

# Model checking

*“perhaps the most important part of applied statistical modelling”*

Simon Wood

# Model checking

- Checking  $\neq$  validation!
- As with detection function, checking is important
- Want to know the model conforms to assumptions
- What assumptions should we check?

# What to check

- Convergence
- Basis size
- Residuals

# Convergence

# Convergence

- Fitting the GAM involves an optimization
- By default this is REstricted Maximum Likelihood (REML) score
- Sometimes this can go wrong
- R will warn you!

# A model that converges

```
gam.check(dsm_tw_xy_depth)
```

Method: REML Optimizer: outer newton  
full convergence after 7 iterations.  
Gradient range [-3.468176e-05,1.090937e-05]  
(score 374.7249 & scale 4.172176).  
Hessian positive definite, eigenvalue range [1.179219,301.267].  
Model rank = 39 / 39

Basis dimension (k) checking results. Low p-value (k-index<1) may indicate that k is too low, especially if edf is close to k'.

	k'	edf	k-index	p-value
s(x,y)	29.00	11.11	0.65	<2e-16 ***
s(Depth)	9.00	3.84	0.81	0.33

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

# A bad model

```
Error in while (mean(ldxx/(ldxx + ldss)) > 0.4) { :  
  missing value where TRUE/FALSE needed  
In addition: Warning message:  
In sqrt(w) : NaNs produced  
Error in while (mean(ldxx/(ldxx + ldss)) > 0.4) { :  
  missing value where TRUE/FALSE needed
```

This is **rare**



# The Folk Theorem of Statistical Computing

“most statistical computational problems are due not to the algorithm being used but rather the model itself”

Andrew Gelman

# Basis size

# Basis size (k)

- Set k per term
- e.g.  $s(x, k=10)$  or  $s(x, y, k=100)$
- Penalty removes “extra” wigglyness
  - *up to a point!*
- (But computation is slower with bigger k)

# Checking basis size

```
gam.check(dsm_x_tw)
```

Method: REML Optimizer: outer newton  
full convergence after 7 iterations.  
Gradient range [-3.08755e-06,4.928064e-07]  
(score 409.936 & scale 6.041307).  
Hessian positive definite, eigenvalue range [0.7645492,302.127].  
Model rank = 10 / 10

Basis dimension (k) checking results. Low p-value (k-index<1) may indicate that k is too low, especially if edf is close to k'.

	k'	edf	k-index	p-value
s(x)	9.00	4.96	0.76	0.44

# Increasing basis size

```
dsm_x_tw_k <- dsm(count~s(x, k=20), ddf.obj=df,  
                  segment.data=segs, observation.data=obs,  
                  family=tw())  
gam.check(dsm_x_tw_k)
```

Method: REML Optimizer: outer newton

full convergence after 7 iterations.

Gradient range [-2.301238e-08,3.930667e-09]

(score 409.9245 & scale 6.033913).

Hessian positive definite, eigenvalue range [0.7678456,302.0336].

Model rank = 20 / 20

Basis dimension (k) checking results. Low p-value (k-index<1) may indicate that k is too low, especially if edf is close to k'.

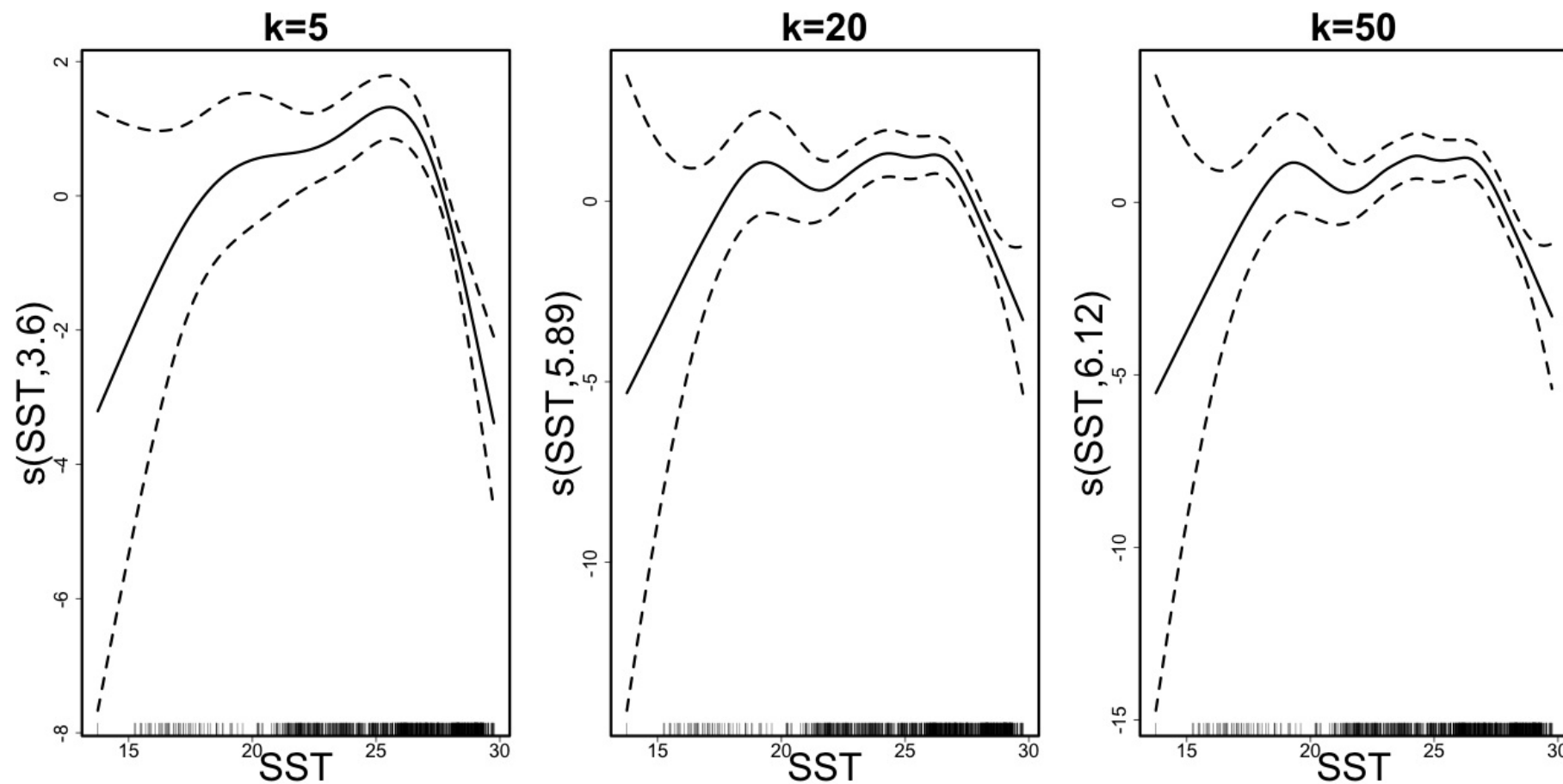
	k'	edf	k-index	p-value
s(x)	19.00	5.25	0.76	0.39

