Purpose

This document extends the analysis of spatial autocorrelation from the document titled Testing_spatial_subsampling.Rmd to the statistical analysis of spatial correlations with semivariograms. Instead of just visually assessing the spatial autocorrelation in the reconstructions, we can fit empirical variograms to the reconstructions at each time step for each taxon.

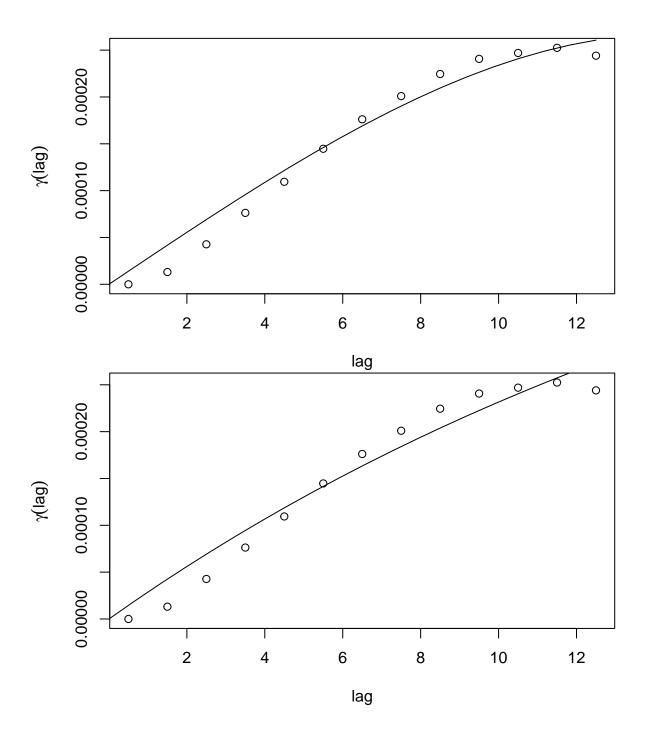
Semivariances

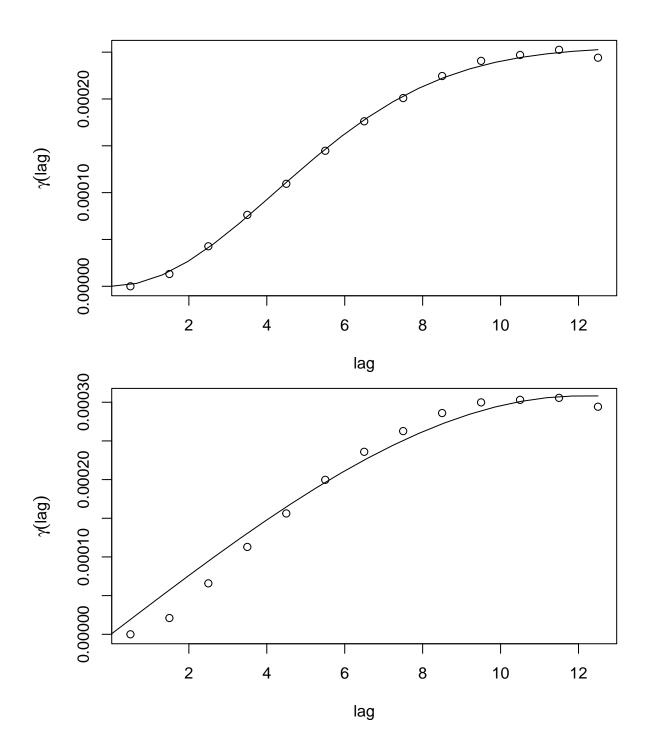
First, I used the semivariogram function from the gstat package to fit spherical, exponential, and Gaussian variograms to each taxon at each time period. This function returns the semivariances of the relative abundances, as well as the sill, nugget, and range parameter estimates. The range is an estimate of how far away the grid cells need to be to be considered independent. The ratio of nugget:sill is an indicator of the strength of spatial dependence in the relative abundance, with high ratio (>75%) indicative of low spatial dependence.

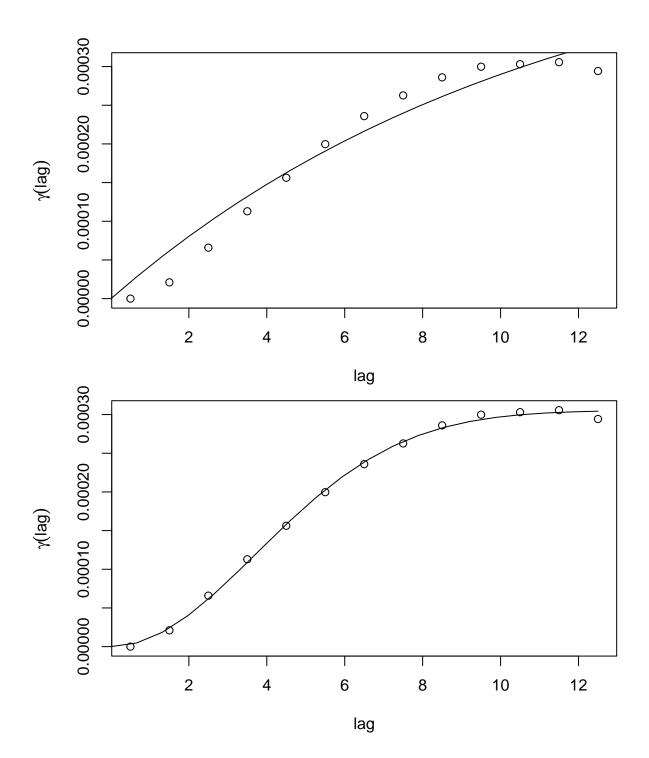
Since I replaced the x-y distances with cell number, the distance and range can be interpreted as number of cells in the following analysis.

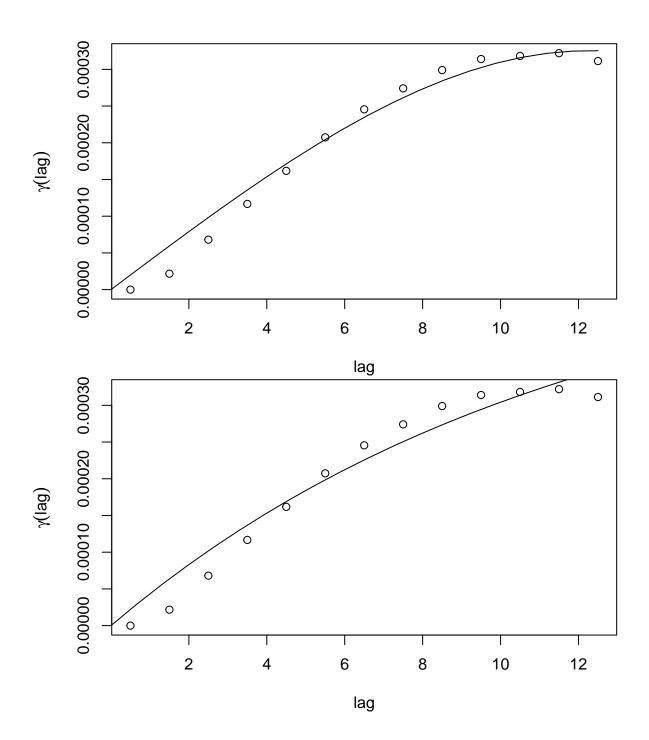
My expectation is that of these three options, the Guassian variogram will be the best fit because this is the closest to the nugget used for model fitting, but the actual nugget is an inverse power law kernel, which I believe is the same as the powered exponential kernel.

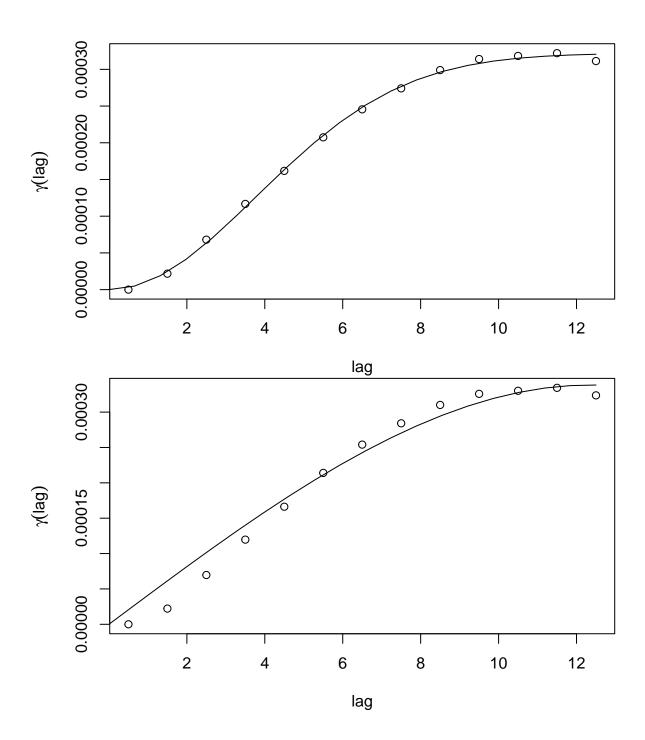
```
ash_spherical_params <- matrix(, nrow = length(time), ncol = 4)
ash_exponential_params <- matrix(, nrow = length(time), ncol = 4)
ash_gaussian_params <- matrix(, nrow = length(time), ncol = 4)</pre>
for(i in 1:length(time)){
  t <- time[i]
  sub <- dplyr::filter(ash_melt,</pre>
                        time == t,
                        !is.na(ash))
  sv <- geostats::semivariogram(x = sub$x, y = sub$y, z = sub$ash,</pre>
                                 model = 'spherical')
  ash_spherical_params[i,1] <- t
  ash_spherical_params[i,2:4] <- sv$snr
  sv <- geostats::semivariogram(x = sub$x, y = sub$y, z = sub$ash,</pre>
                                 model = 'exponential')
  ash exponential params[i,1] <- t
  ash_exponential_params[i,2:4] <- sv$snr
  sv <- geostats::semivariogram(x = sub$x, y = sub$y, z = sub$ash,</pre>
                                 model = 'gaussian')
  ash_gaussian_params[i,1] <- t
  ash_gaussian_params[i,2:4] <- sv$snr
}
```

