

What is mgcv?

mgcv is an R package for fitting generalized additive models (GAMs). That means we can fit models where the predictors are smooth functions of the covariates. Often these smooth functions are splines, but that's not all they can be.

The main functions in mgcv

gam

For fitting GAMs

gamm

For fitting generalized additive mixed models. Can include correlation structures and performance can be better for random effects. You can specify random effects using lme syntax.

bar

For fitting big additive models. Includes some special tricks for fitting to large datasets.

Formula

formula=

We can write a model formula in mgcv just as we can when we use lm or glm, with some additions.

 ${\tt s}$ () is the general setup for a smooth.

te() interaction via tensor product.

Distributions

Sinh-arcsinh

location-scale-shape General family

family=

shash

qfam

Binomial binomial Normal gaussian Gamma Gamma Inverse normal inverse.gaussian Poisson poisson Quasi quasi Quasi-binomial quasibinomial Quasi-Poisson quasipoisson Tweedie tw/Tweedie Negative binomial nb/neabin Beta betar Censored normal cnorm Ordered categorical ocat Scaled t scat. Zero inflated Poisson ziP Zero inflated Poisson ziplss location-scale Cox proportional cox.ph hazards Generalized extreme gevlss value location-scale Normal gaulss location-scale model Multivariate normal mvn Gamma gammals location-scale Gumbel qumbls location-scale Multinomial multinom Tweedie twlss location-scale

Smoothers

Using the bs= argument in s(), te(), etc. Further details can be found in ?smooth.construct.*.smooth.spec

Univariate only smoothers

Cubic regression splines cr

Cubic regression splines with shrinkage cs

Cyclic cubic splines cc

B-splines bs

P-splines ps

Special smoothers

Adaptive smoothers ad

Factor-smooth interactions sz

Random factor-smooth interactions fs

Smoothers in > 1 dimension

Thin plate regression splines tp

Thin plate regression splines within shrinkage ts

Duchon splines ds

Random effects re

Markov random fields mrf

Gaussian process smooths gp

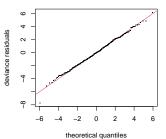
Smoothers in 2 dimensions

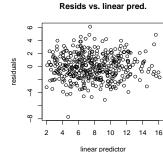
Splines on the sphere sos

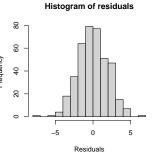
Soap film smoothing so (sw and sf)

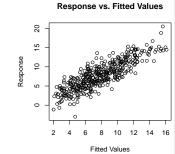
Model checking

gam.check









Top left: Quantile-quantile plot: points should be close to the line, meaning residuals are normally distributed.

Bottom left: Histogram of residuals: again, looking for normal(ish) distribution.

Top right: Residuals vs. linear predictor: looking for no increase or decrease in spread with increasing linear predictor value, otherwise we have heteroskedasticity.

Bottom right: Response vs. linear predictor, expecting tight line relationship indicating good agreement between the predictions and data.

Knots and basis complexity

General strategy: check ${\tt k}$ and double if too small. When do we know ${\tt k}$ is too small?

Fitting criterion method=

"GCV.Cp"	Generalized cross validation, default
"REML"	REstricted Maximum Likelihood, preferred
"ML"	Maximum Likelihood
"NCV"	Neighbourhood Cross-Validation

Extras

gam.mh	Metropolis-Hastings sampling of the posterior
concurvity	Assess concurvity between terms
gam.vcomp	Random effects style output
gamSim	Simulate GAM-type data
inSide/in.out	point-in-polygon test
jagam	Generate JAGS/Nimble code
new.name	Generate a variable name
place.knots	Place knots evenly
rmvn	Generate multivariate normal deviates

Extra help

?gam.models	Fitting fancy models
?linear.functionals	How to use by=
?random.effects	Random effects syntax
?mgcv.FAQ	frequently asked questions
?mgcv.parallel	Info on parallelisation
?missing.data	What to do about missing data
?choose.k	How to select basis size
?one.se.rule	Making smoother smooth models

Other packages

scam	Shape constrained smoothing
gratia	Plotting with ggplot2
mgcViz	Fancy plotting
qgam	Quantile GAMs
gamm4	Bandom effects besed on lme4

Useful references

Wood. Generalized Additive Models. An Introduction with R. 2nd ed. CRC Press, 2017

Pedersen, Miller, Simpson and Ross. Hierarchical Generalized Additive Models in Ecology: An Introduction with mgcv . PeerJ (2019). https://doi.org/10.7717/peerj.