# Sometimes basis size isn't the issue...

- Generally, double  $k$  and see what happens
- Didn't increase the EDF much here
- Other things can cause low “p-value” and “k-index”
- Increasing  $k$  can cause problems (nullspace)

# k is a maximum

- (Usually) Don't need to worry about things being too wiggly
- k gives the maximum complexity
- Penalty deals with the rest



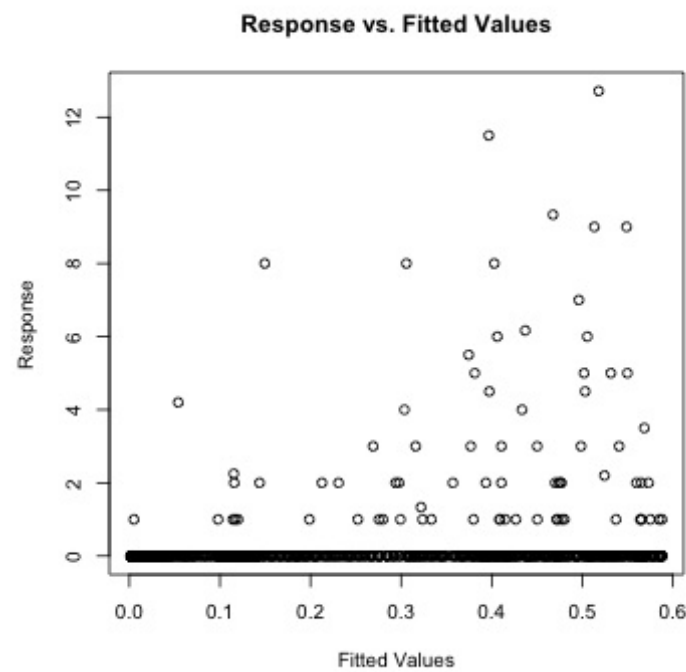
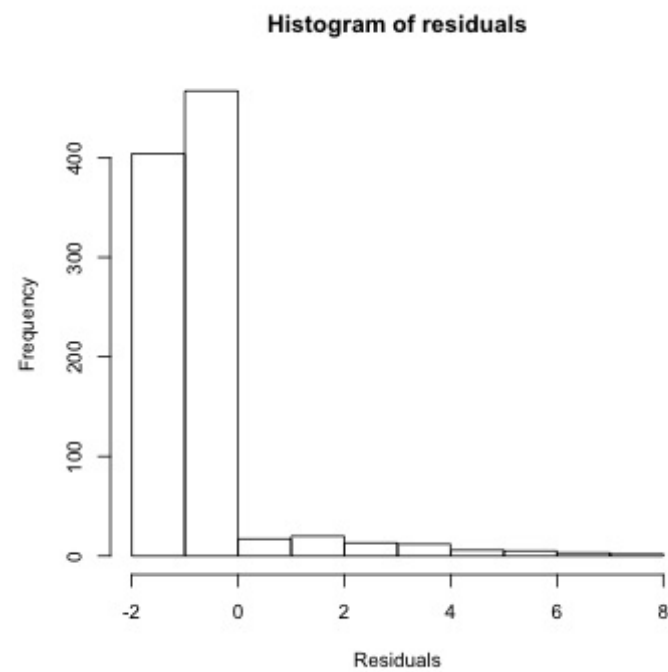
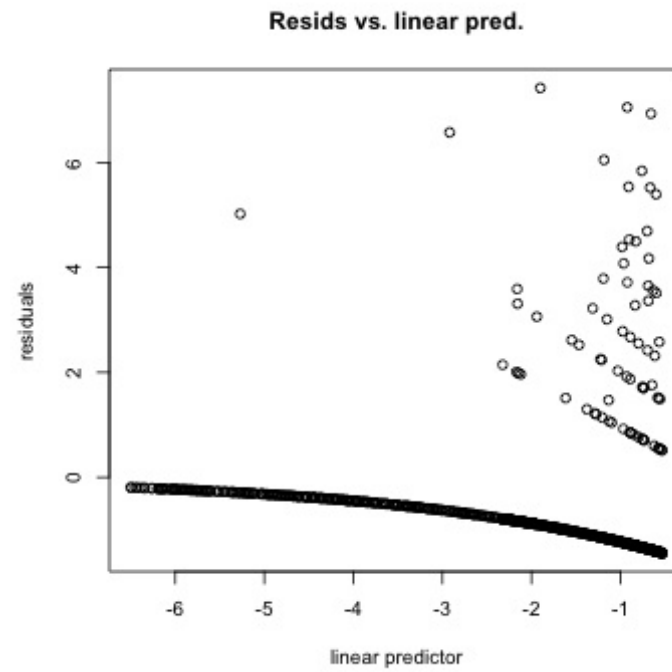
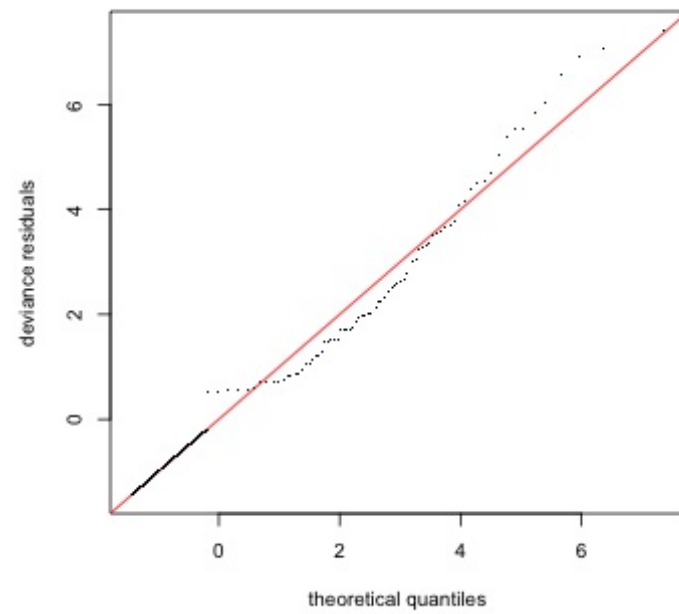
# Residuals