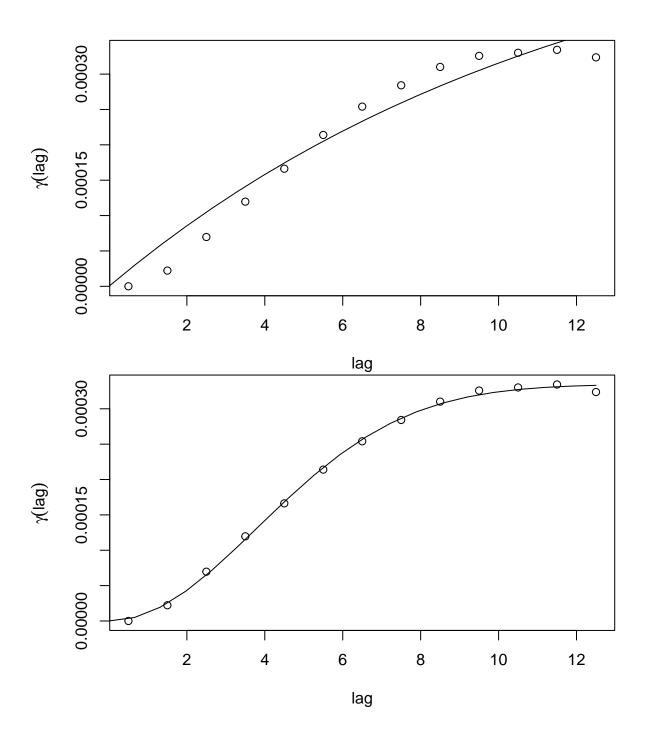


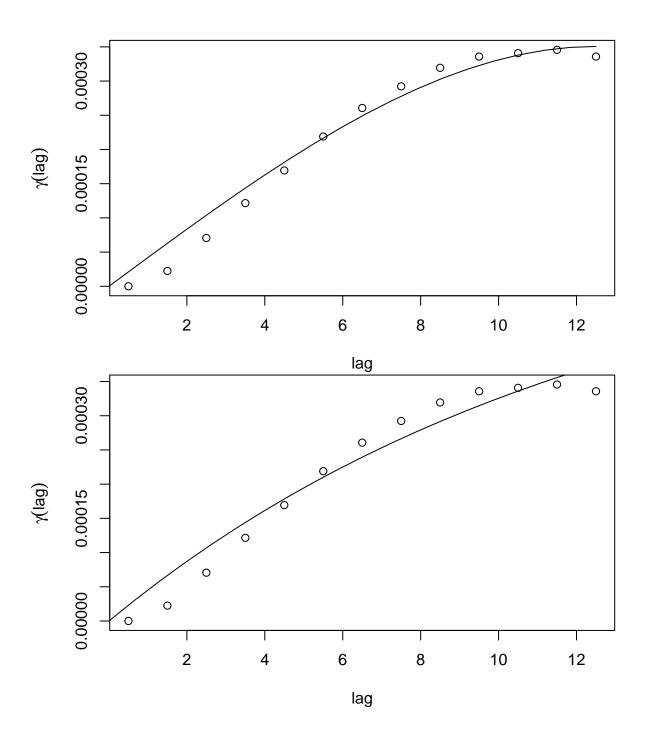


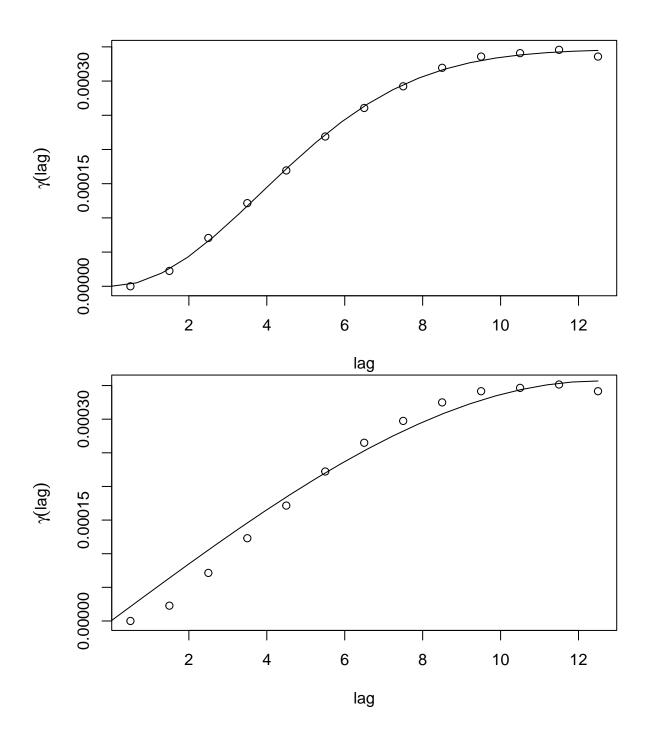


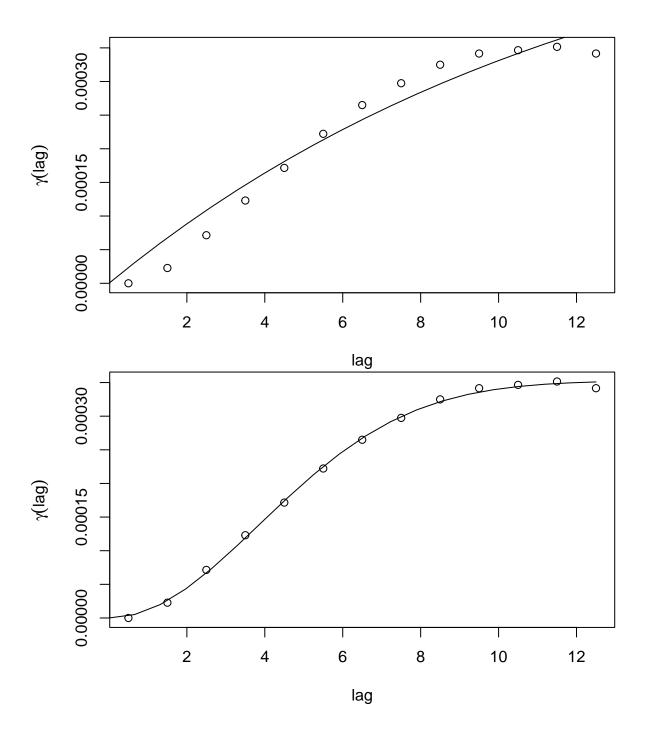


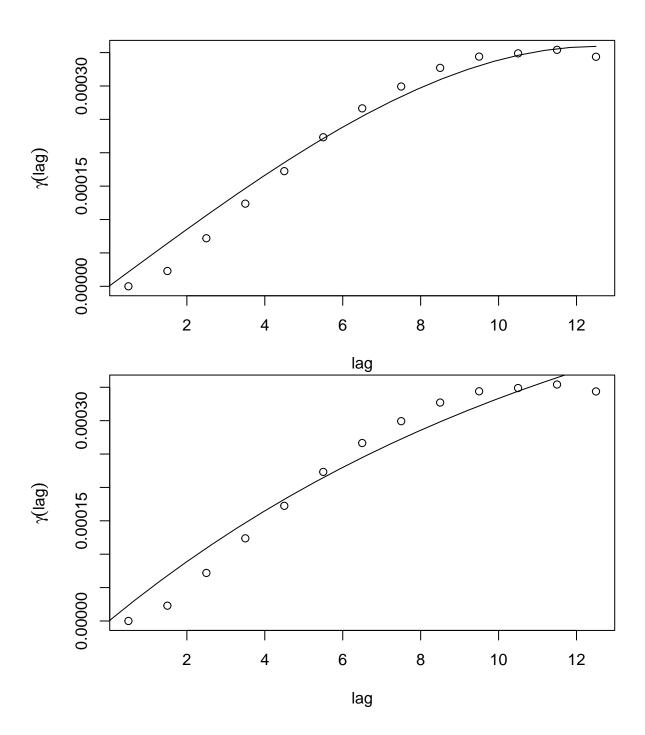


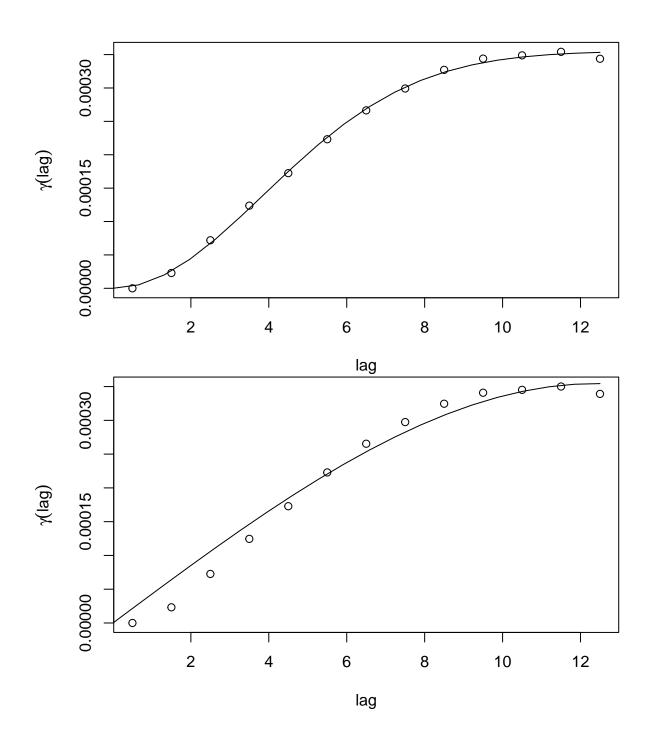


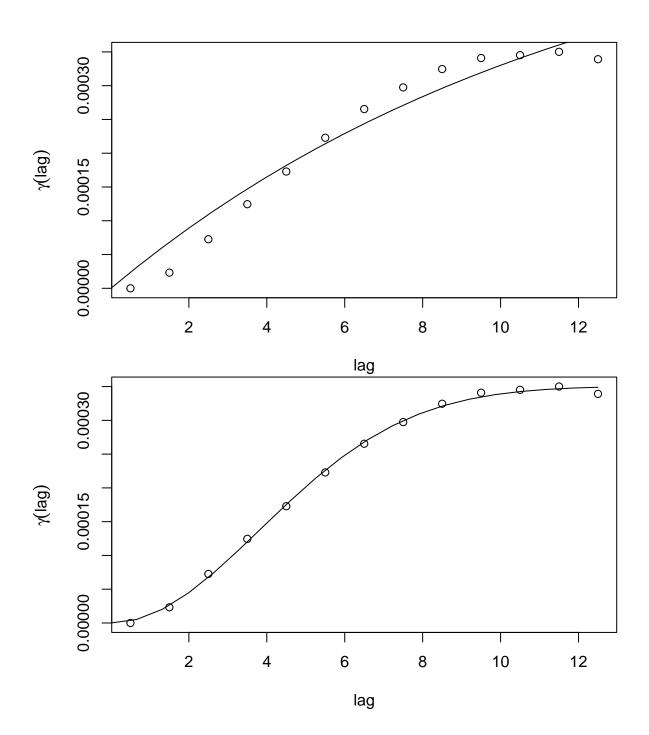


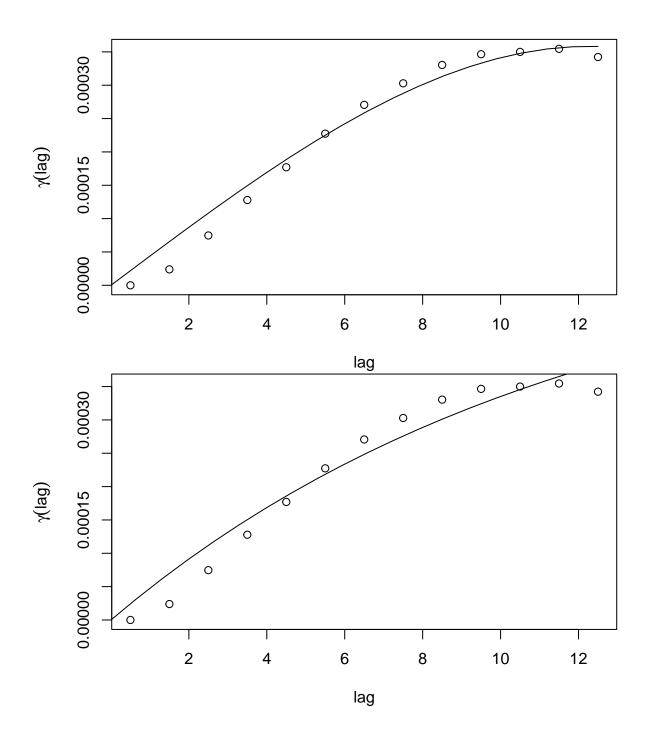


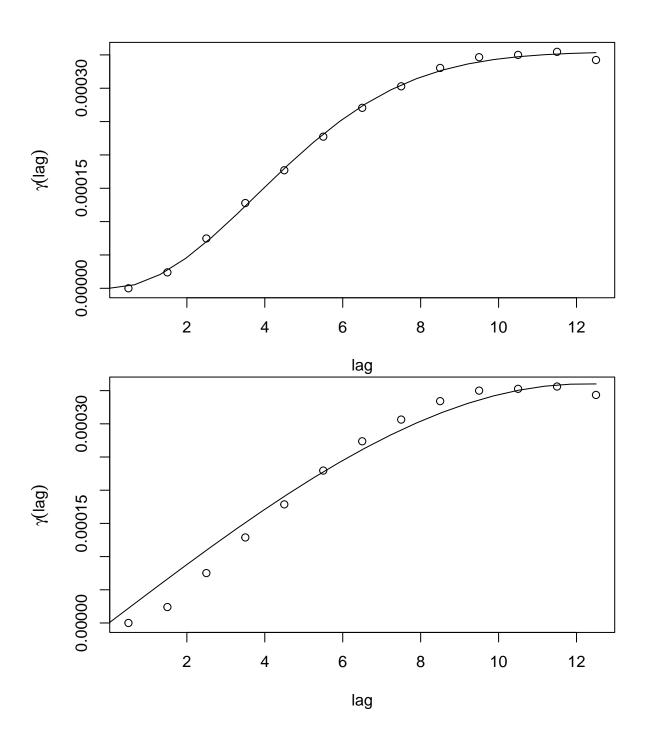


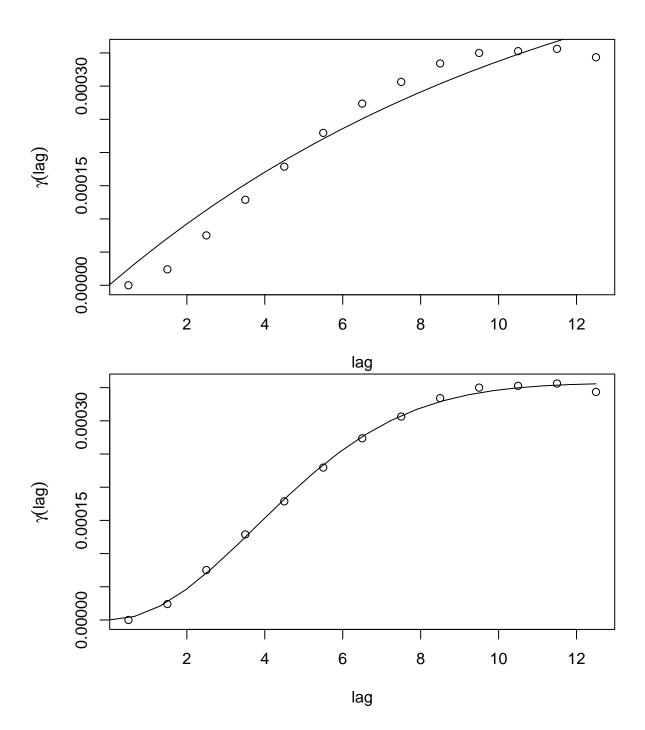


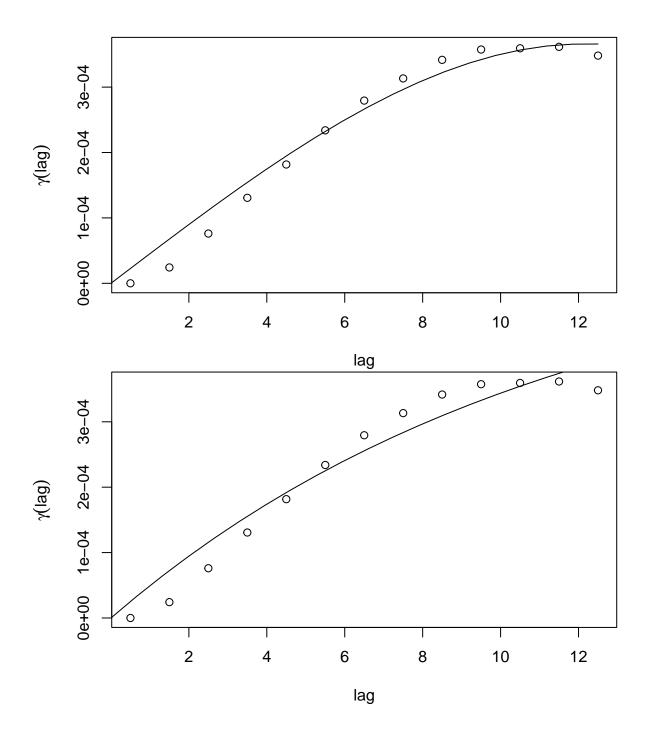


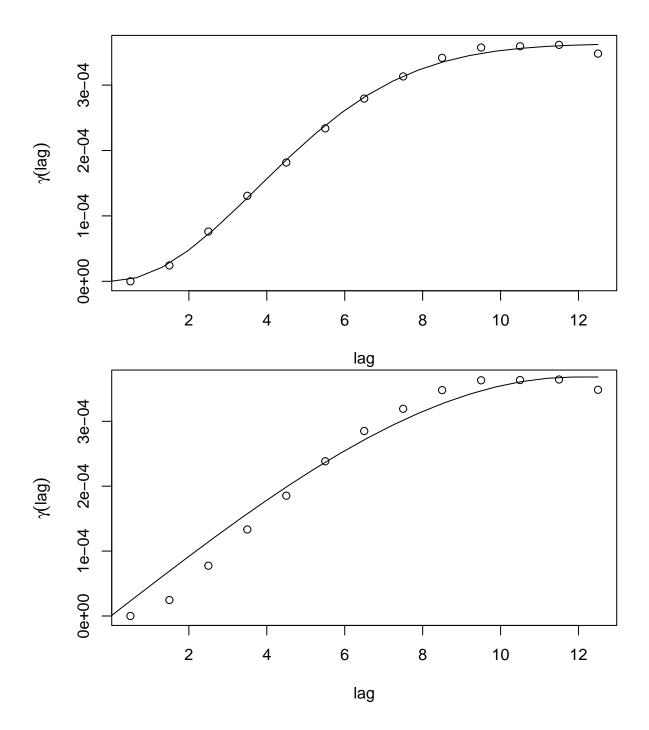


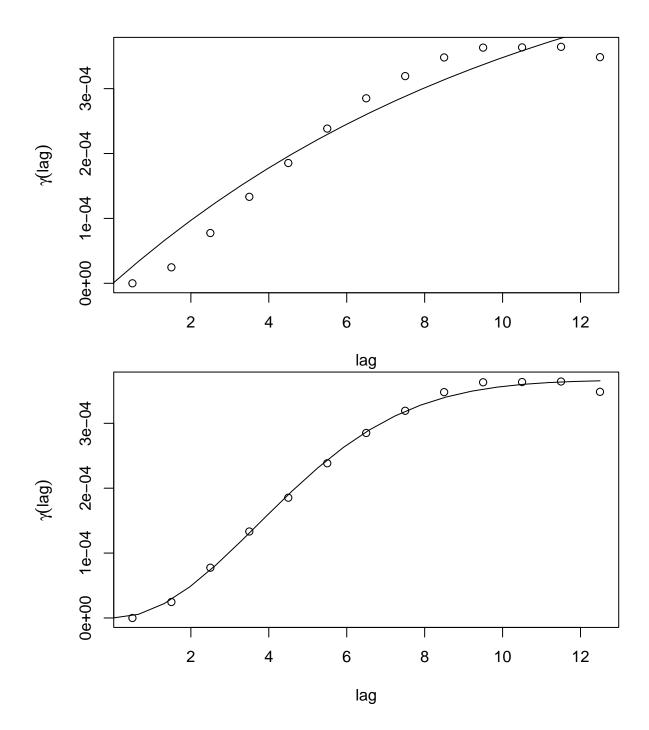


