRUE.

RR If RR_studies is TRUE, a data frame containing information for all RR and SIR

studies with following columns where each row corresponds to a single study: for studies that do not provide age-related summaries or if age related column

not provided, default values will be used

reported RR/SIR value. R.est RR.ci.lower lower 95% confidence limit. upper 95% confidence limit. RR.ci.upper

> mean age of onset for carriers. Default is 63. \boldsymbol{A}

sd of age of onset for carriers. Default is 14.00726.

A.lo minimum age of among for carriers. Default is 20. A.hi maximum age of among for carriers. Default is 95.

mean age of onset for non-carriers. Default is 63. A0

V0 sd of age of onset for non-carriers. Default is 14.00726. A0.lo minimum age of onset for non-carriers. Default is 20.

A0.hi maximum age of onset for non-carriers. Default is 95.

OR_studies A logical variable to indicate whether studies reporting OR are included in the

meta analysis. Default is TRUE.

OR If OR_studies is TRUE, a data frame containing information for all OR studies with following columns where each row corresponds to a single study: for studies that do not provide age-related summaries or if age related column not

provided, default values will be used

OR.est reported OR value.

OR.ci.lower lower 95% confidence limit. OR.ci.upper upper 95% confidence limit.

> mean age of onset for cases. Default is 63. \boldsymbol{A}

 \boldsymbol{V} sd of age of onset for cases. Default is 14.00726.

A.lo minimum age of onset among cases. Default is 20.

A.hi maximum age of onset among cases. Default is 95.

A0mean age of controls (at study inclusion). Default is value A.

V0 sd of age of controls (at study inclusion). Default is value V.

A0.lo minimum age of controls (at study inclusion). Default is value A.lo.

A0.hi maximum age of controls (at study inclusion). Default is value A.hi.

zero_studies A logical variable to indicate whether information from case control studies where no mutations were detected in controls are included in the meta analysis. Default if TRUE.

zero_OR If zero_studies is TRUE, a data frame containing information for such studies with following columns where each row corresponds to a single study: for studies that do not provide age-related summaries or if age related column not provided, default values will be used

carrier.cases number of carrier cases.non_carrier.cases number of non-carrier cases.number of non-carrier controls.

A.lo minimum age of onset among cases. Default is 20.A.hi maximum age of onset among cases. Default is 95.

A0 mean age of controls (at study inclusion). Default is value A.V0 sd of age of controls (at study inclusion). Default is value V.

A0.lo minimum age of controls (at study inclusion). Default is value A.lo. **A0.hi** maximum age of controls (at study inclusion). Default is value A.hi.

ages Ages at which penetrance values are required. Default is 20:85 at increments of

1. Maximum possible age is 85

pl If pl=TRUE, returns a plot of the estimated age specific penetrance value vs age

with confidence intervals. Default is FALSE.

ylim If pl=TRUE, numeric vector of length 2, giving the y coordinates. Default is

c(0,1).

x1im If pl=TRUE, numeric vector of length 2, giving the x coordinates. Default is

c(20,85).

Details

The bayesmeta function estimates consensus age-specific risks of developing cancer associated with mutations in a specific gene, i.e., penetrance estimation. A meta-analysis approach based on a Bayesian hierarchical random-effects model is used to obtain penetrance estimates integrating studies reporting different types of risk measures (e.g., penetrance, relative risk, odds ratio, and standard incidence ratio) while accounting for associated uncertainties.

The cumulative penetrance $F_s(t|\kappa_s, \lambda_s)$ at age t for study s is assumed to be given by the c.d.f. of a Weibull distribution with shape parameter κ_s and scale parameter λ_s .

The prior distributions are $\pi(\kappa_s|a,b) = \text{Gamma}(a,b), \ \pi(\lambda_s|c,d) = \text{Gamma}(c,d), \ \text{where } a \text{ and } c$ are shape parameters and b and d are scale parameters.

Continuous uniform distributions are assumed for all the hyper-parameters. Specifically, $\pi(a|l_a,u_a)=U(l_a,u_a), \ \pi(b|l_b,u_b)=U(l_b,u_b), \ \pi(c|l_c,u_c)=U(l_c,u_c), \ \text{ and } \ \pi(d|l_d,u_d)=U(l_d,u_d) \text{ with } l_a,u_a,l_a,u_a,l_a,u_a,l_a,u_a \text{ pre-specified.}$

Posterior distributions are obtained via a Markov chain Monte Carlo algorithm, that employ a Metropolis-Hastings algorithm within Gibbs sampling. The default values for age distributions are based on the distribution of age of onset of breast cancer for US general population (https://seer.cancer.gov/statfacts/html/breast.html). Credible interval at a given age is the 0.025^{th} and 0.0975^{th} quantiles of the posterior dstribution of penetrance estimate at that age.

Value

bayesmeta returns a list of two objects—the estimated penetrance values and 95% credible interval.

```
penetrance Estimated age-specif penetrance values penetrance_CI Credible interval
```

Author(s)

Thanthirige Lakshika Maduwanthi Ruberu

References

1. Ruberu, T. L. M, Braun, D, Parmigiani, G. and Biswas, S. Bayesian Meta-Analysis of Penetrance and its Application to Breast Cancer Risk among ATM Mutation Carriers. (Manuscript) 2. Marabelli, M., Cheng, S. C., and Parmigiani, G. (2016). Penetrance of ATM Gene Mutations in Breast Cancer: A Meta-Analysis of Different Measures of Risk. Genetic Epidemiology, 40, 425-431.