# What are residuals?

- Generally residuals = observed value - fitted value
- BUT hard to see patterns in these “raw” residuals
- Need to standardise  $\Rightarrow$  **deviance residuals**
- Residual sum of squares  $\Rightarrow$  linear model
  - deviance  $\Rightarrow$  GAM
- Expect these residuals  $\sim N(0, 1)$

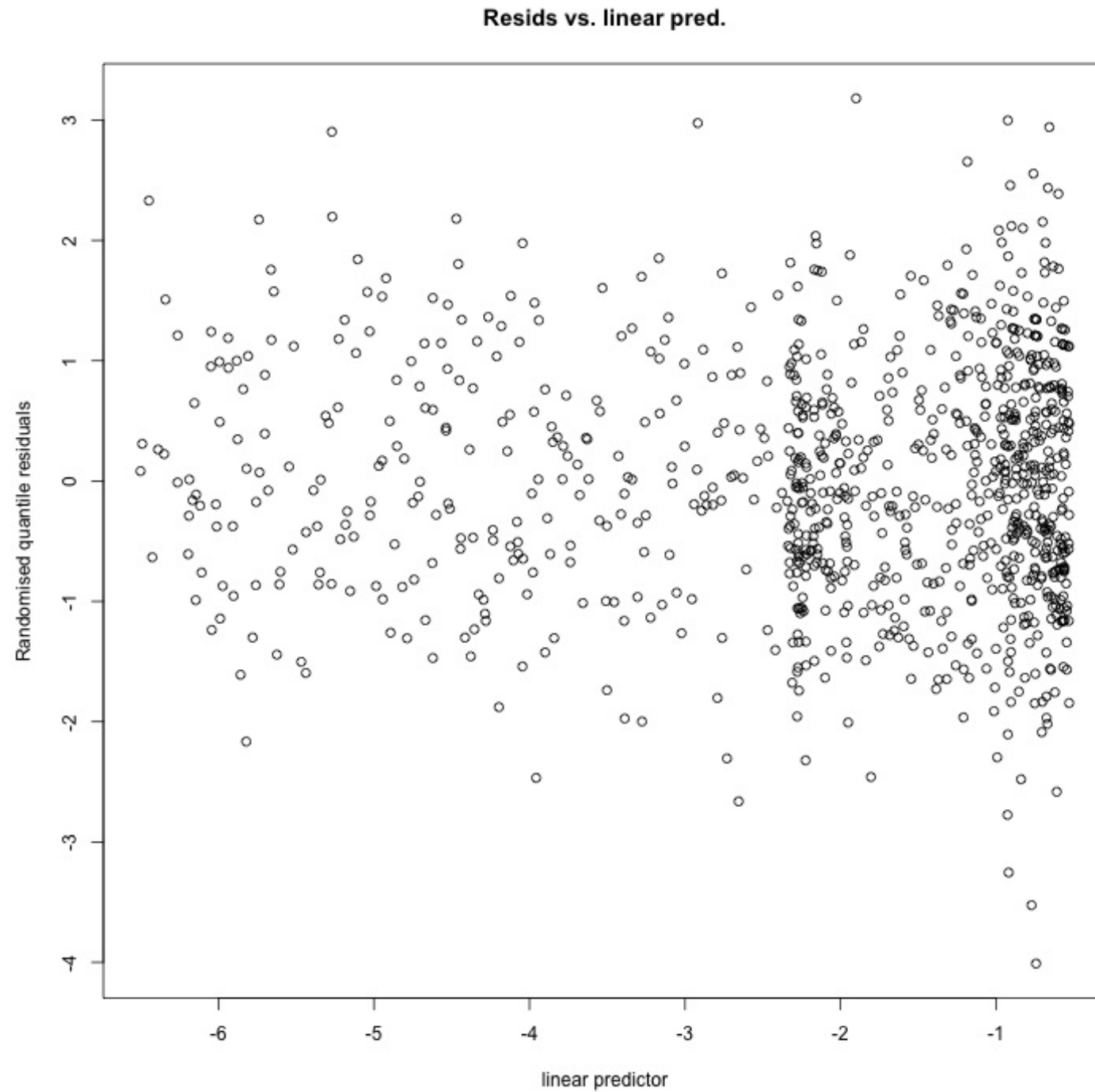
# Residual checking



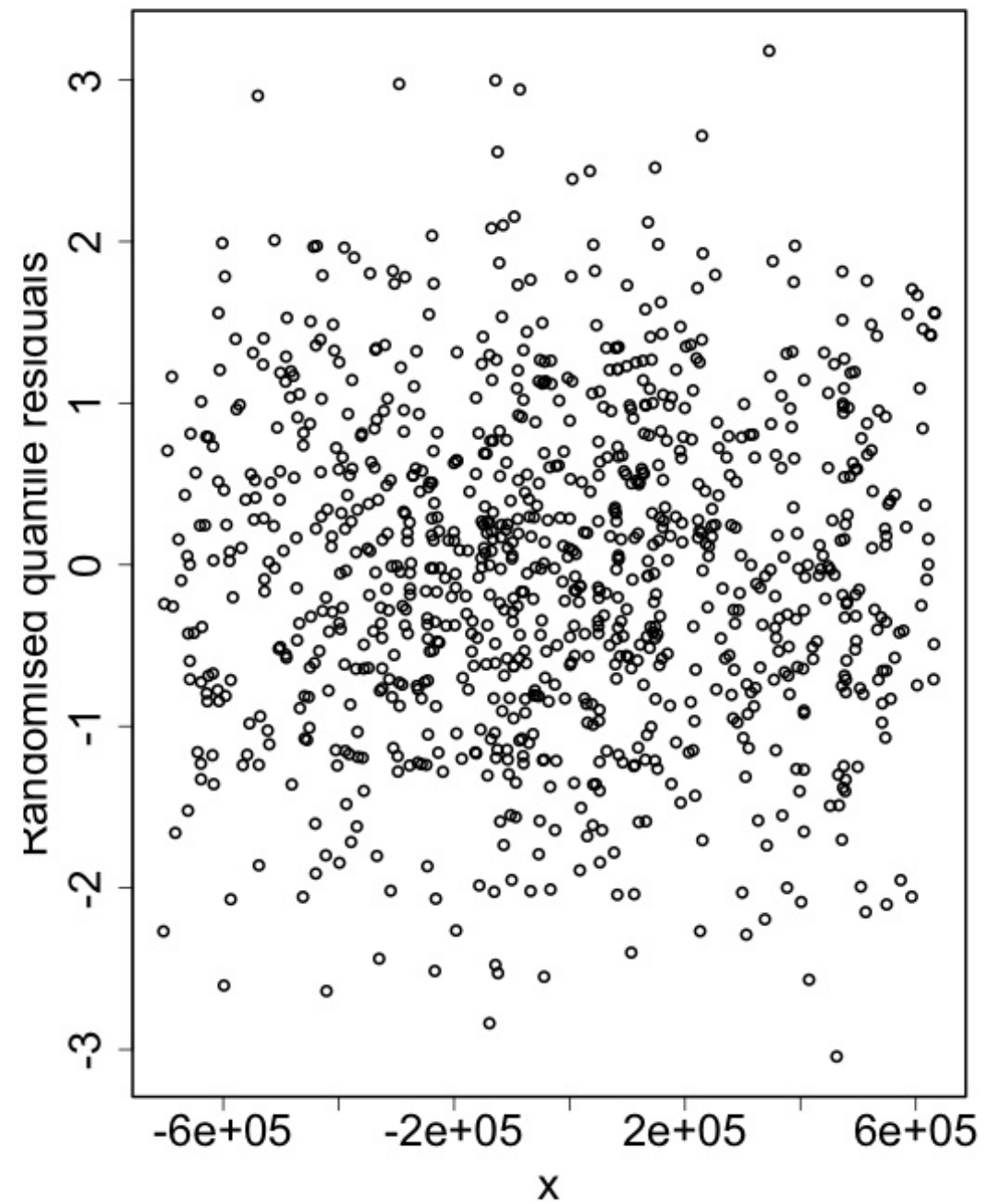
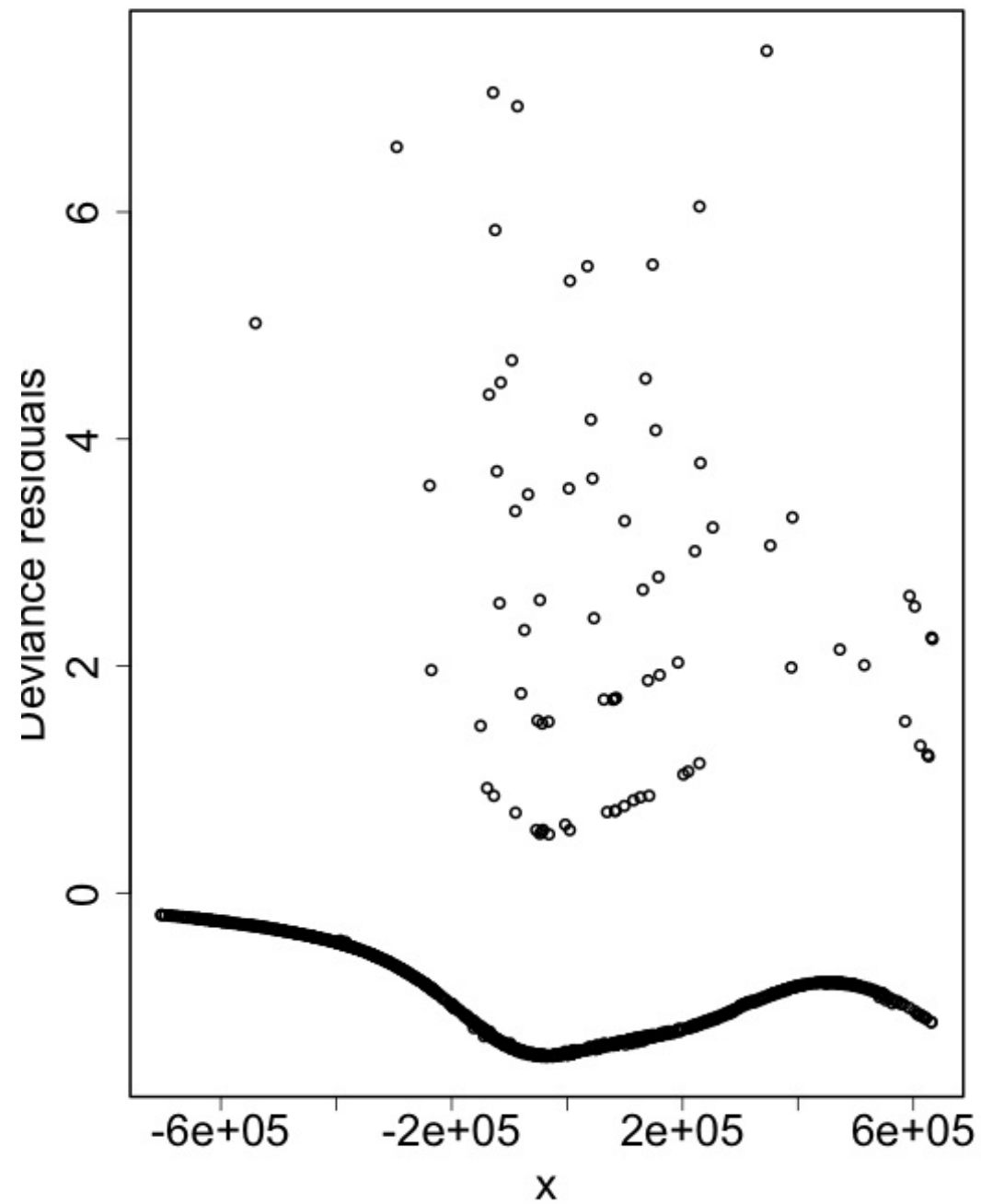
# Shortcomings