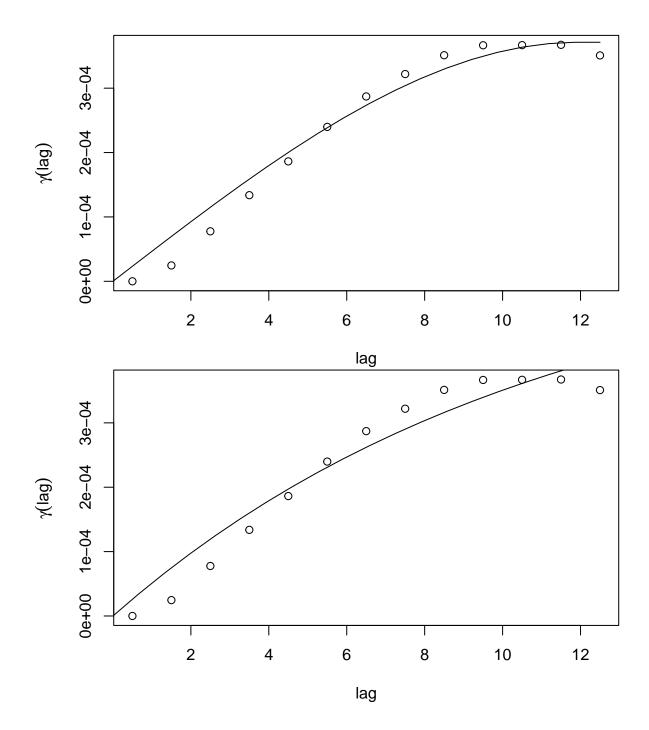


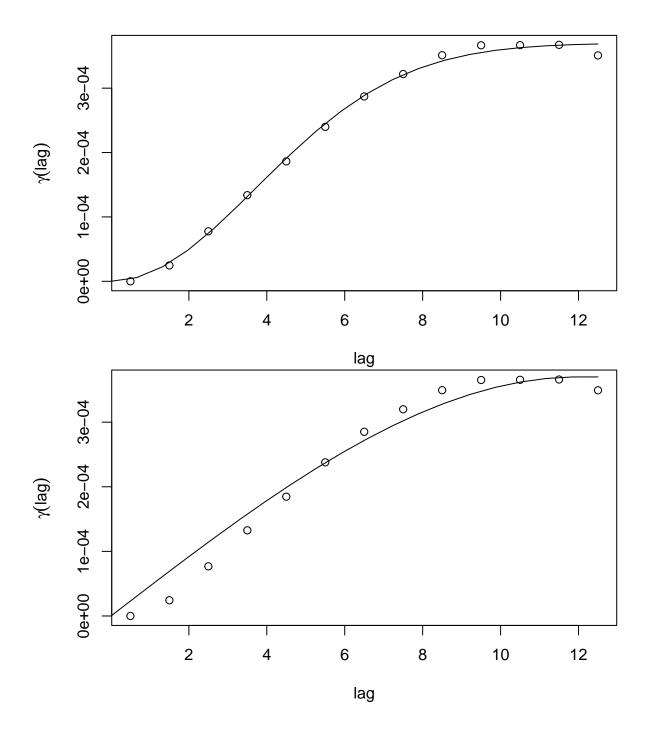


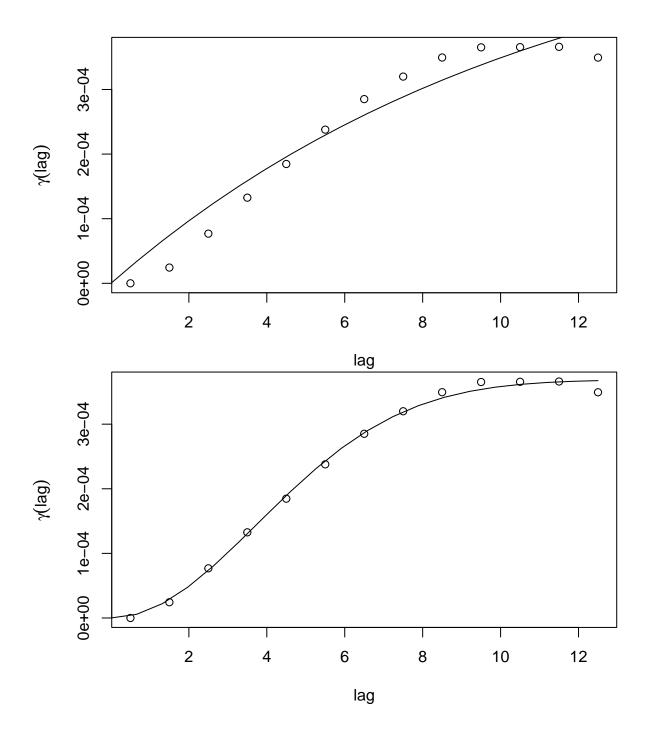


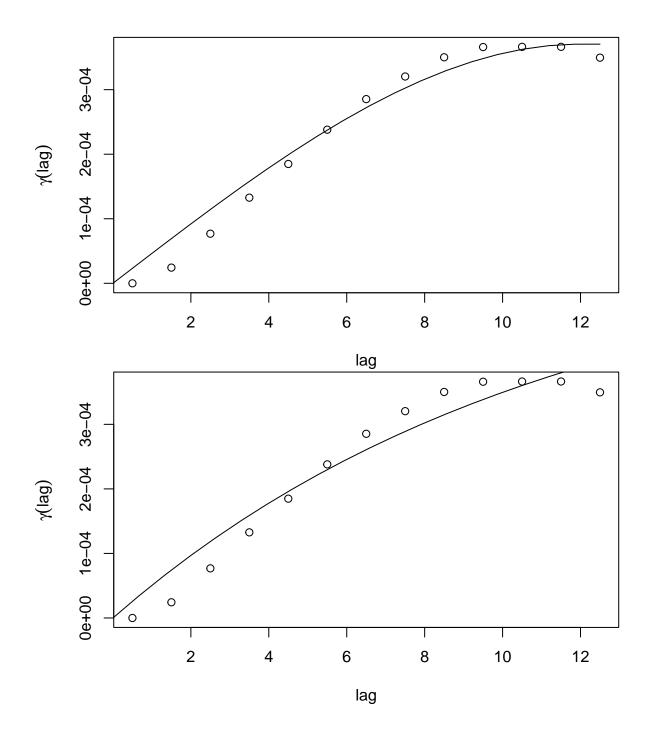


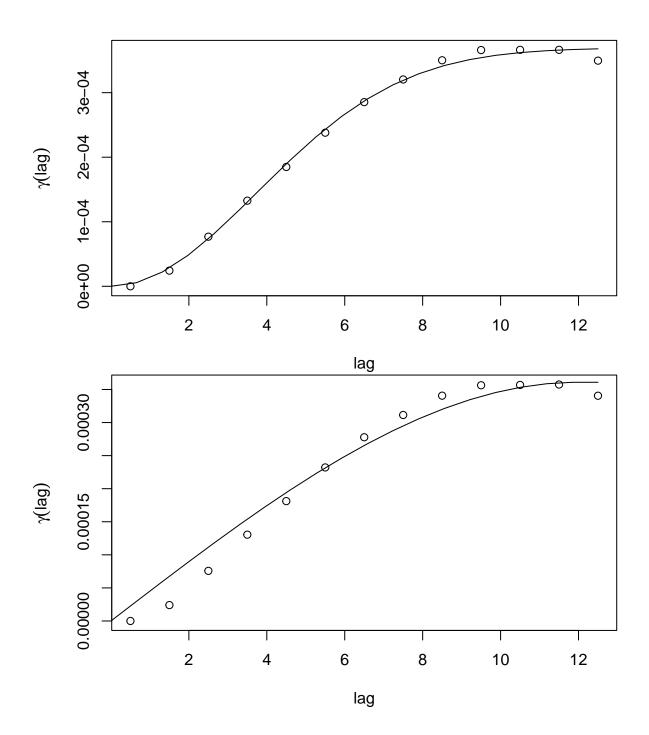


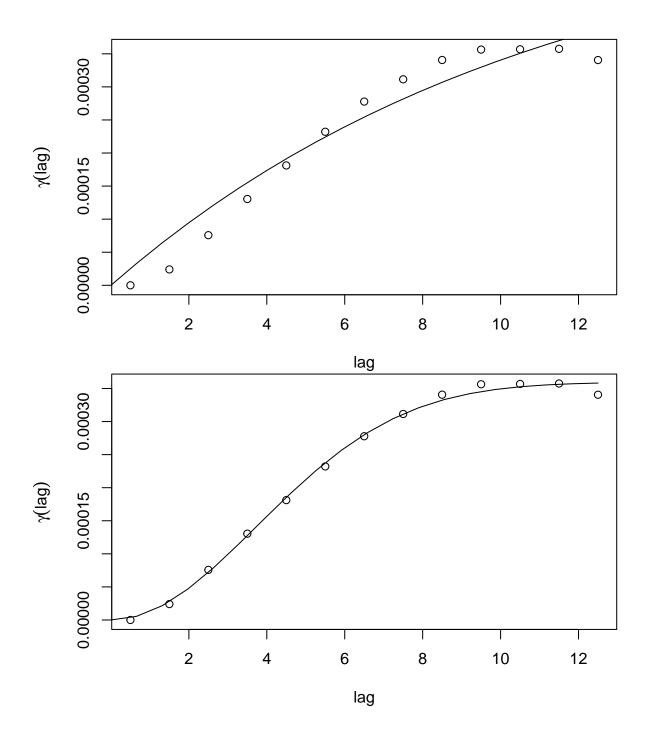


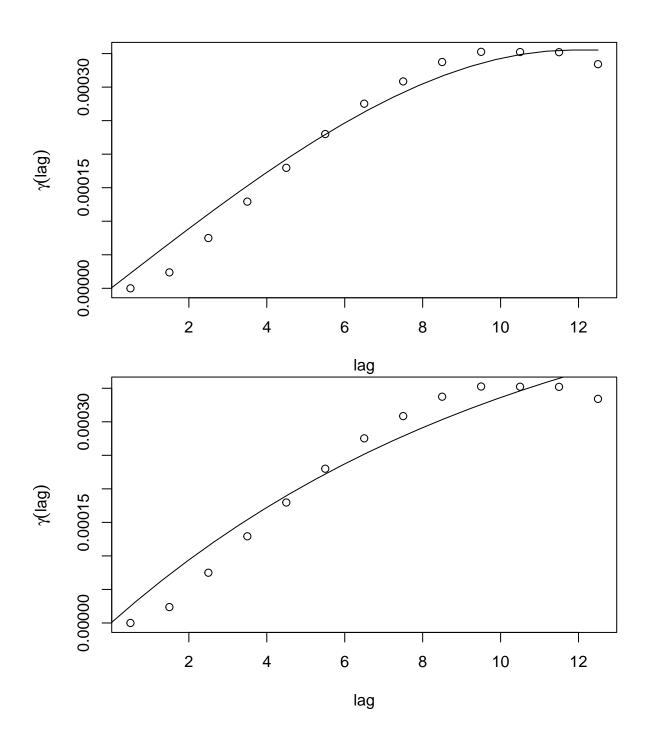


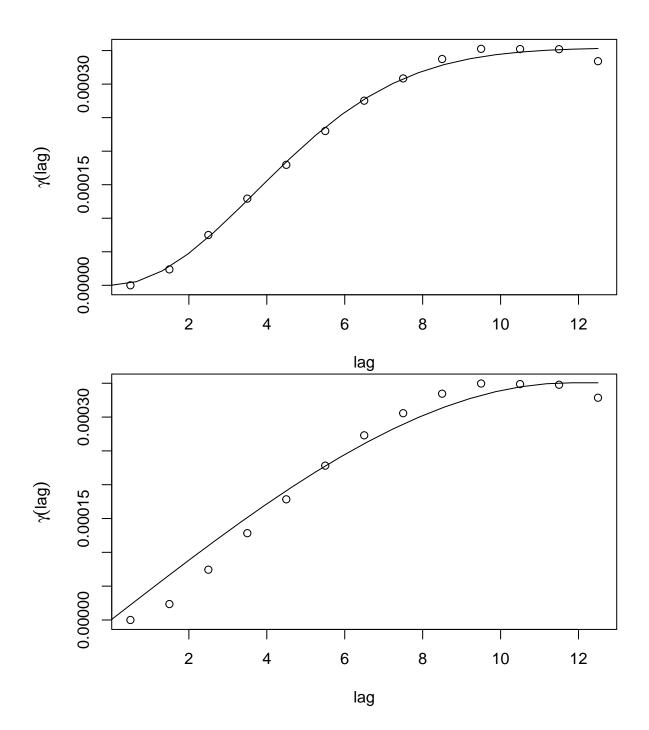


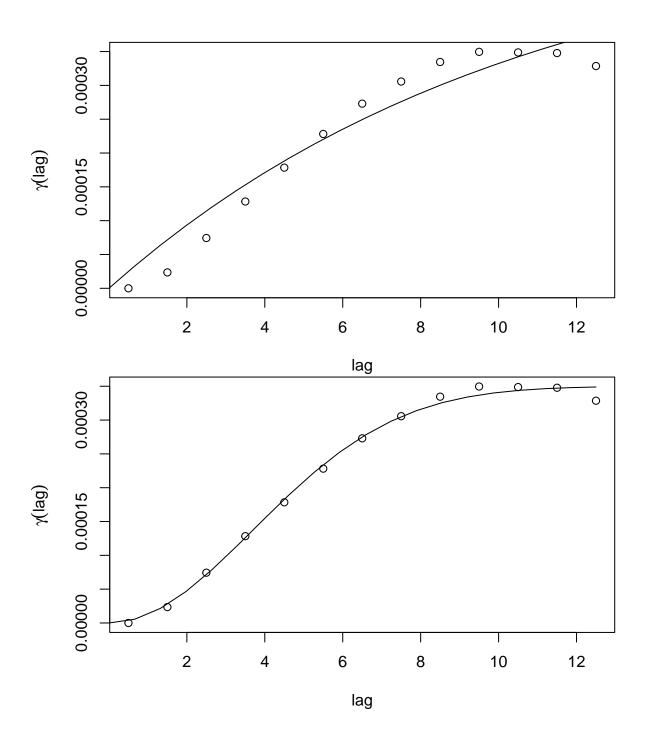


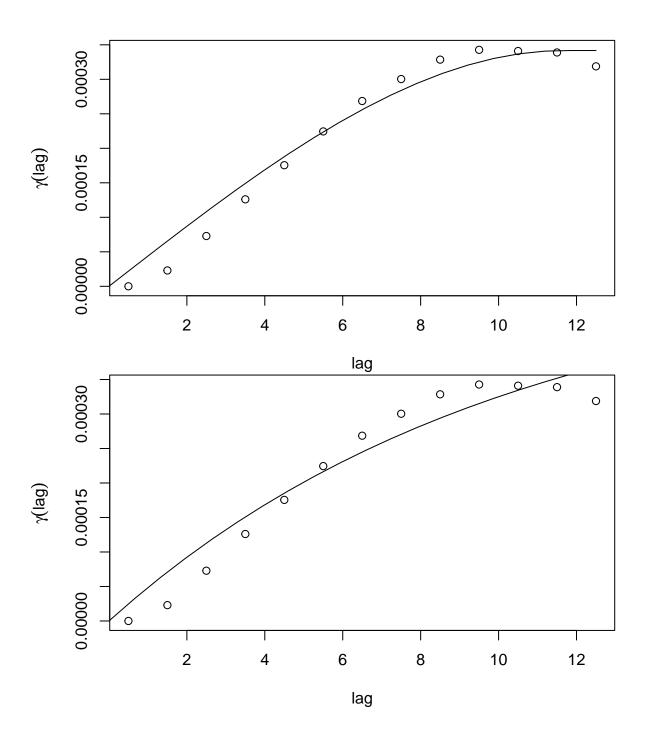


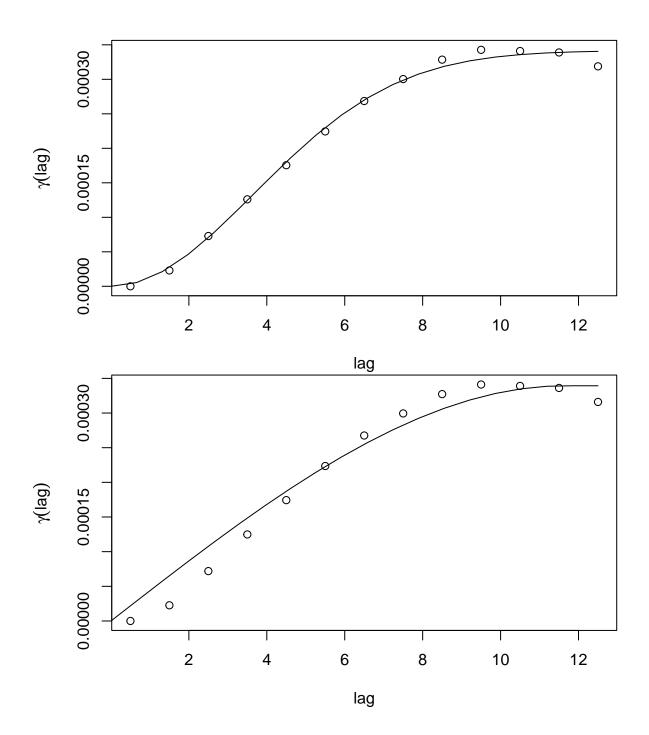


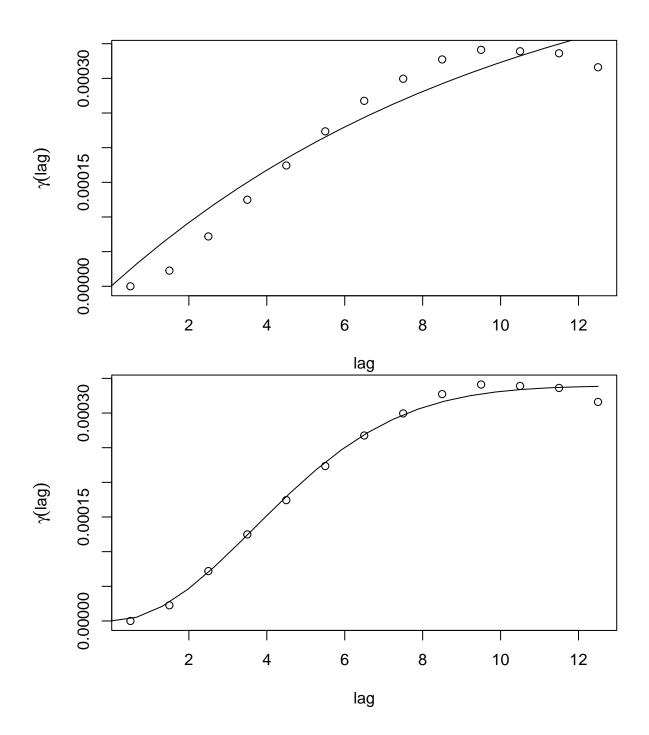


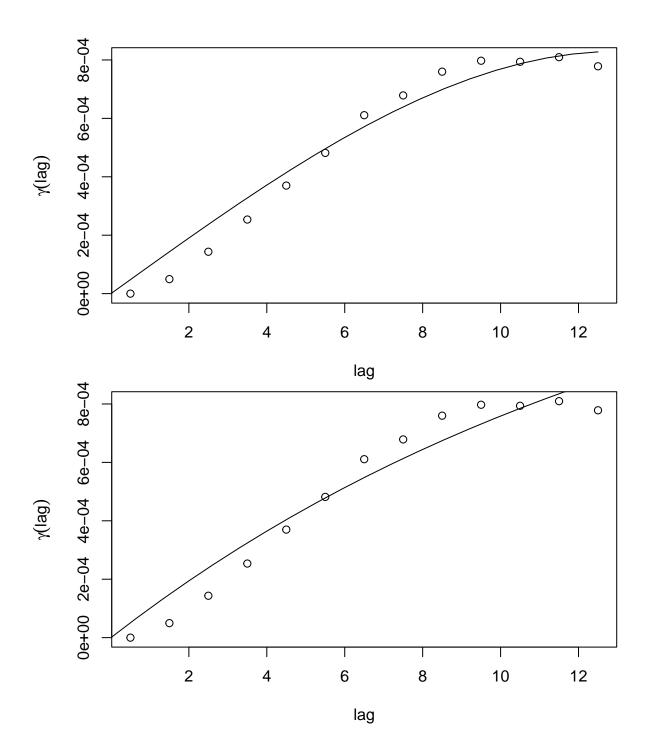