Examples

```
#penetrance value
set.seed(123)
penet1=c(0.134, 0.230, 0.400, 0.571, 0.706) #penetrance values from study 1
penet2=c(0.092, 0.146, 0.260, 0.403, 0.556) #penetrance values from study 2
penet3=c(0.034,0.072,0.131, 0.243,0.370) #penetrance values from study 3
penet4=c(0.082,0.160,0.256, 0.422 ,0.588) # #penetrance values from study 4
penet1_ci_low=c( 0.113, 0.203,0.369,0.539,0.675) #lower 95% confidence limit of values in penet1
penet1_ci_hi=c(0.155, 0.256,0.430,0.601,0.734) #upper 95% confidence limit of values in penet1
penet2_ci_low=c( 0.066,0.114, 0.221 , 0.359,0.508) #lower 95% confidence limit of values in penet2
penet2_ci_hi=c(0.117,0.176,0.298,0.446,0.600) #upper 95% confidence limit of values in penet2
penet3_ci_low=c( 0.023,0.056,0.110,0.216,0.338)#lower 95% confidence limit of values in penet3
penet3_ci_hi=c(0.045,0.088,0.152,0.270,0.402) #upper 95% confidence limit of values in penet3
penet4_ci_low=c(0.058,0.127,0.217,0.377,0.540)#lower 95% confidence limit of values in penet4
penet4\_ci\_hi=c(\emptyset.106~,\emptyset.192,\emptyset.294,\emptyset.465,\emptyset.632)~ \#upper~95\%~ confidence~ limit~of~ values~ in~penet4~ and the confidence~ limit~ of~ values~ in~penet4~ and the confidence~ limit~ limi
ages_penet1=c(35,40,50,60,70) # ages corresponding to values in penet 1
ages\_penet2=c(40,50,60,70,80) # ages corresponding to values in penet 2
ages\_penet3=c(35,40,50,60,70) # ages corresponding to values in penet 3
ages\_penet4=c(40,50,60,70,80) # ages\_corresponding\_to\_values\_in\_penet\_4
study_number=c(rep(1,5),rep(2,5),rep(3,5),rep(4,5)) # 1 for 5 records of study 1, 2 for study 2 ...
penet=data.frame('penetrance'=c(penet1,penet2,penet3,penet4),
                         'penet_ci_lower'=c(penet1_ci_low,penet2_ci_low,penet3_ci_low,penet4_ci_low),
                                 'penet_ci_upper'=c(penet1_ci_hi,penet2_ci_hi,penet3_ci_hi,penet4_ci_hi),
                                 'ages_penet'=c(ages_penet1,ages_penet2,ages_penet3,ages_penet4),
                                 'study_label'=study_number)
## RR values
R.est=c(4.71,4.31,3.31, 3.27) #RR or SIR values reported by each study
RR.ci.lower=c(3.80,3.69,2.87,2.97) #corresponding 95% lower limit
RR.ci.upper=c(5.84,5.04,3.82,3.60) #corresponding 95% upper limit
```

```
#Age related summaries from each study
A=c(60.21,64.88,65.81,63)
V=c(17.34, 15.07, 15.18, 14.01)
A.lo=c(20, 20, 20, 20)
A.hi=c(95,95,95,95)
A0=c(67.61, 67.34, 68.17, 67.18)
V0=c( 14.33, 16.30, 15.18, 14.01)
A0.lo=c(20, 20, 20, 20)
A0.hi=c(95,95,95,95)
RR=data.frame(R.est,RR.ci.lower,RR.ci.upper,A,V,A.lo,A.hi,A0,V0,A0.lo,A0.hi)
colnames(RR)=c("R.est","RR.ci.lower","RR.ci.upper","A","V","A.lo","A.hi","A0","V0","A0.lo","A0.hi")
###OR values
OR.est=c(6.25, 8.91, 22.97) #OR value reported by each study
OR.ci.lower=c(1.38,4.89, 3.08) #corresponding 95% lower limit
OR.ci.upper=c(28.23,16.24, 171.05) #corresponding 95% lower limit
#Age related summaries from each study
A=c(66.69, 66.58, 63.00)
V=c(15.34, 16.32, 14.01)
A.lo=c(20, 20, 20)
A.hi=c(95,95,95)
A0=c(66.69, 66.58, 63.00)
V0=c(15.34, 16.32, 14.01)
A0.lo=c(20, 20, 20)
A0.hi=c(95,95,95)
OR=data.frame(OR.est,OR.ci.lower,OR.ci.upper,A,V,A.lo,A.hi,A0,V0,A0.lo,A0.hi)
colnames(OR)=c("OR.est","OR.ci.lower","OR.ci.upper","A","V","A.lo","A.hi","A0","V0","A0.lo","A0.hi")
######studies with no mutations in controls
carrier.cases=c(1)
non_carrier.cases=c(99)
non_carrier.controls=c(100)
#Age related summaries from each study
A=c(48)
V=c(13.57)
A.lo=c(25)
A.hi=c(78)
A0=c(48)
V0=c(13.57)
A0.1o=c(25)
A0.hi=c(78)
zero_OR=data.frame(carrier.cases,non_carrier.cases,non_carrier.controls,
```

A, V, A. lo, A. hi, A0, V0, A0. lo, A0. hi)

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