- `gam.check` can be helpful
- “Resids vs. linear pred” is victim of artifacts
- Need an alternative
- “Randomised quantile residuals” (*experimental*)
  - `rqqam.check`
  - Exactly normal residuals

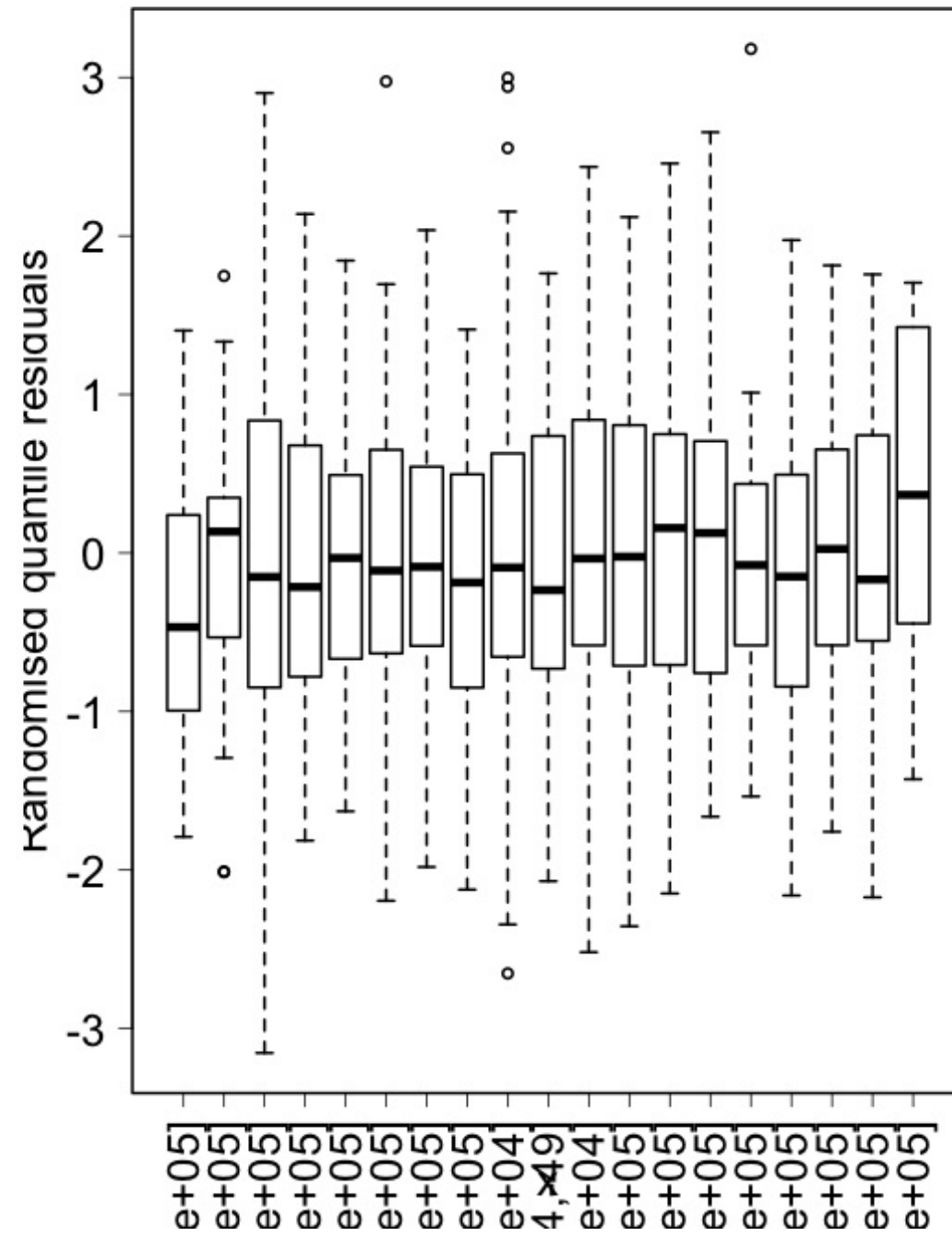
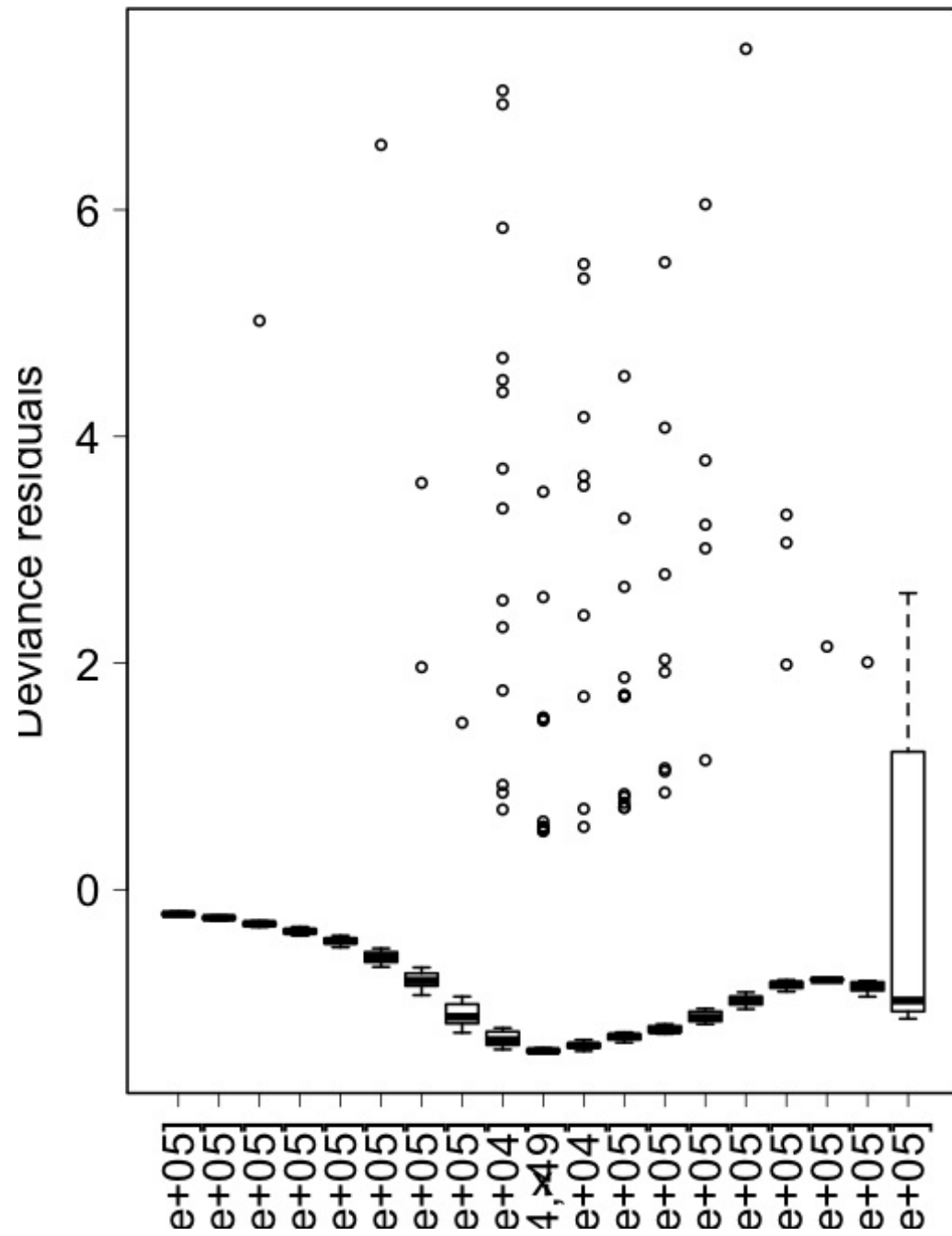
# Randomised quantile residuals