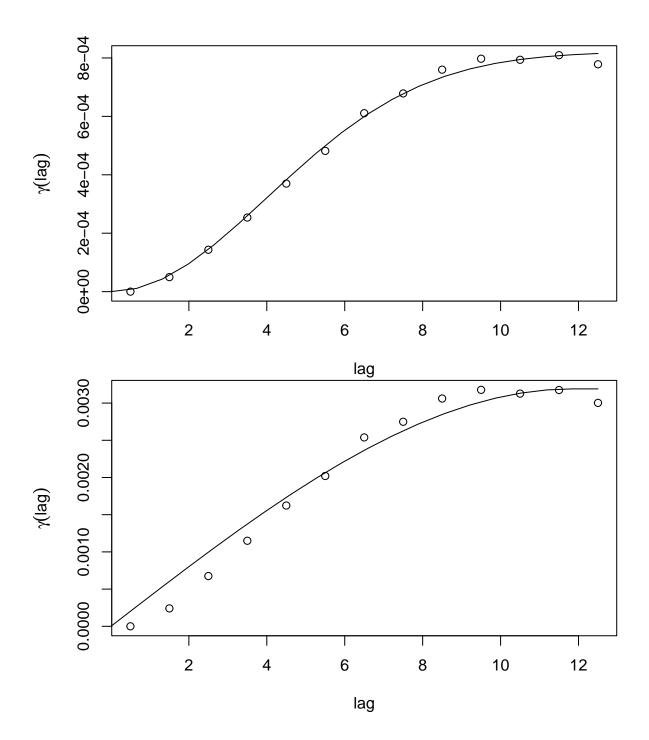


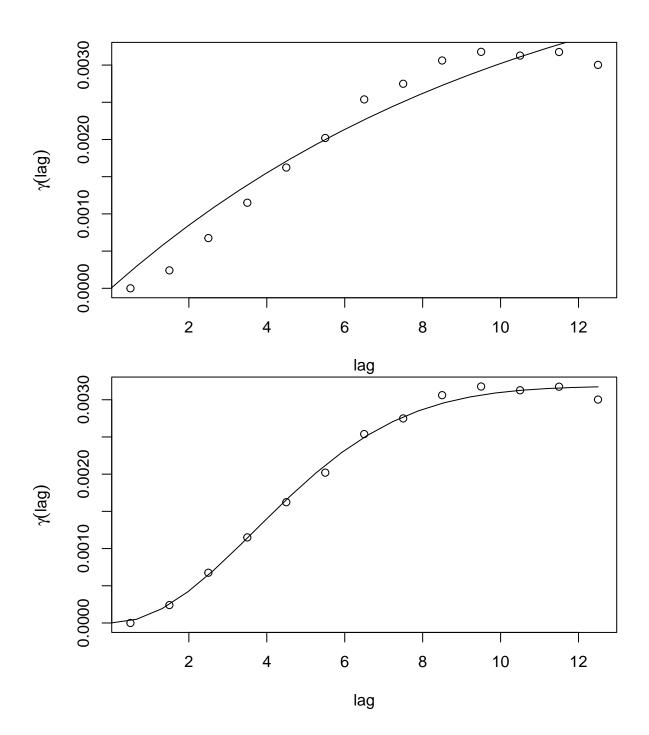


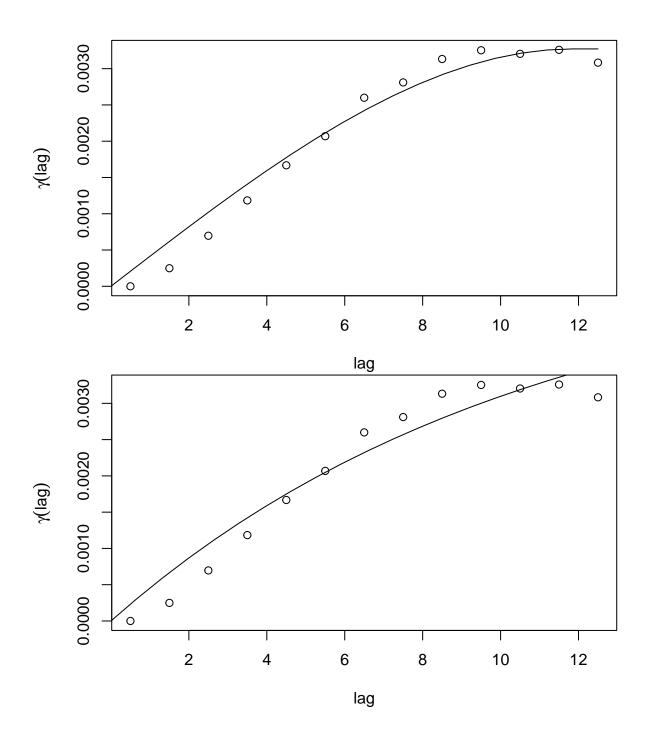


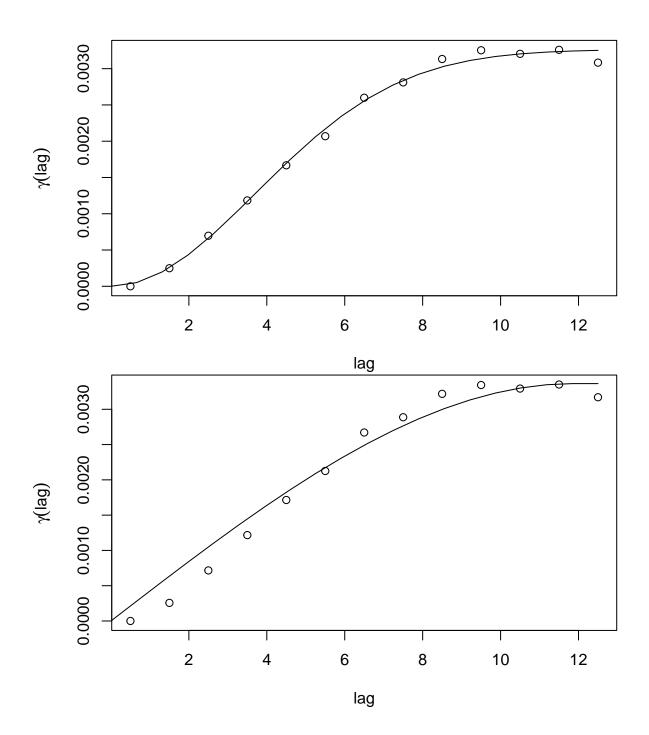


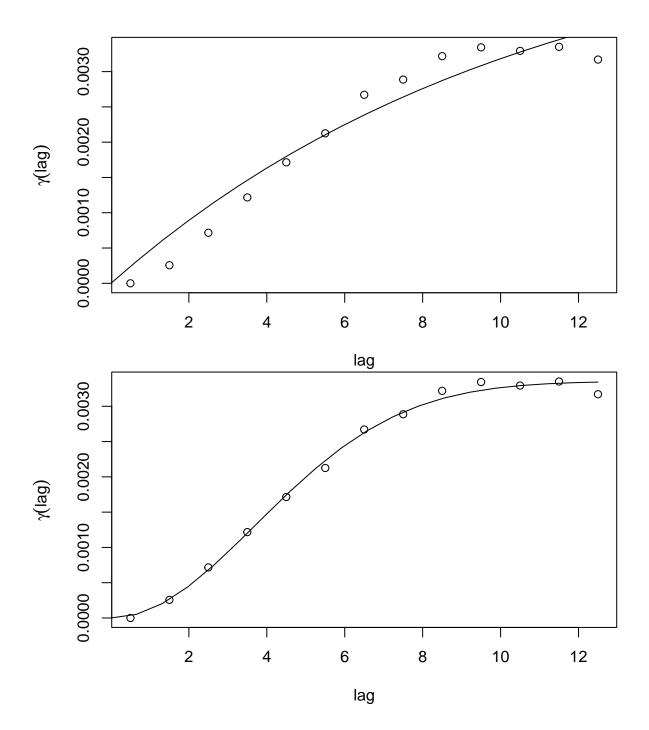


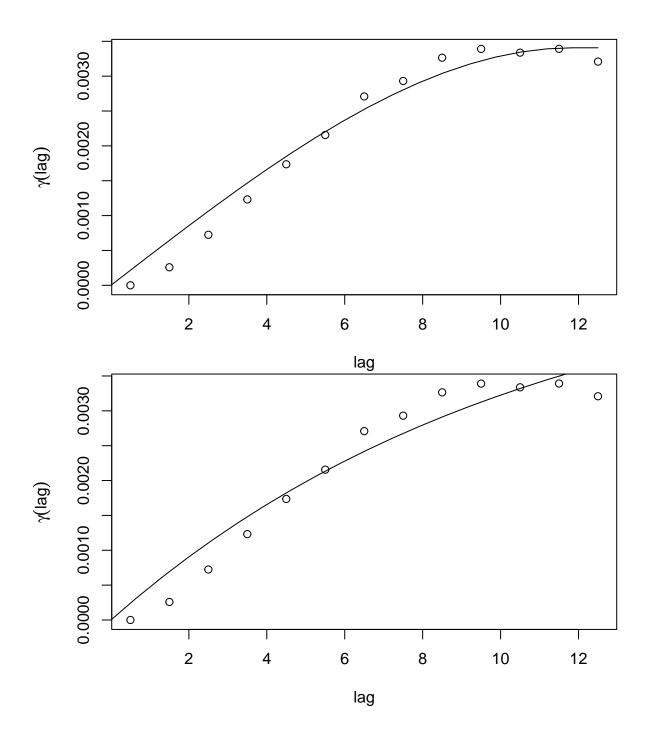


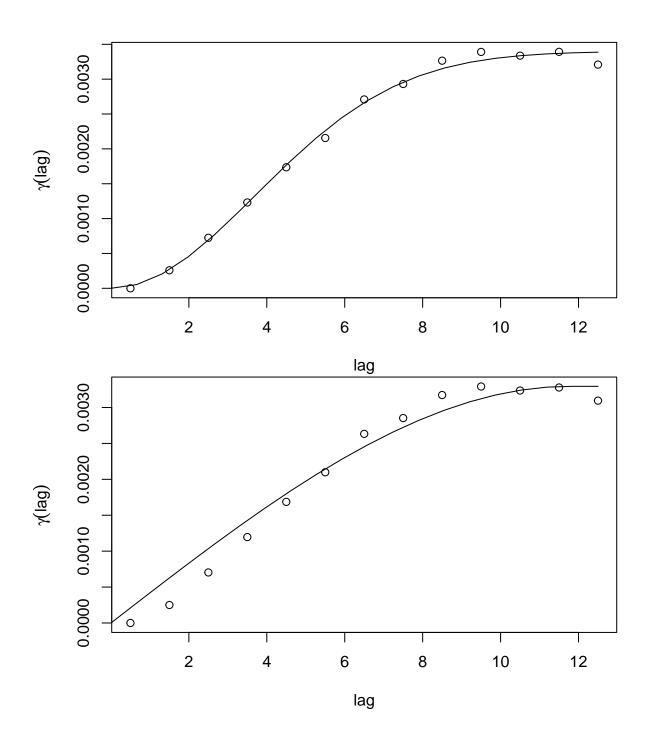


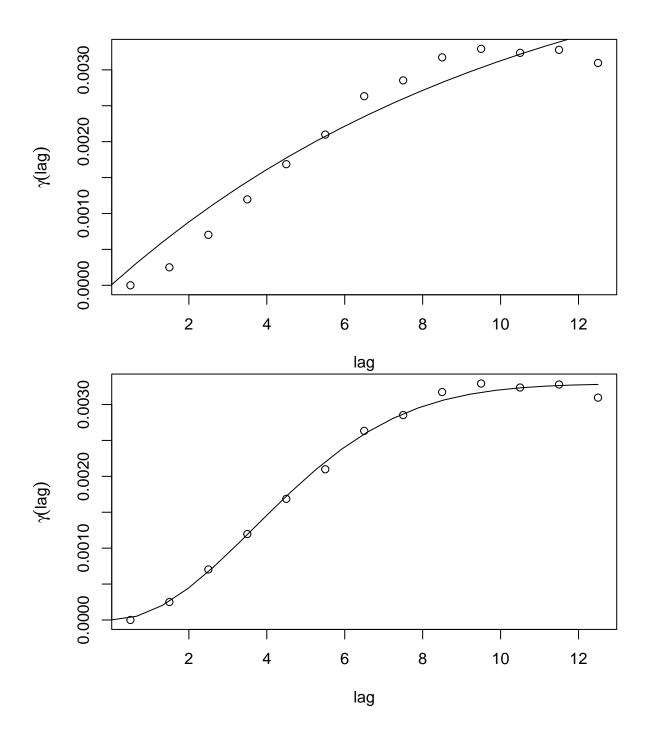


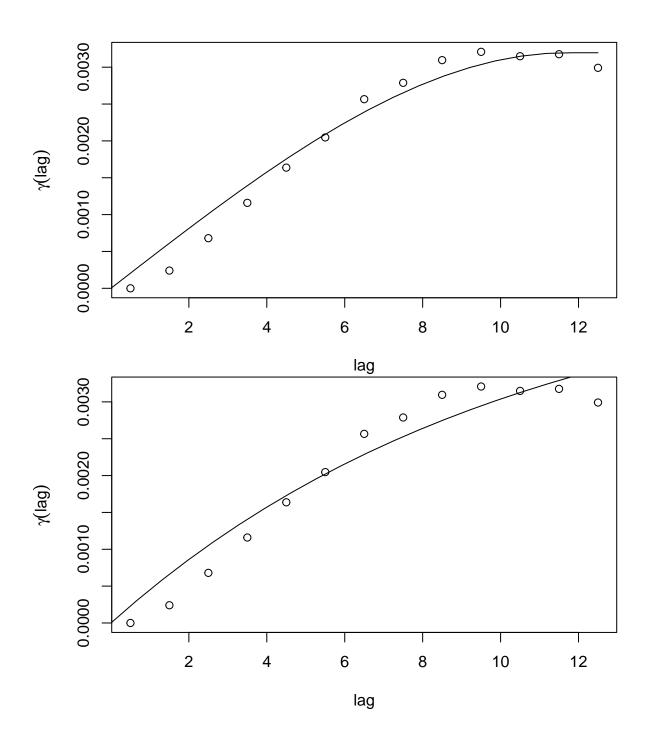


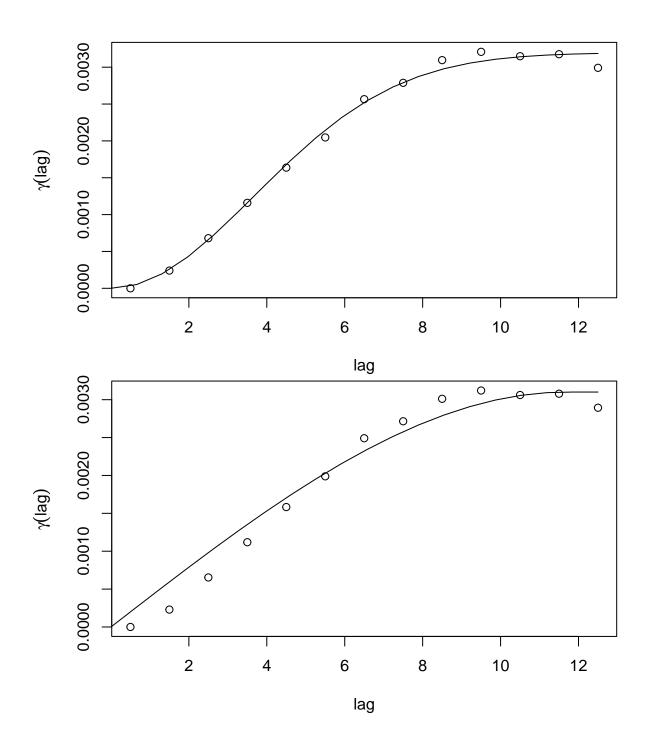


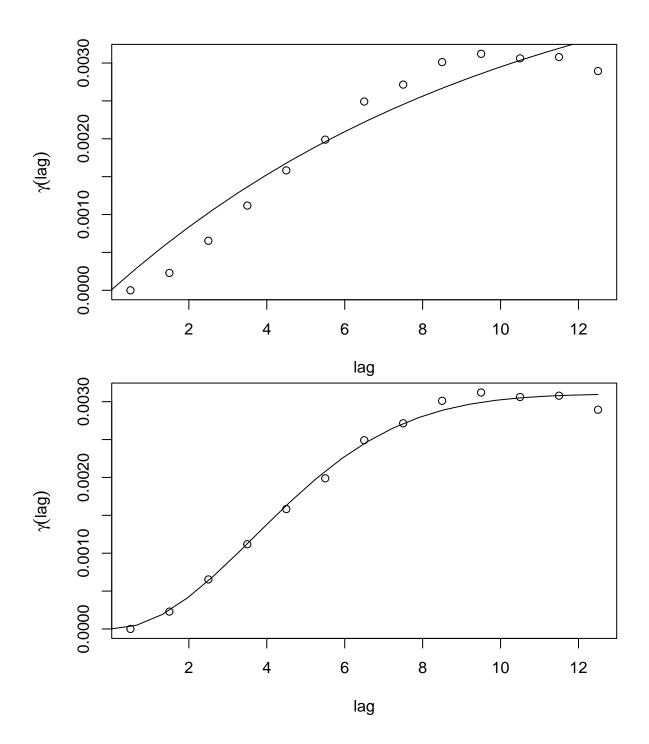


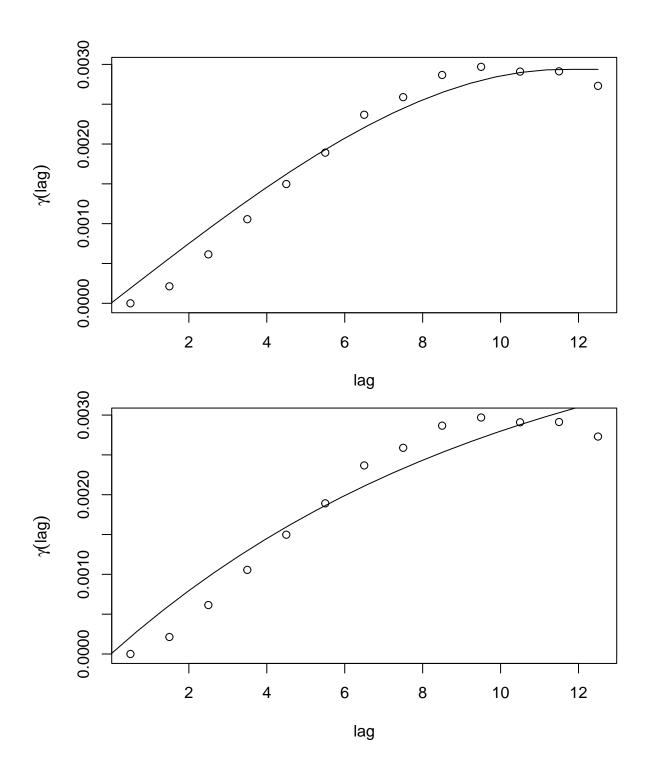


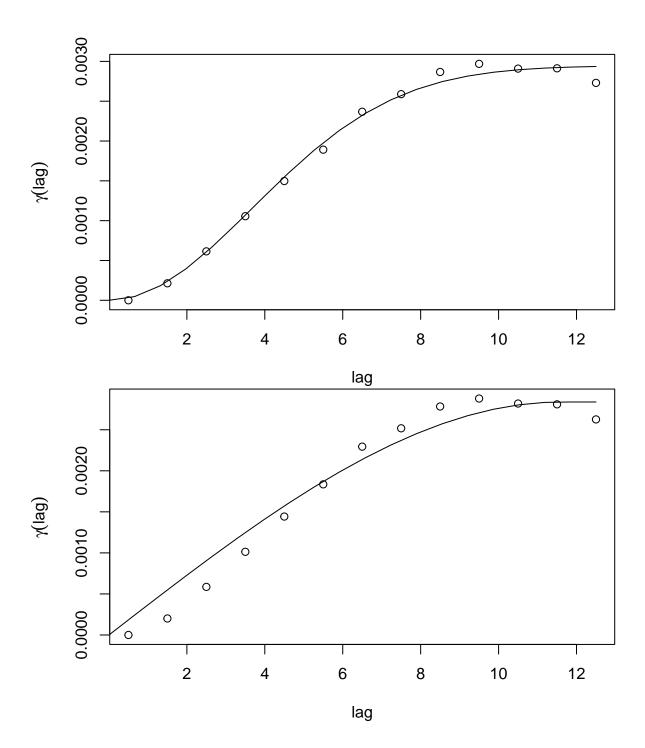


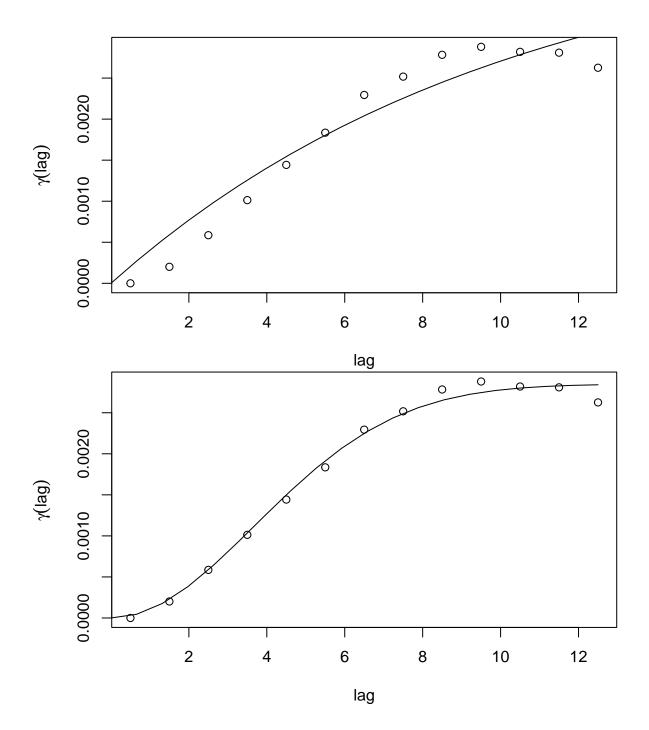


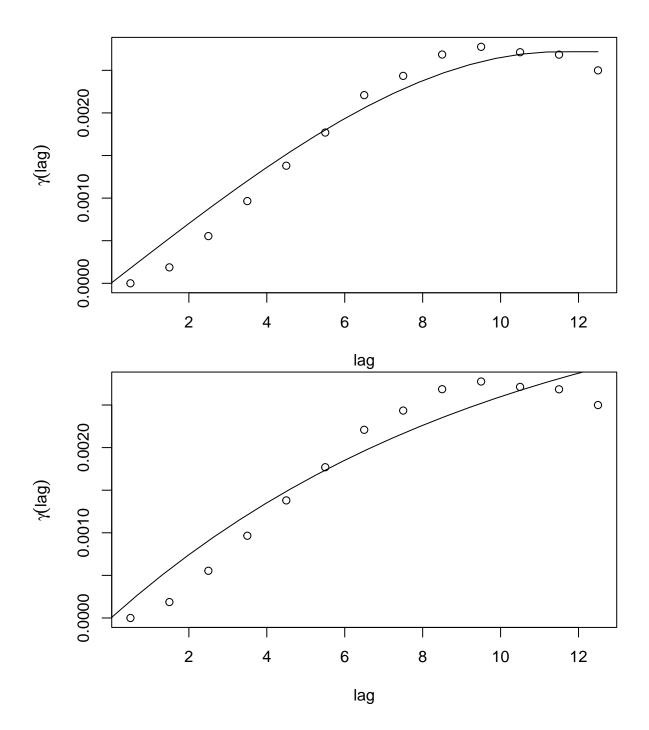


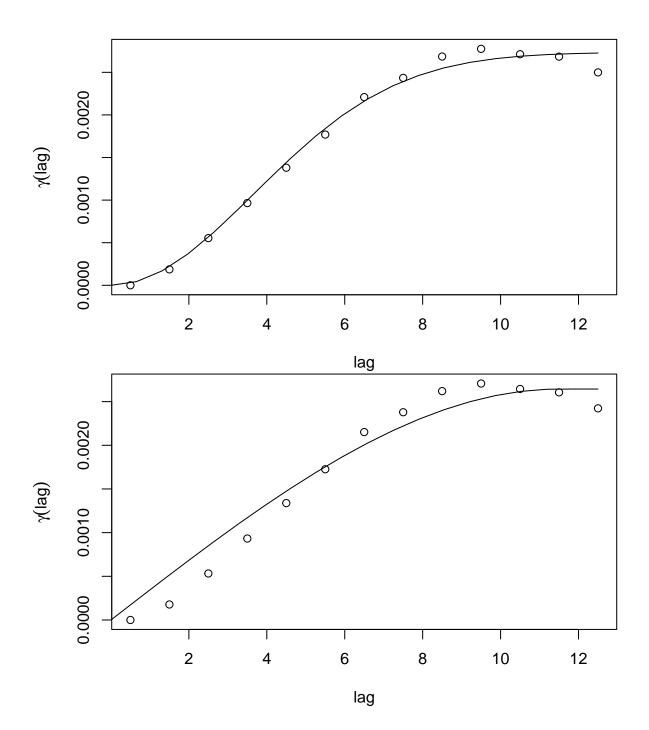


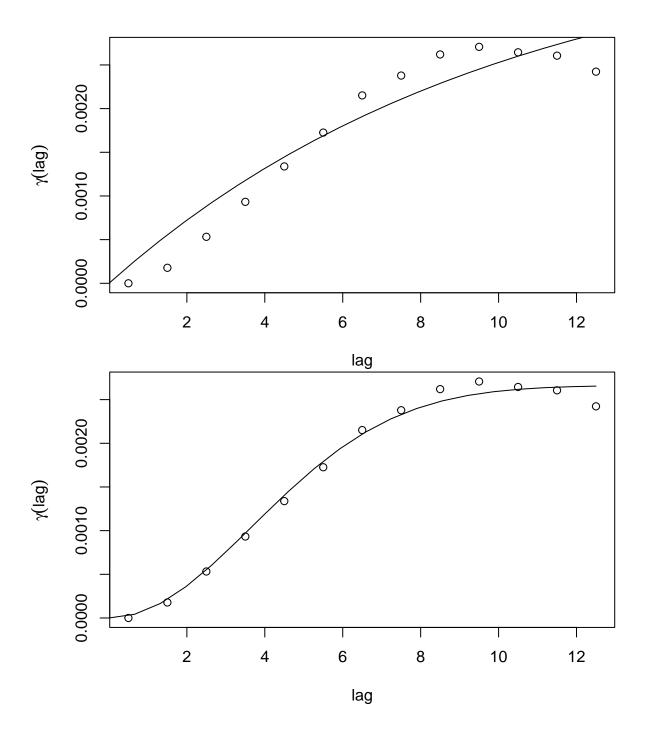


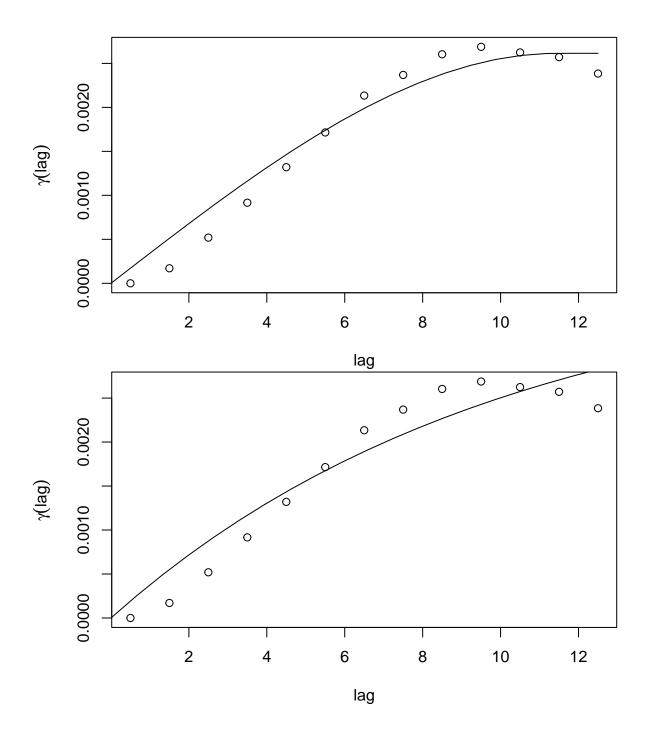


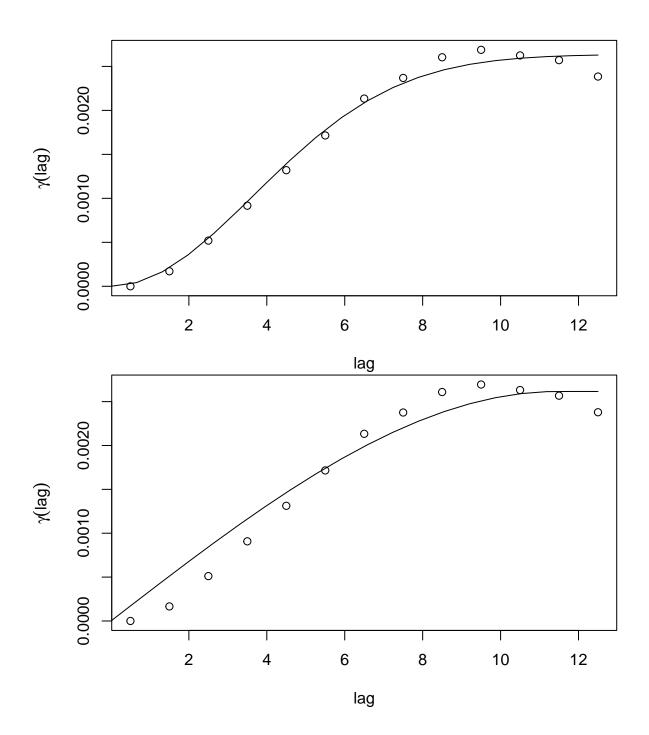


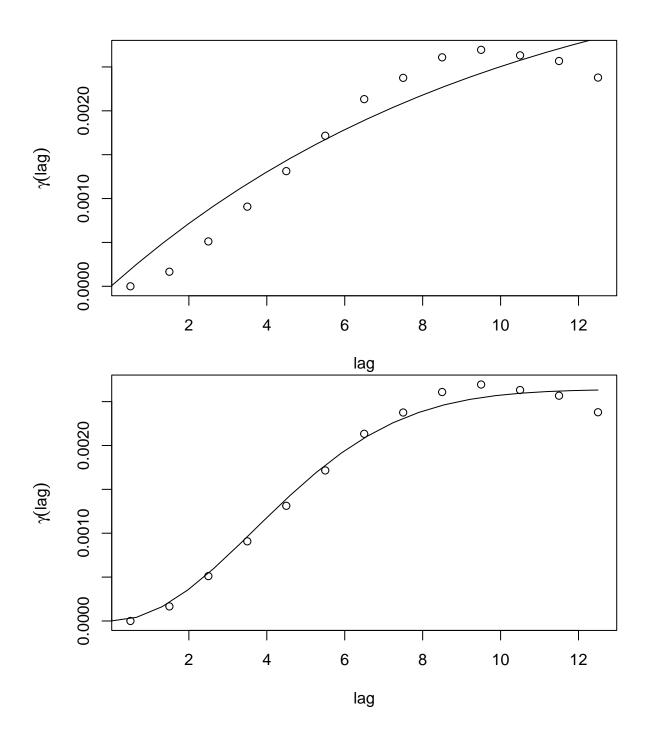


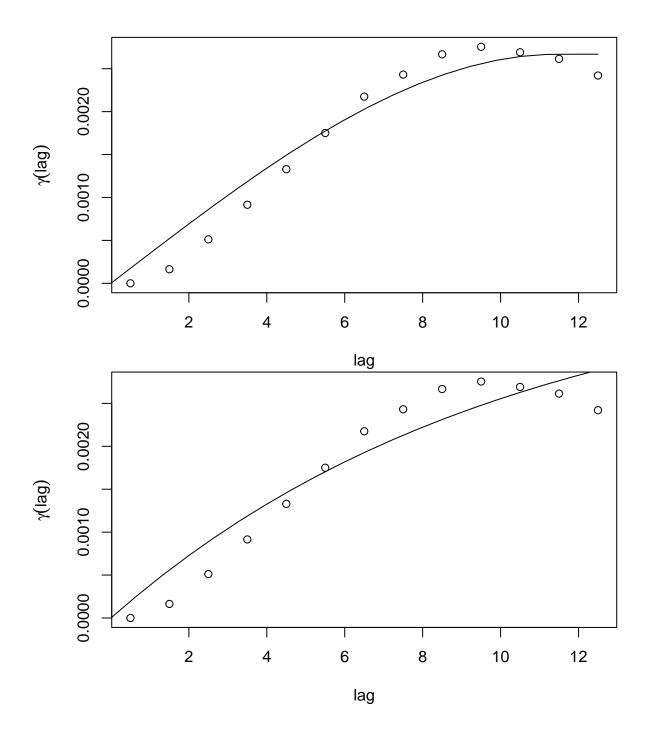


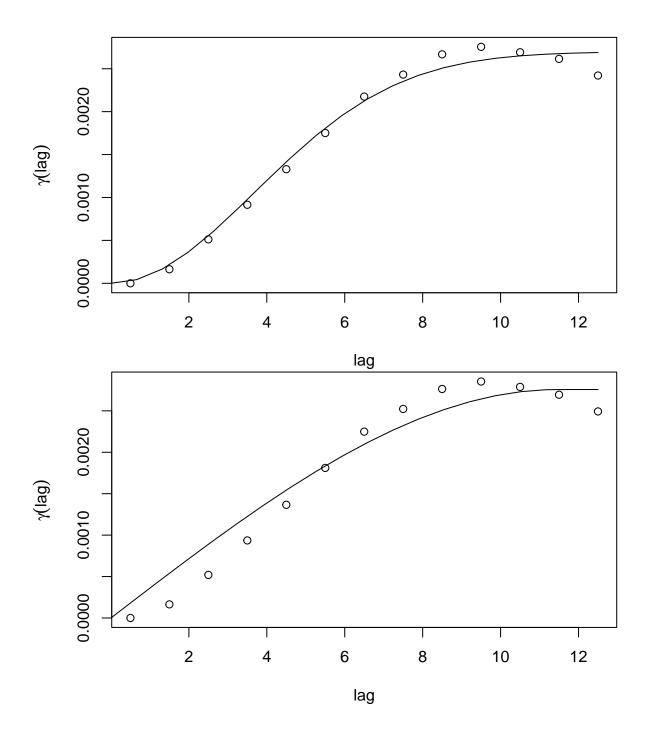


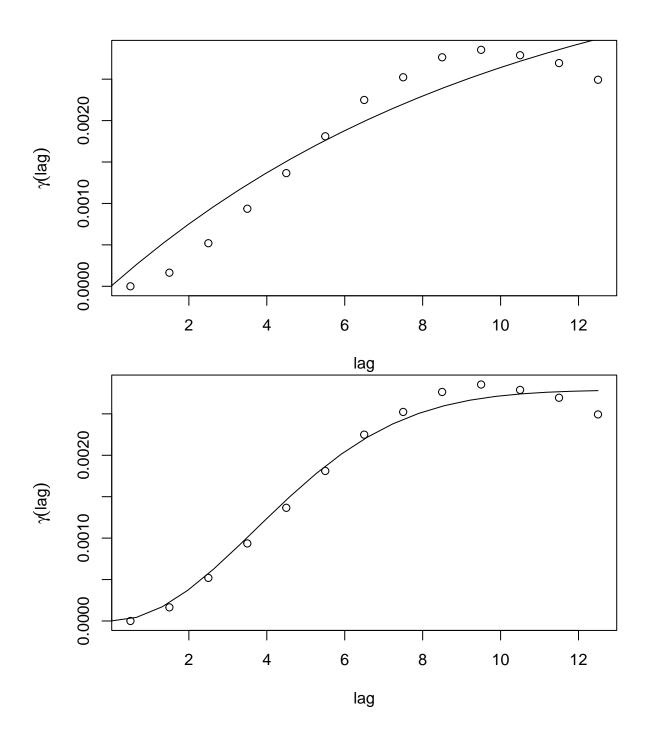


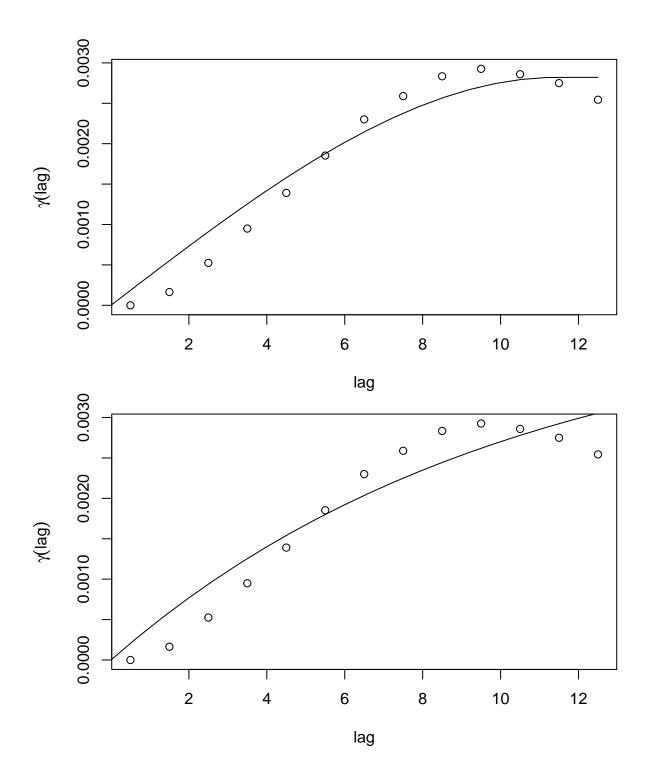


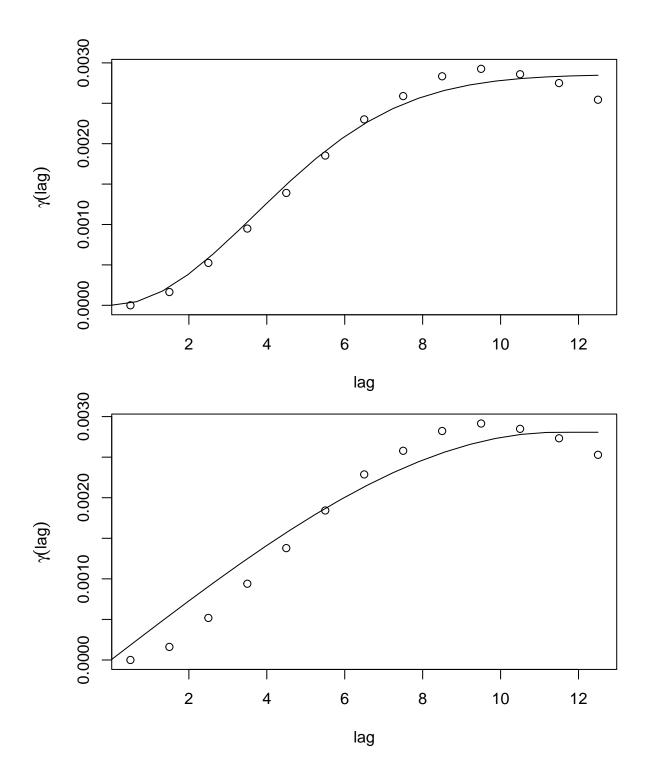


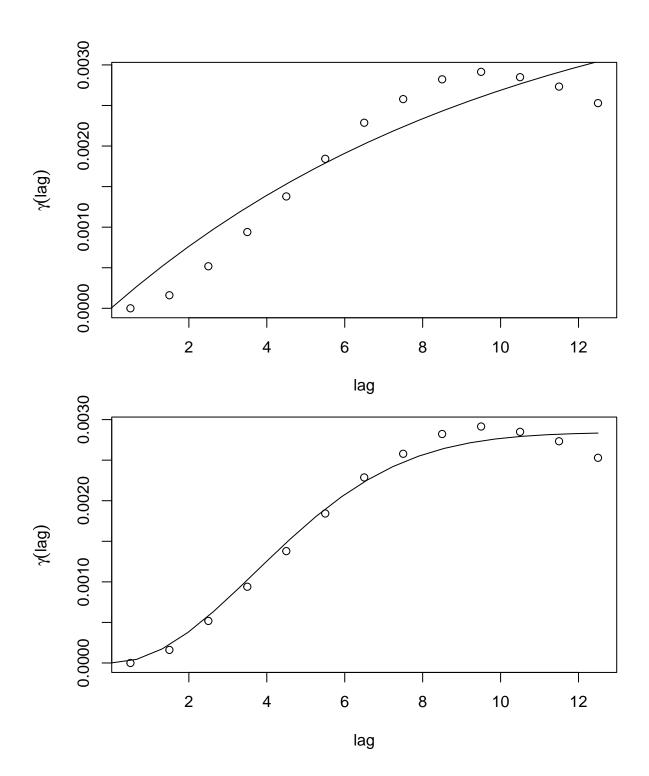


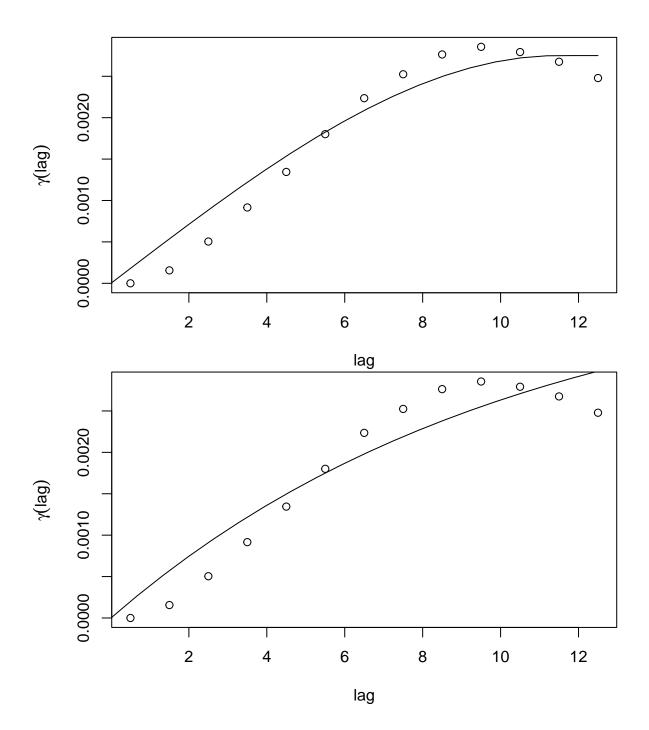