# Residuals vs. covariates

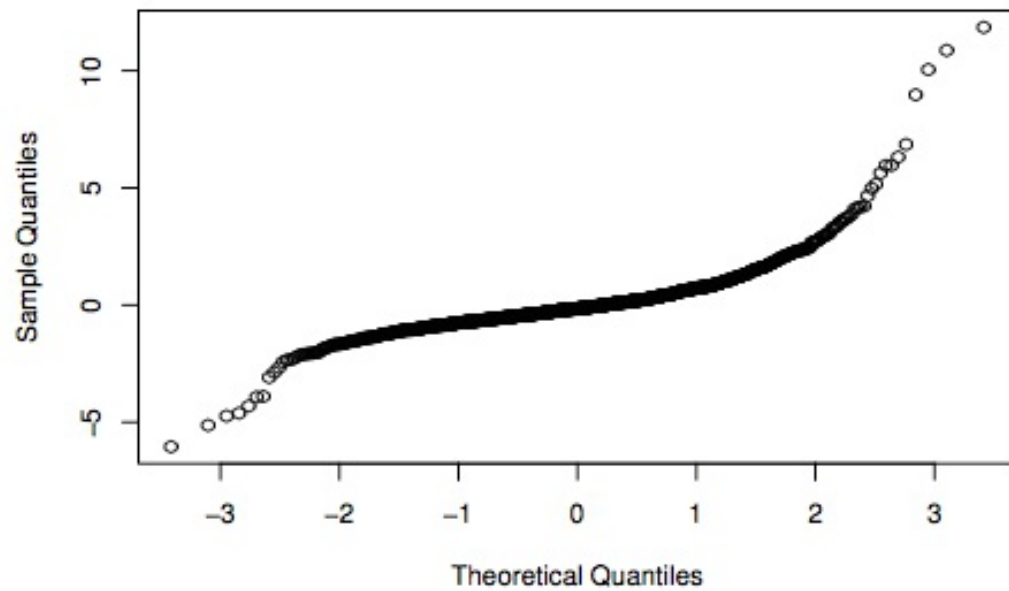


# Residuals vs. covariates (boxplots)

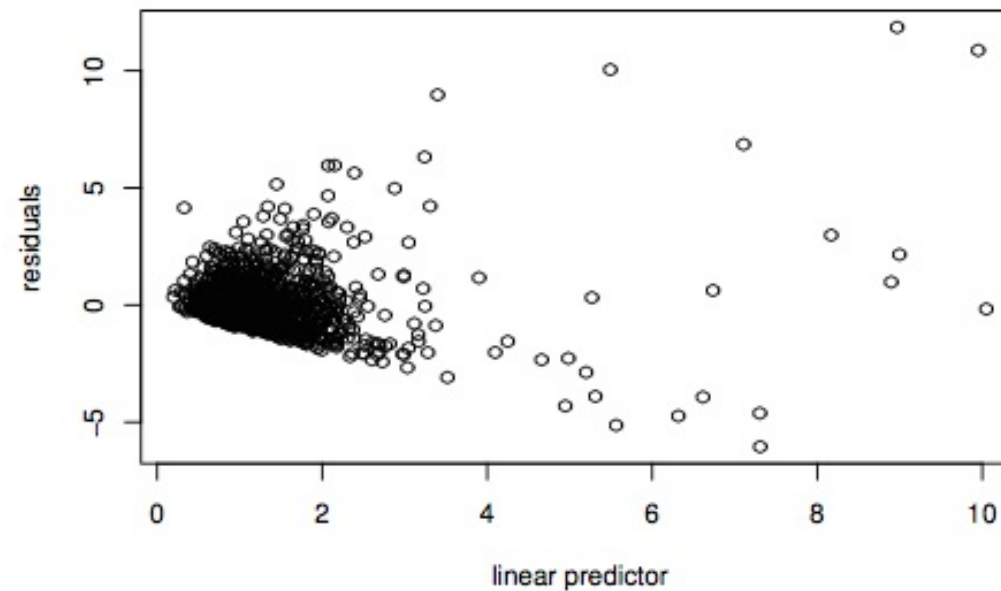


# Example of "bad" plots

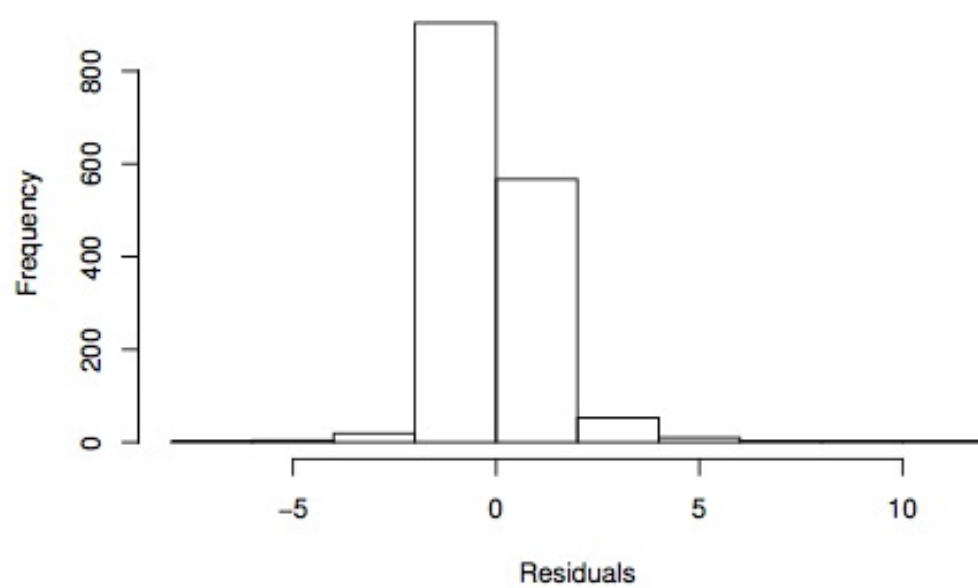
Normal Q-Q Plot



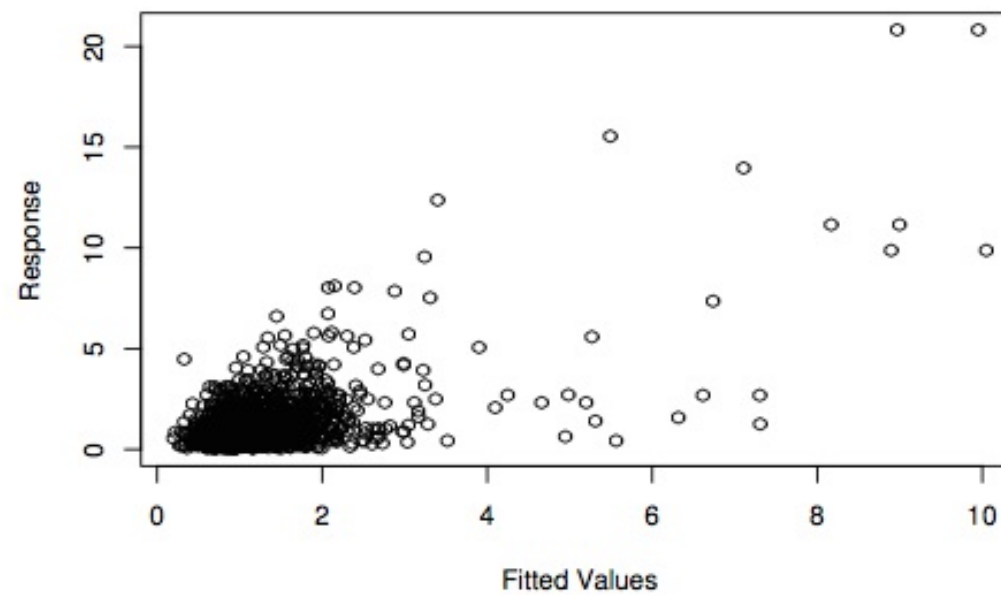
Resids vs. linear pred.



Histogram of residuals

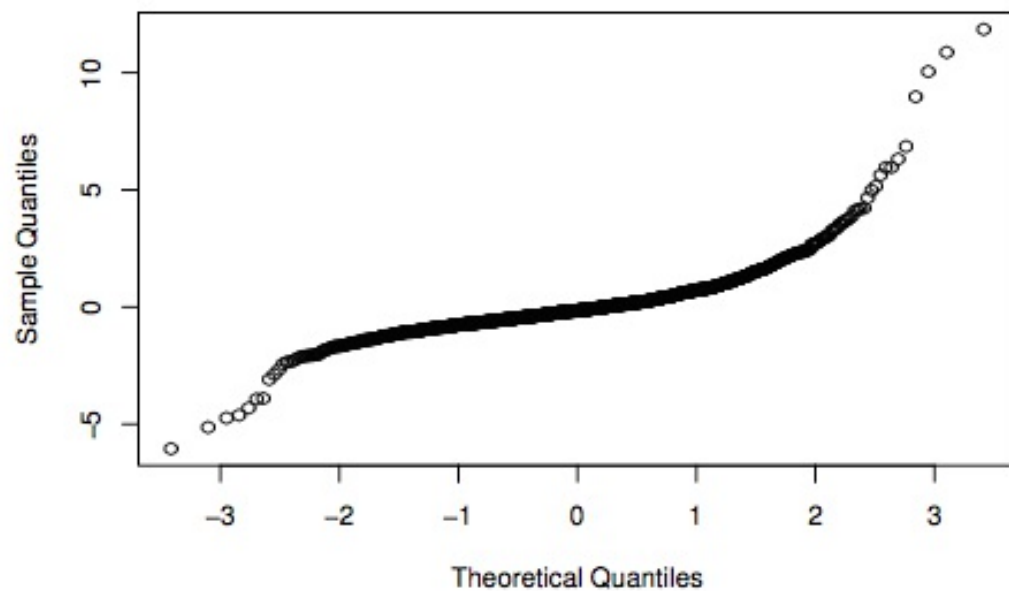


Response vs. Fitted Values

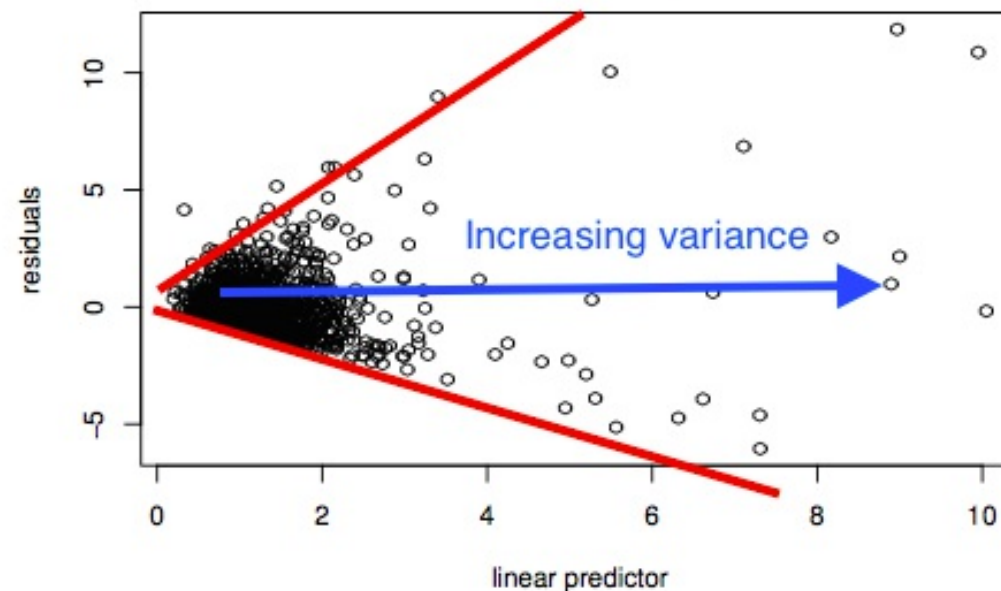


# Example of "bad" plots

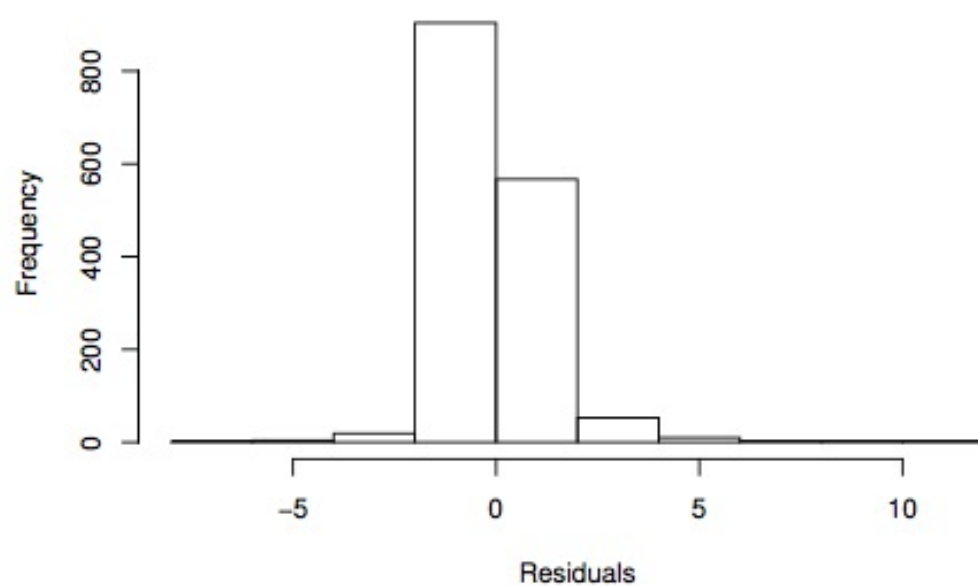
Normal Q-Q Plot



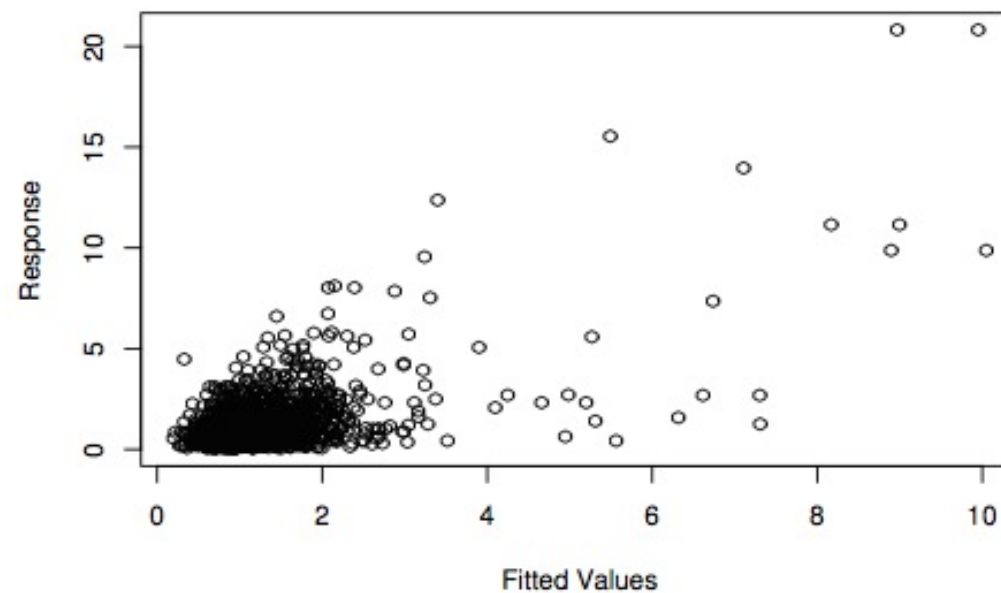
Resids vs. linear pred.



Histogram of residuals



Response vs. Fitted Values





# Residual checks

- Looking for patterns (not artifacts)
- This can be tricky
- Need to use a mixture of techniques
- Cycle through checks, make changes recheck
- Each dataset is different

# Summary

- Convergence
  - Rarely an issue
  - Check your thinking about the model
- Basis size
  - $k$  is a maximum
  - Double and see what happens
- Residuals
  - Deviance and randomised quantile
  - check for artifacts
- `gam.check` is your friend