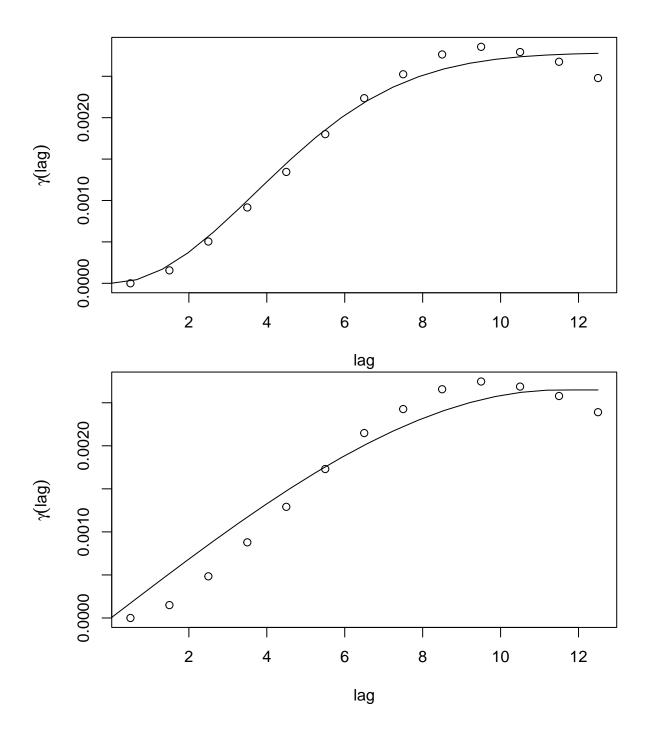


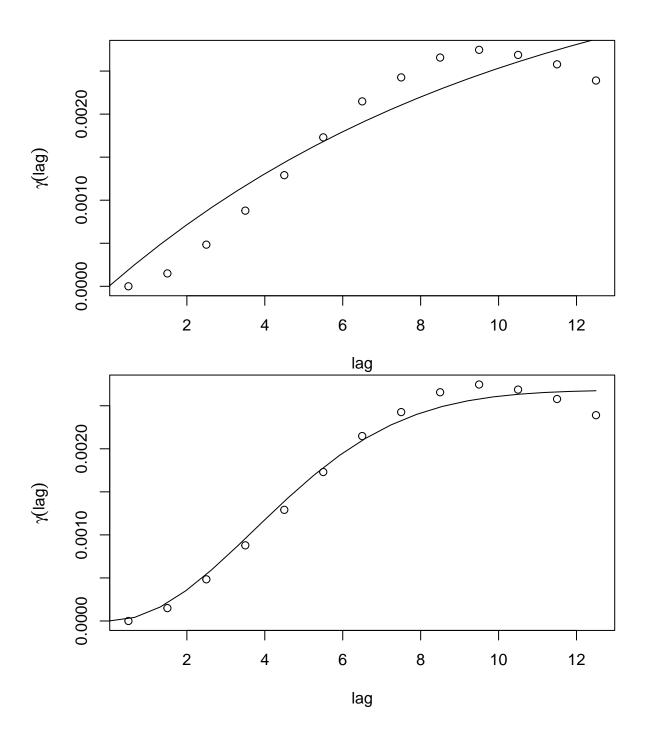


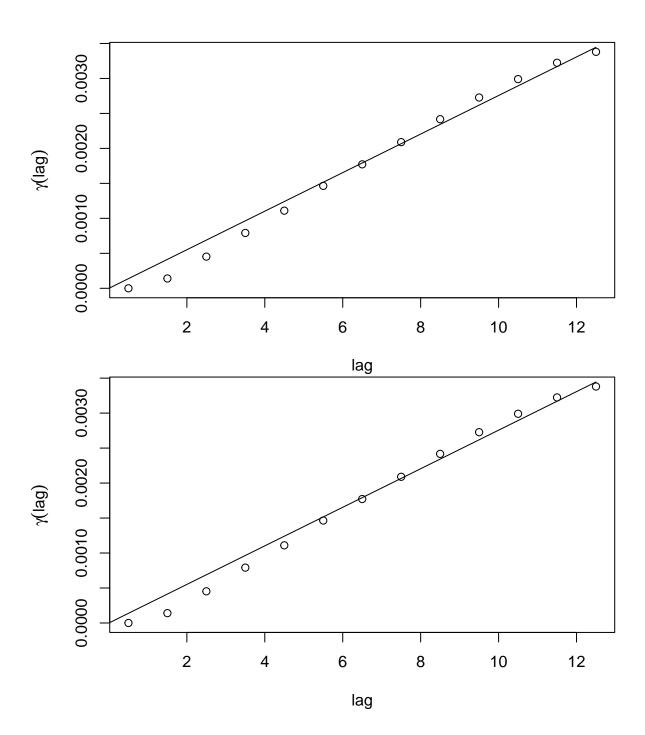


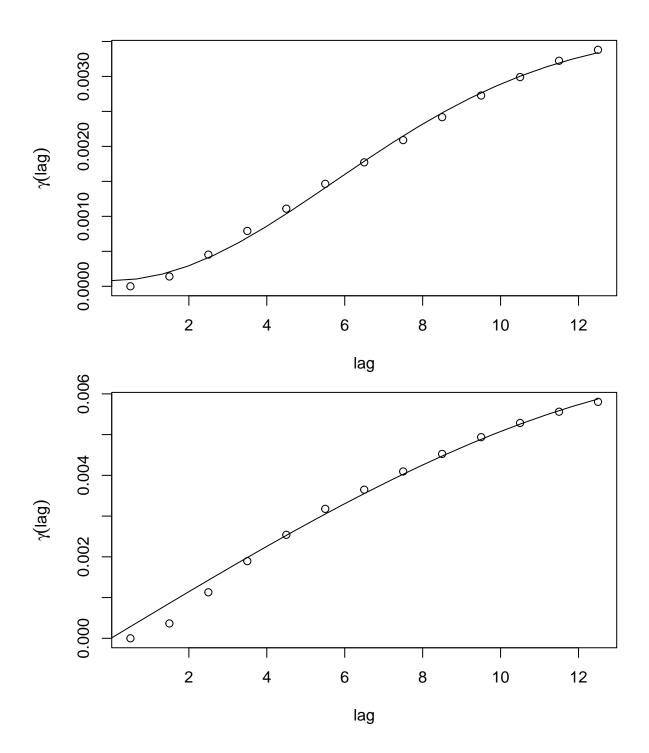


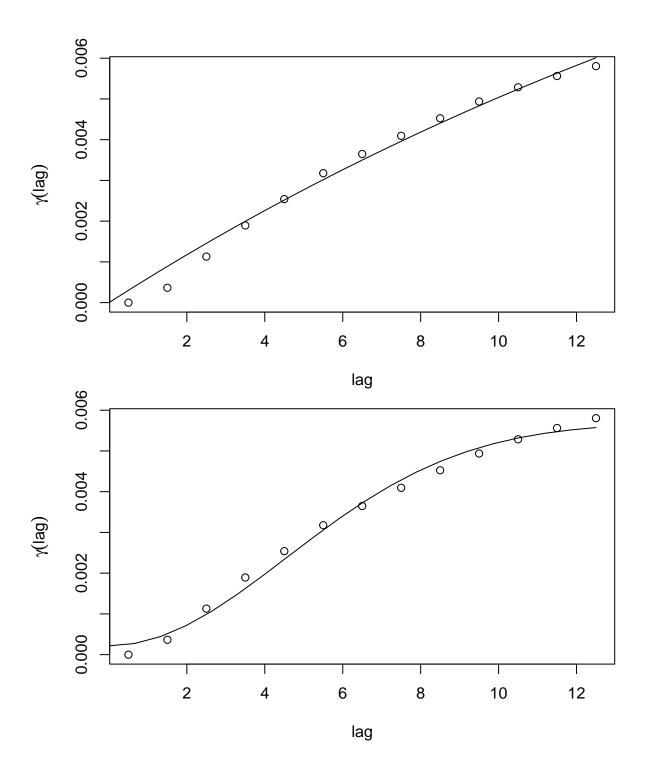


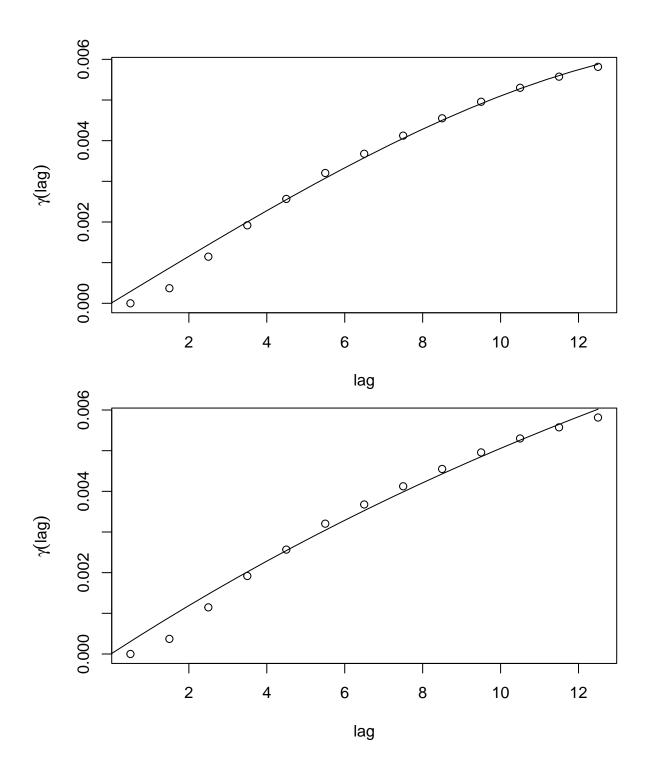


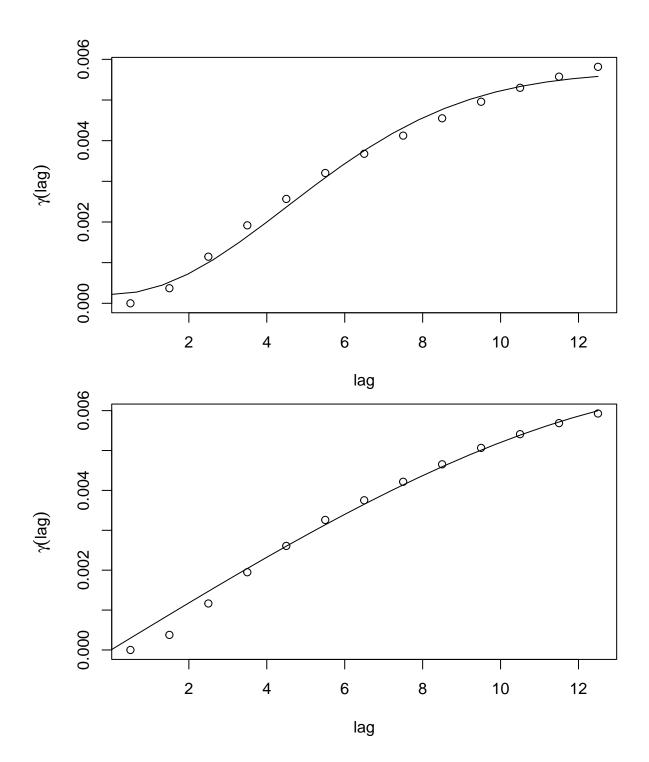


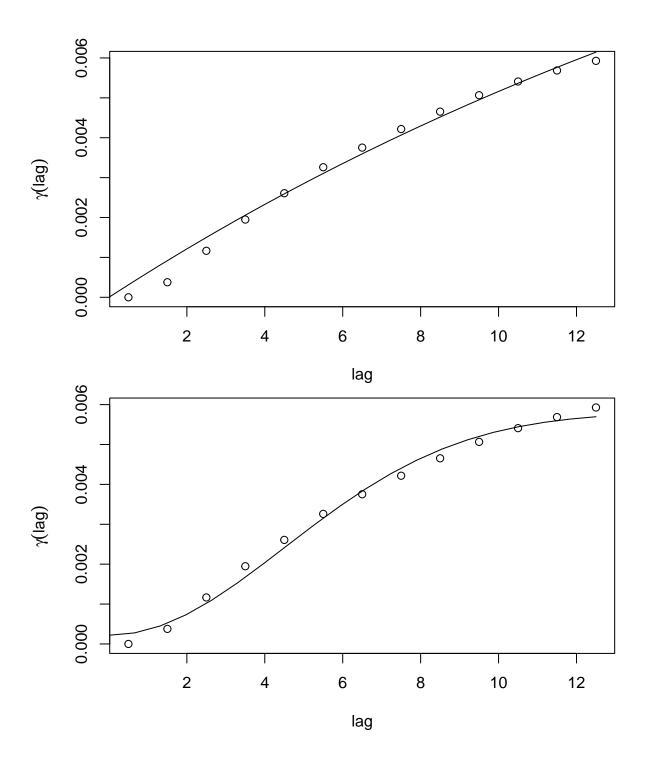


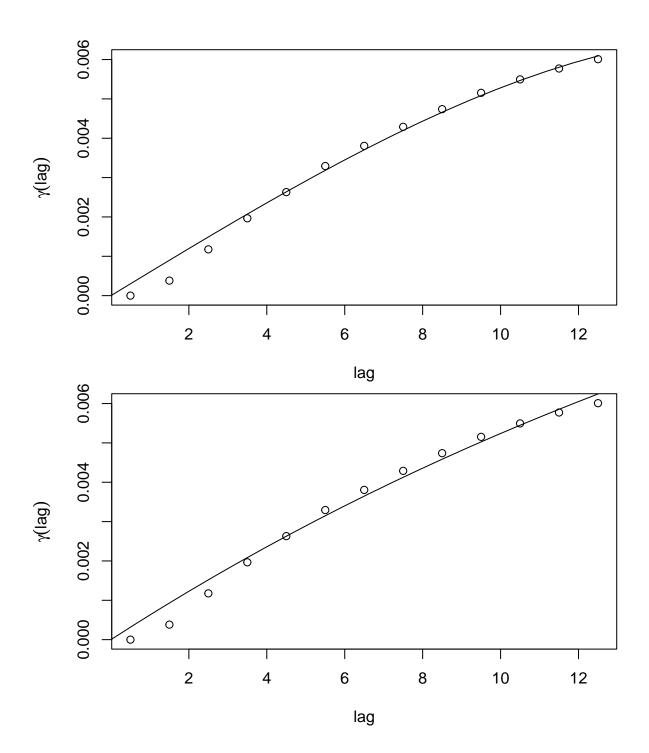


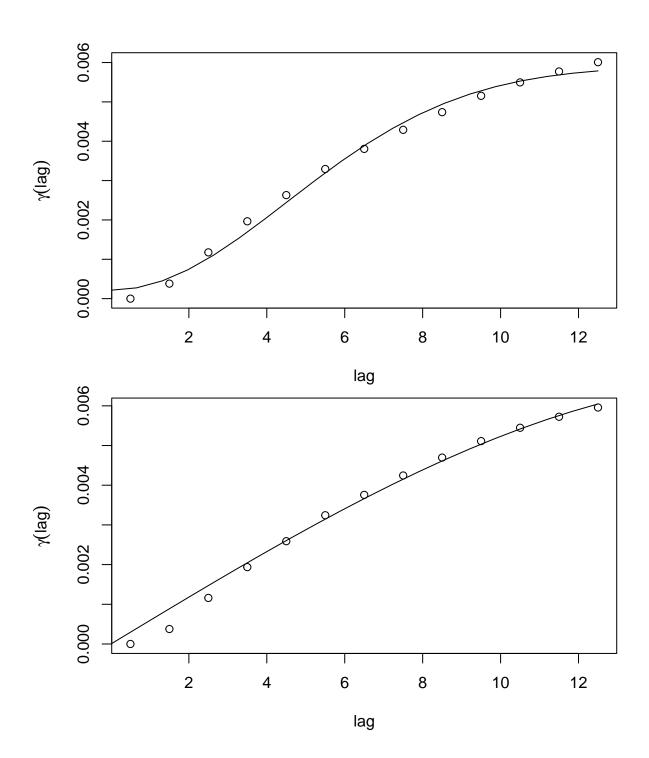


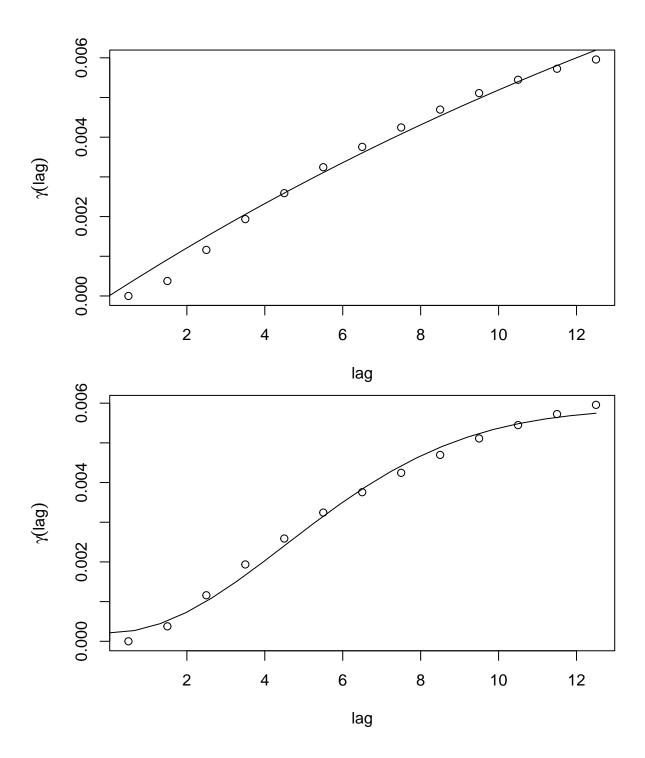


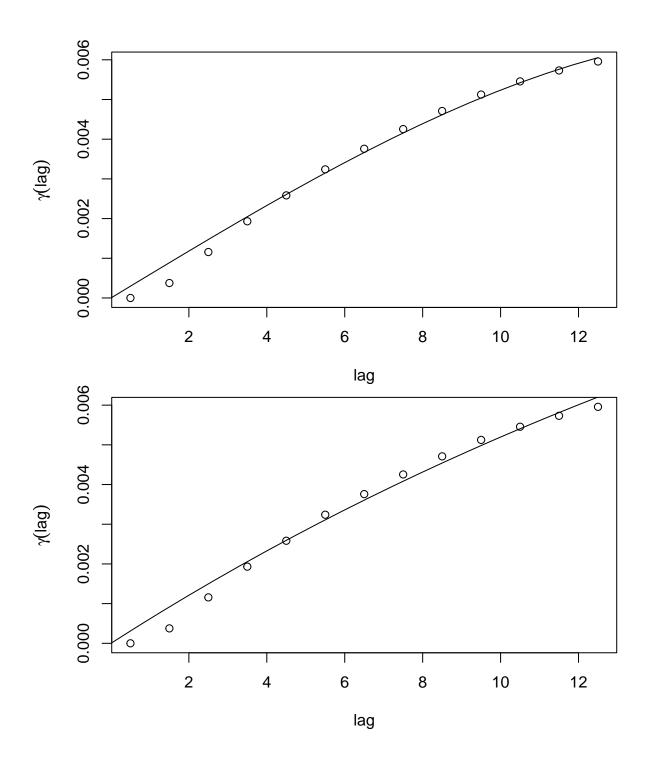


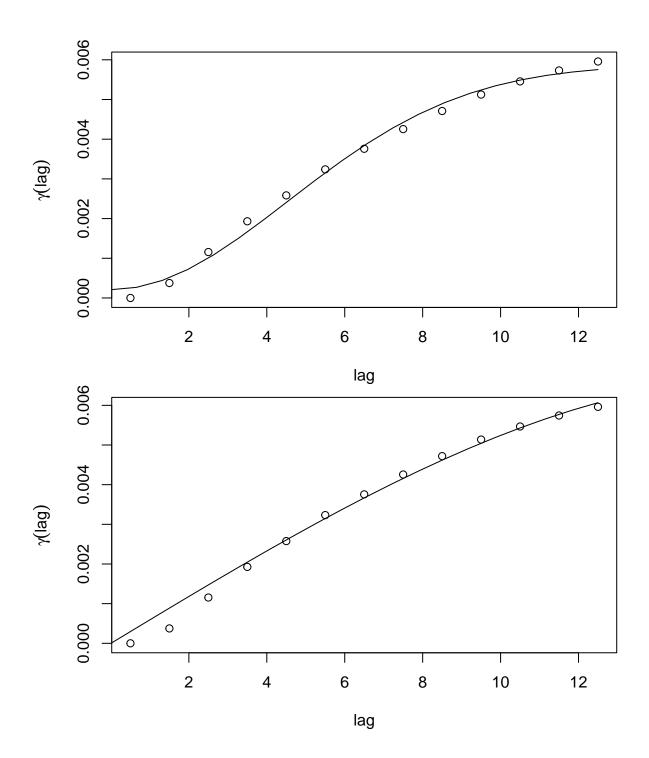


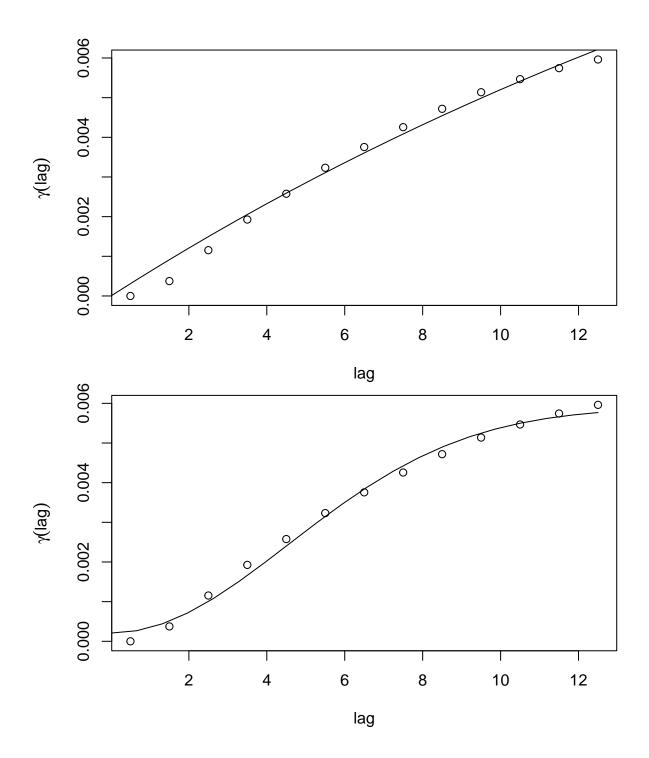


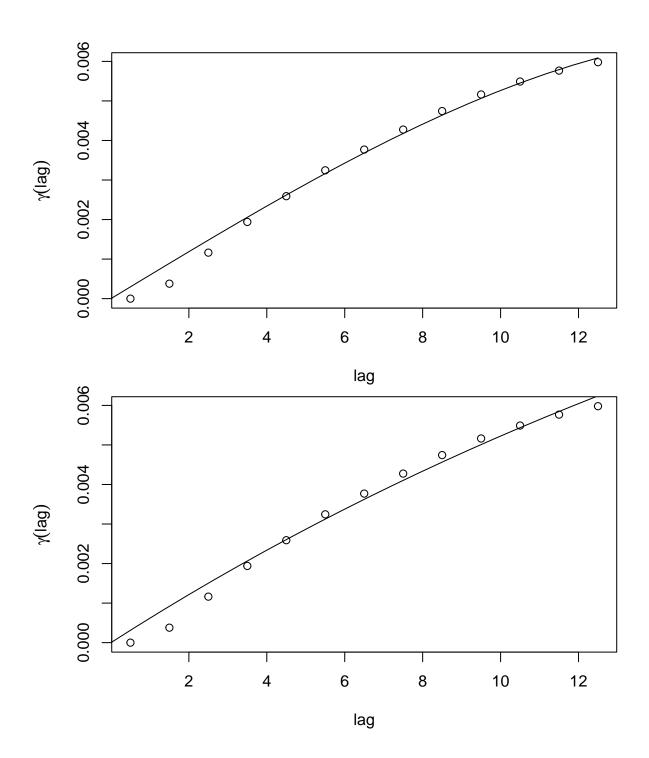


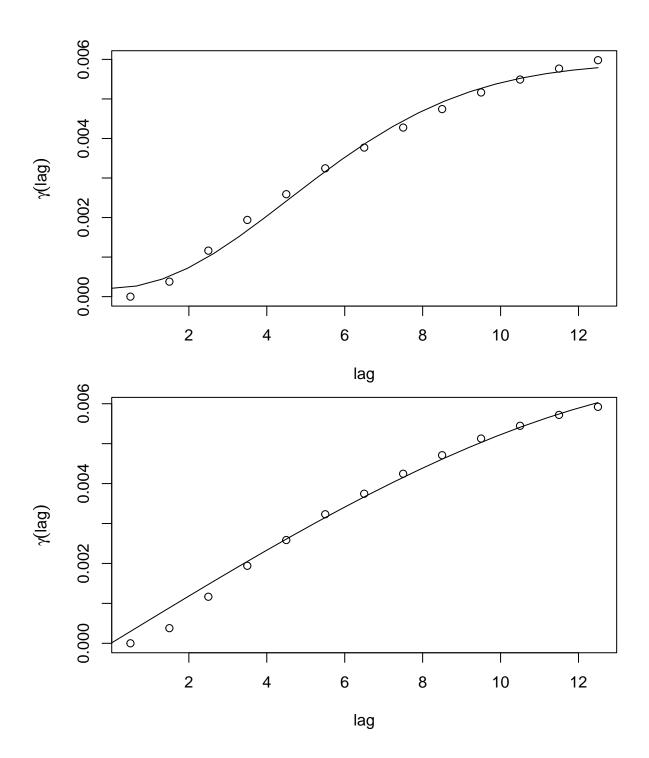


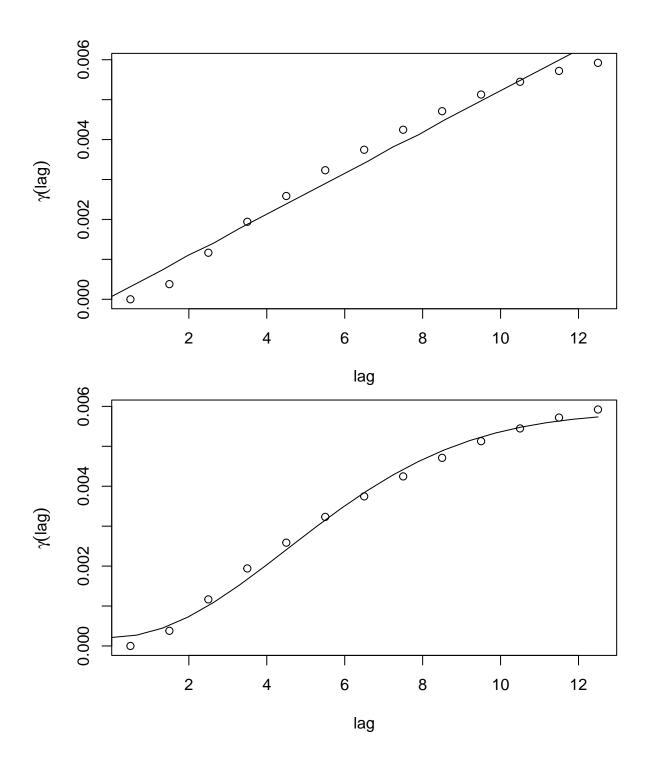


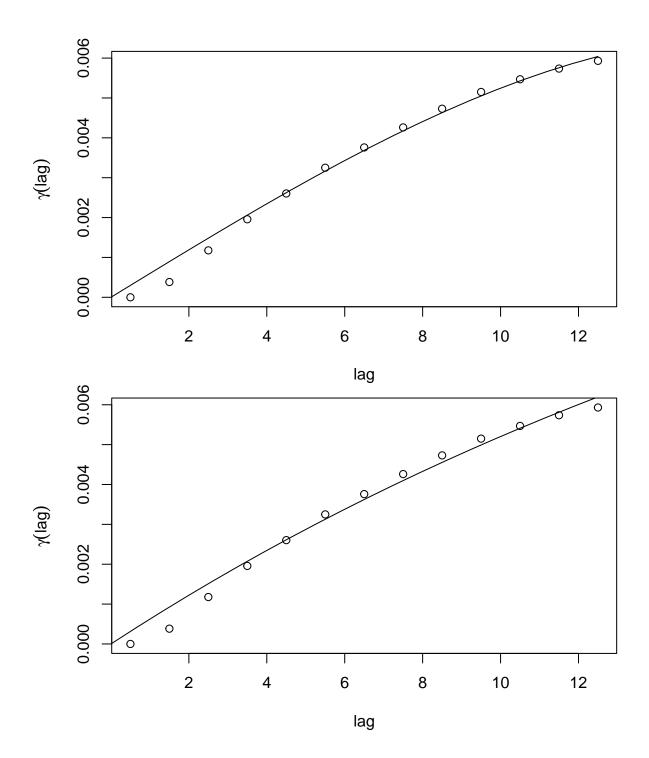


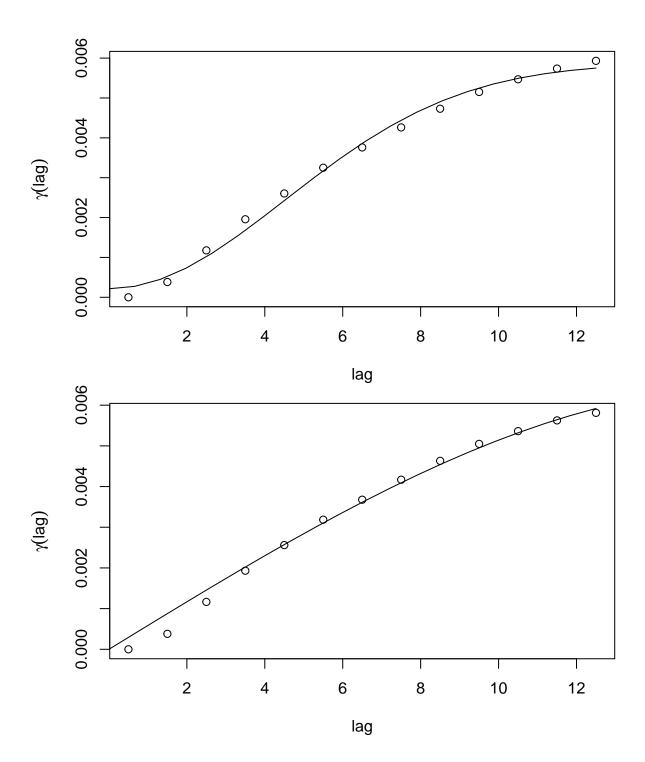


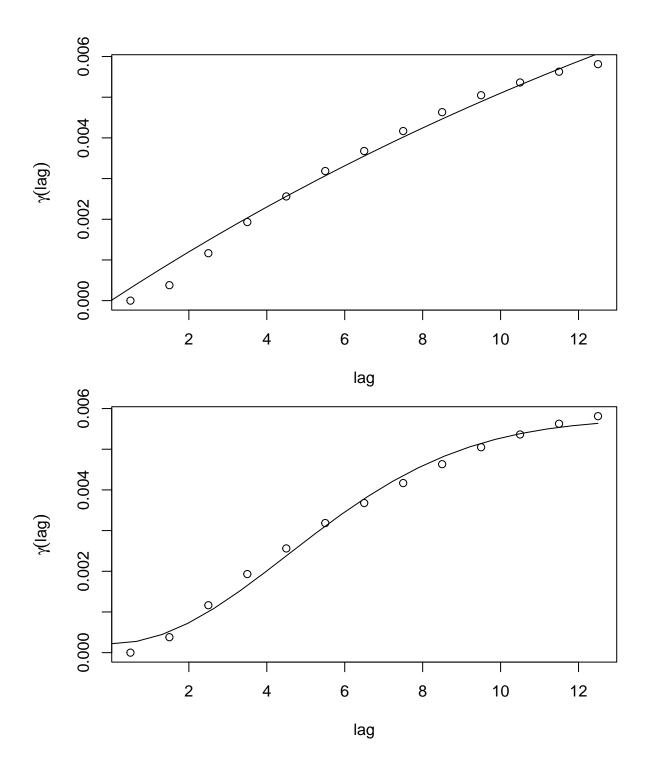


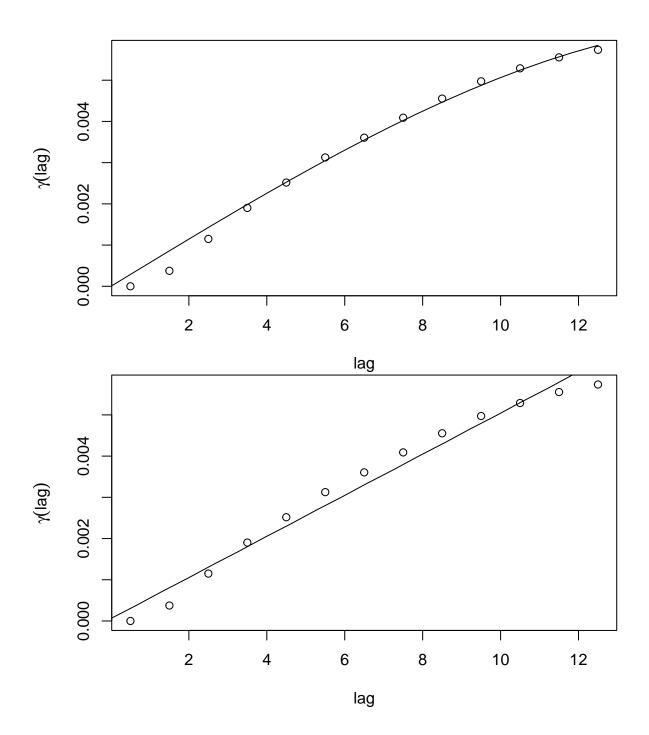


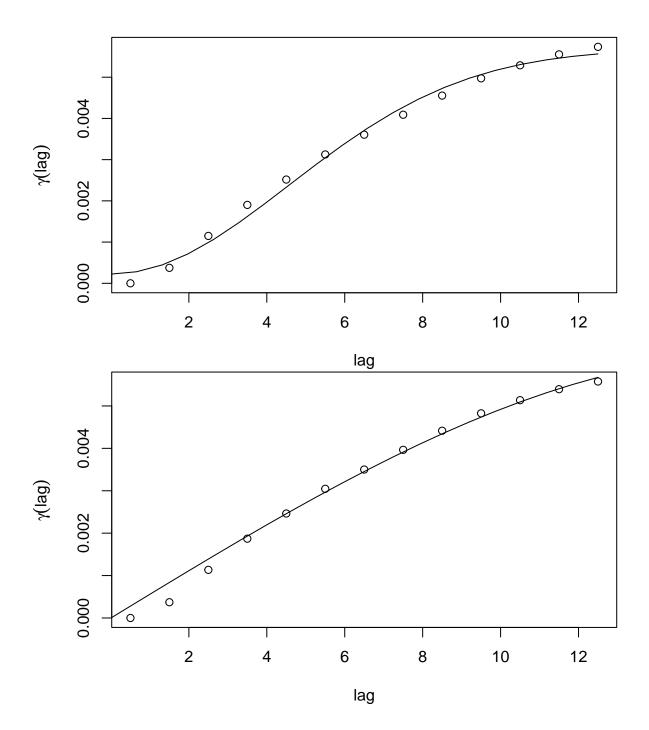


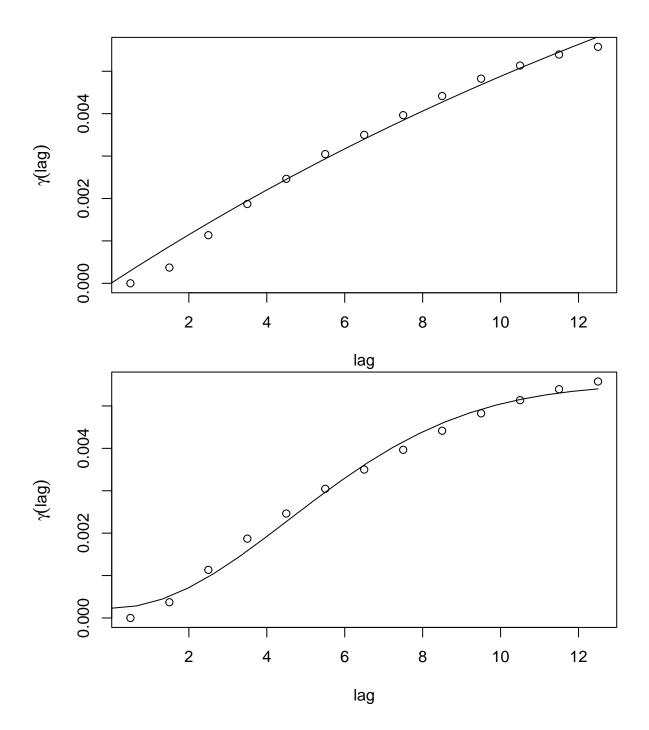


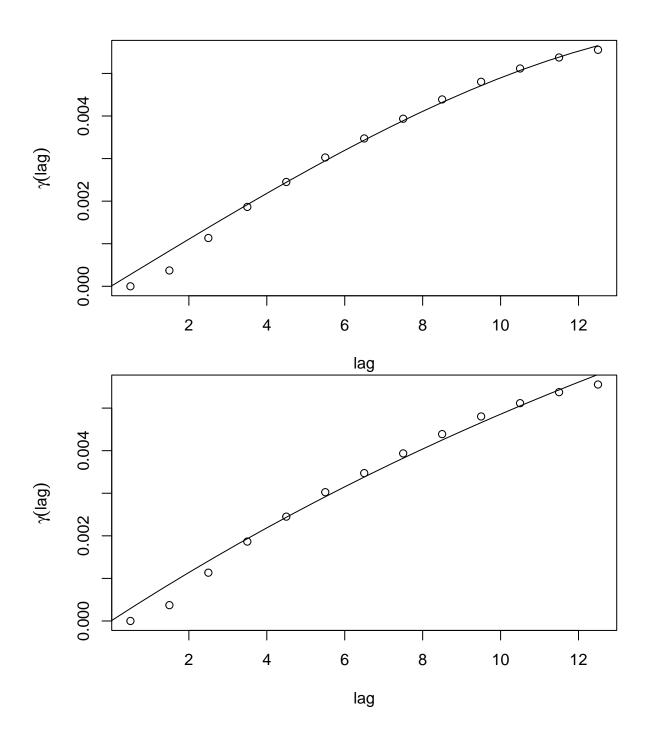


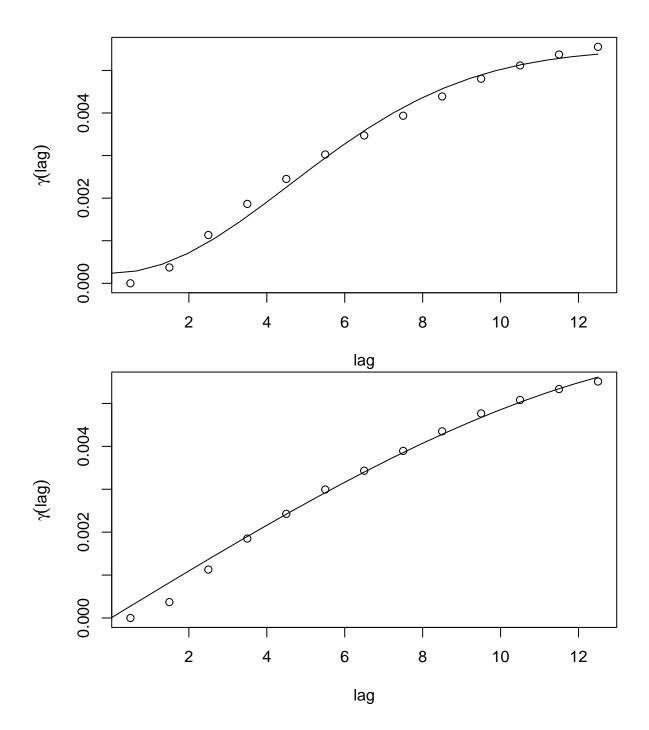


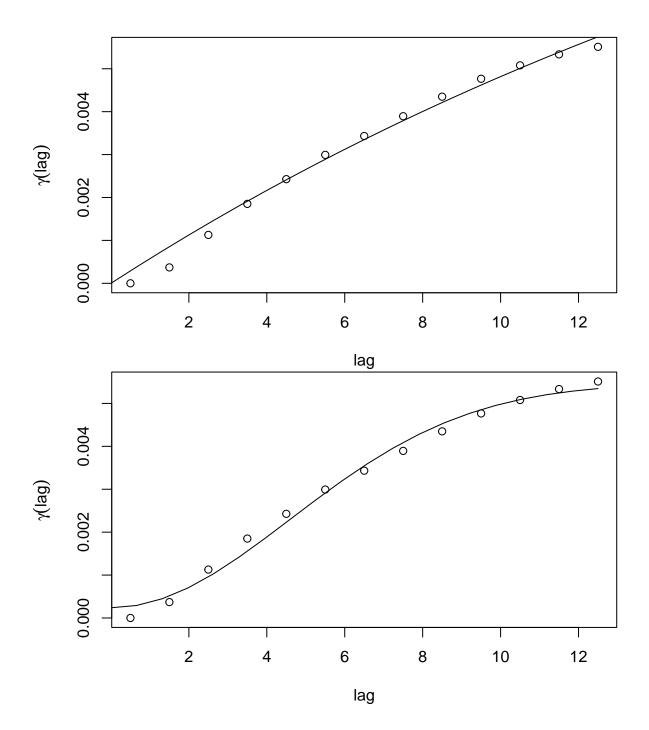


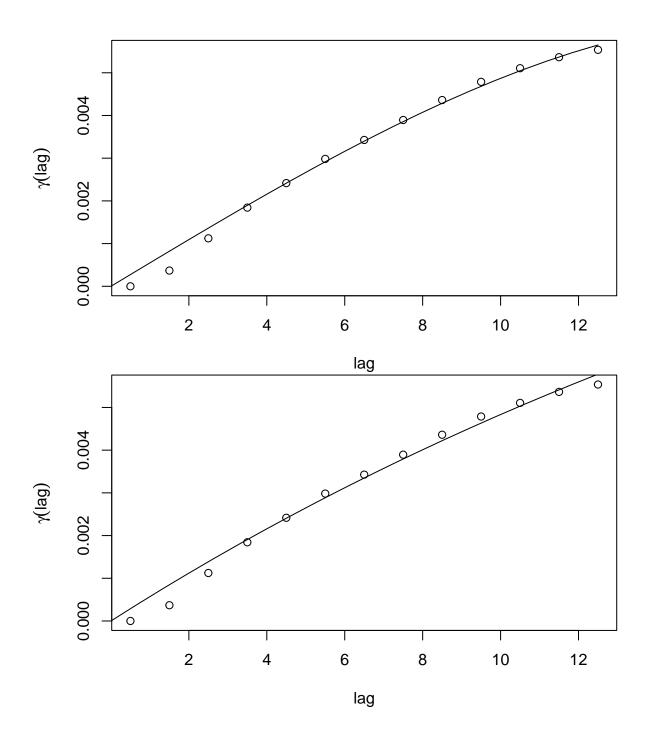


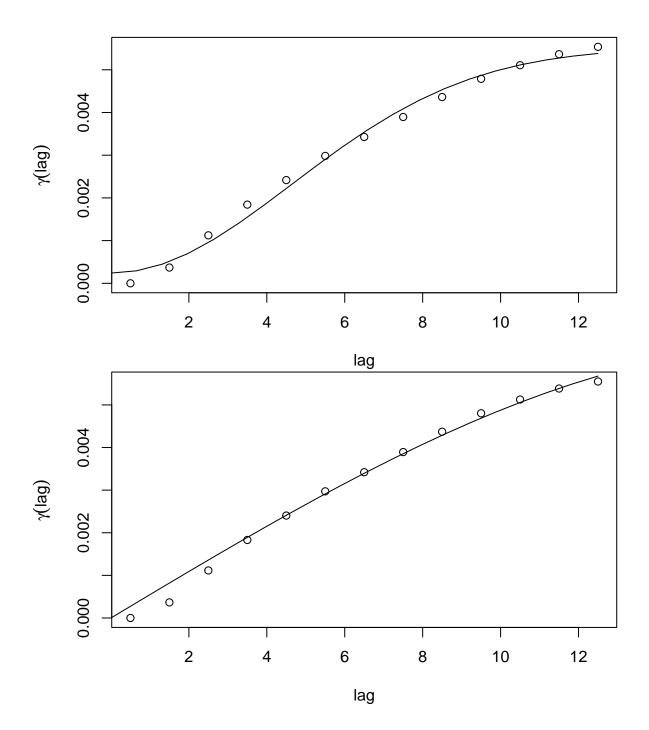


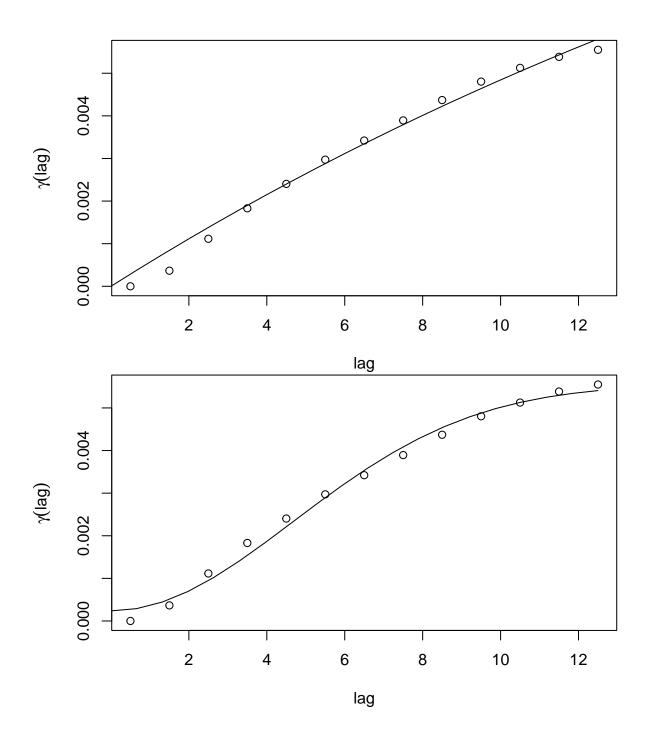


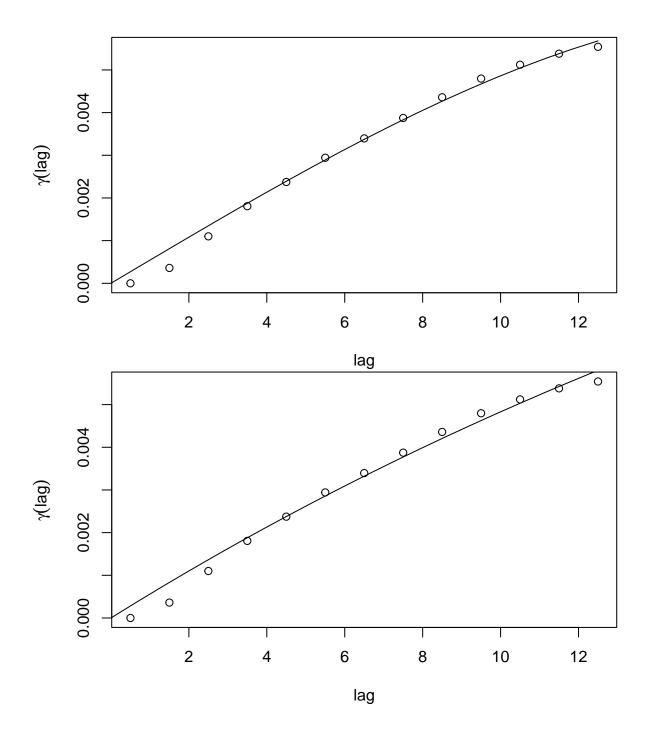


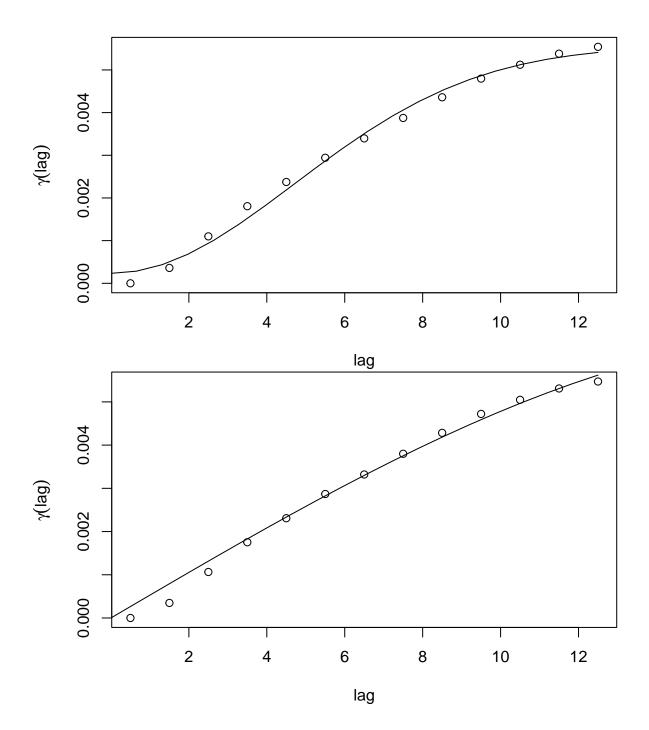


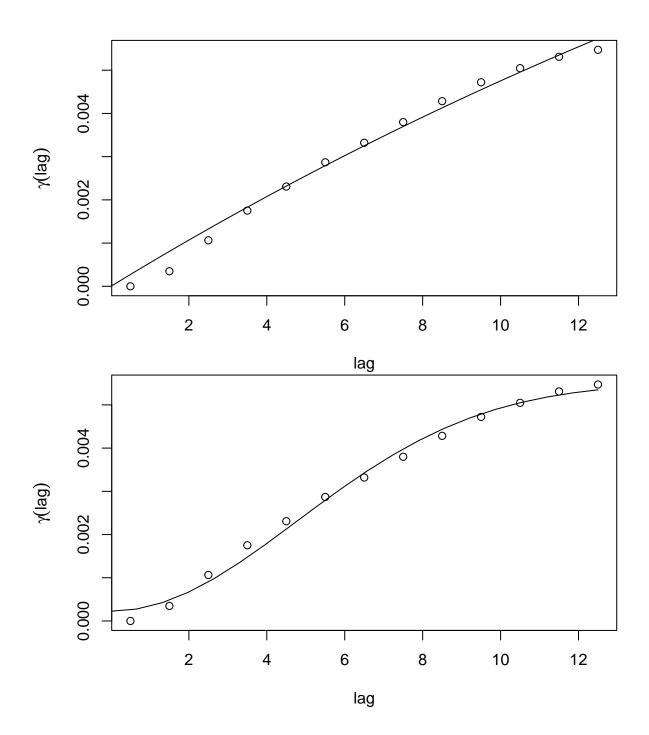


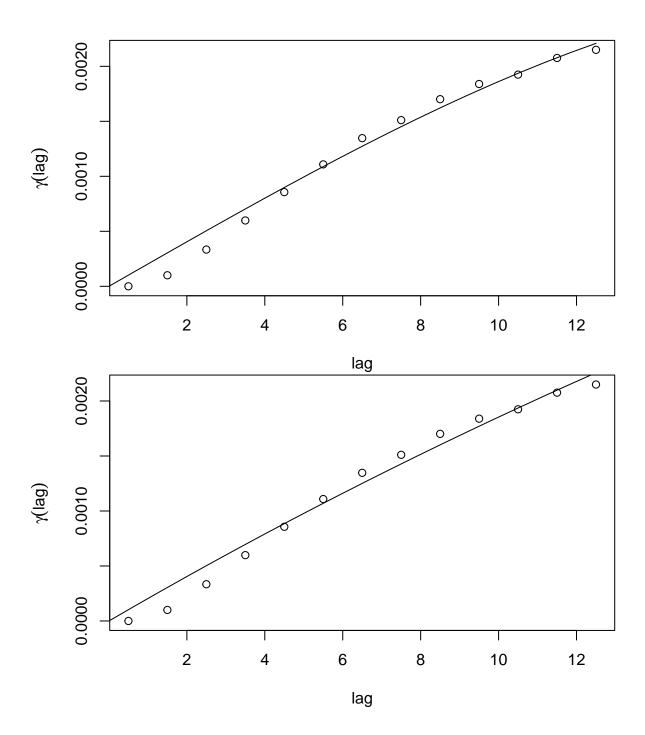


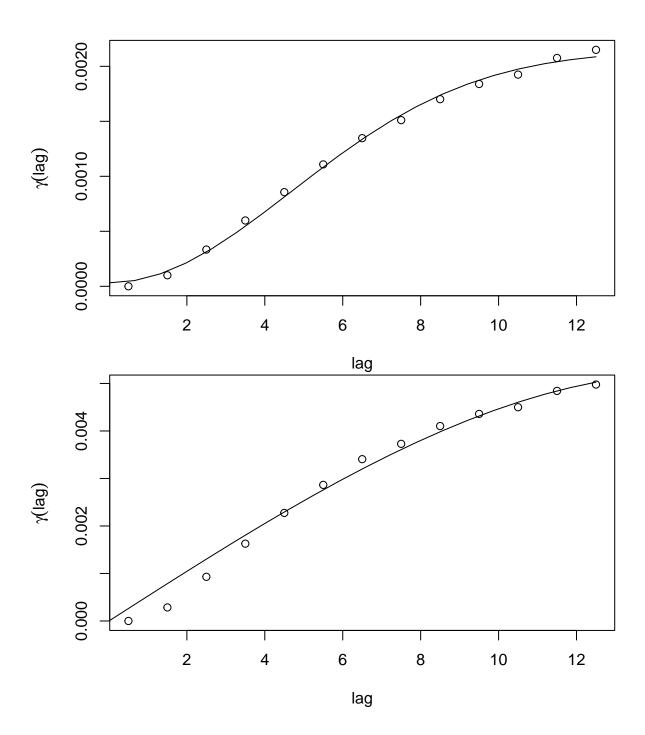


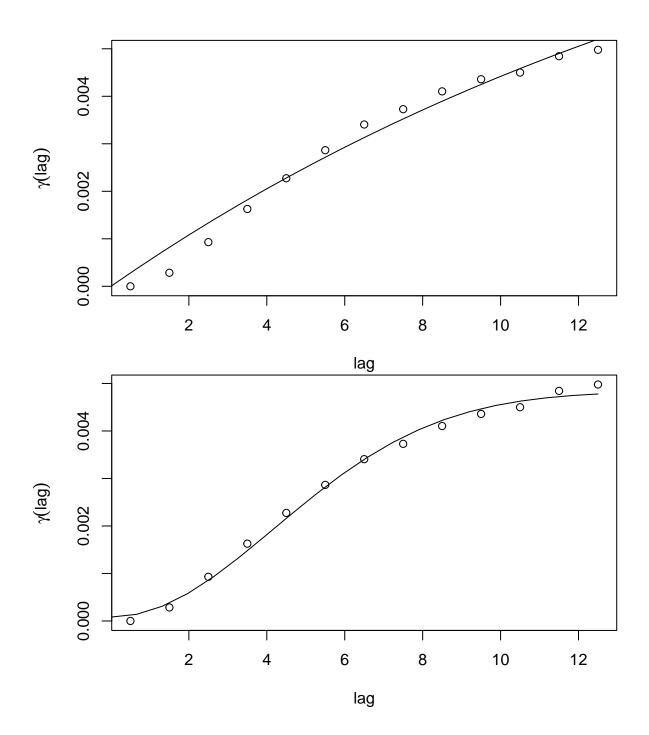


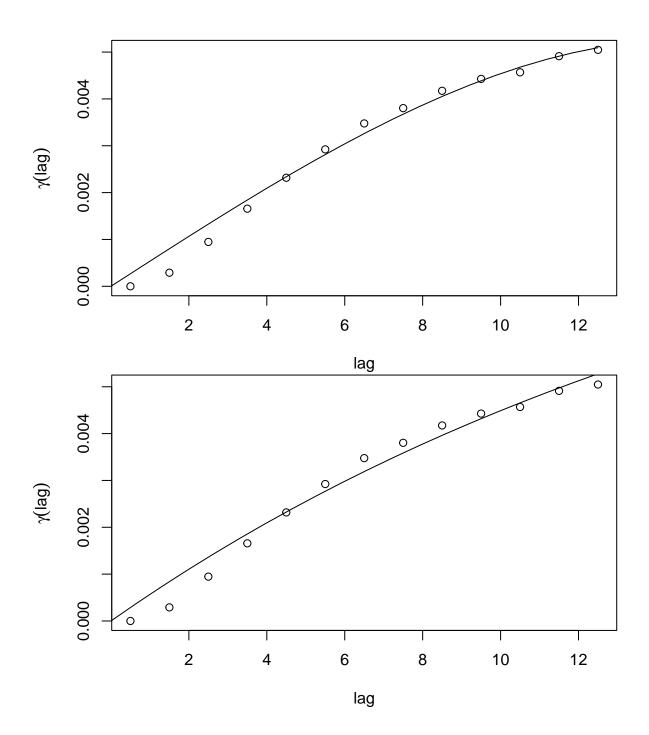


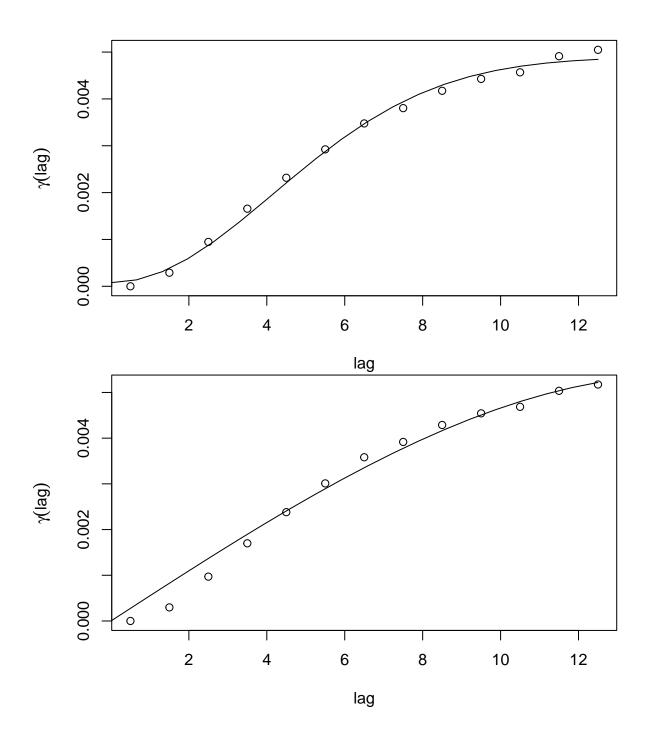


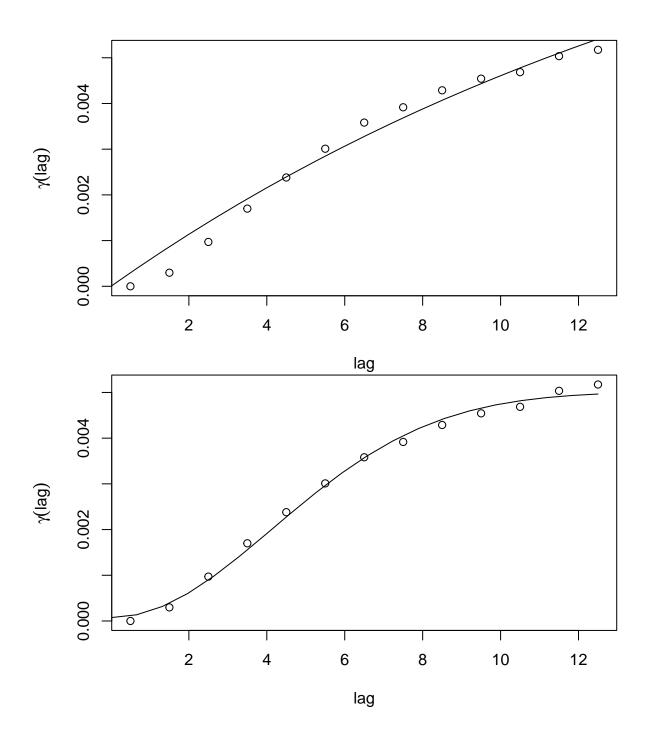


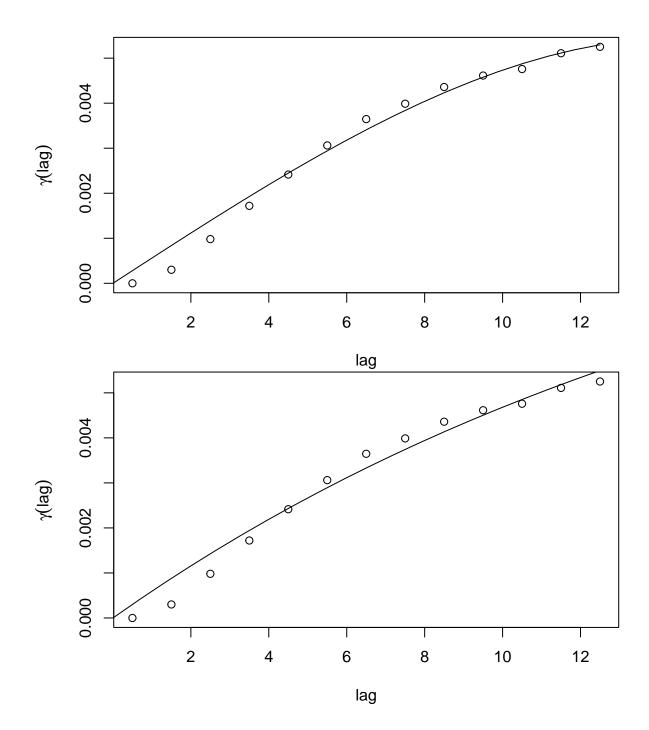


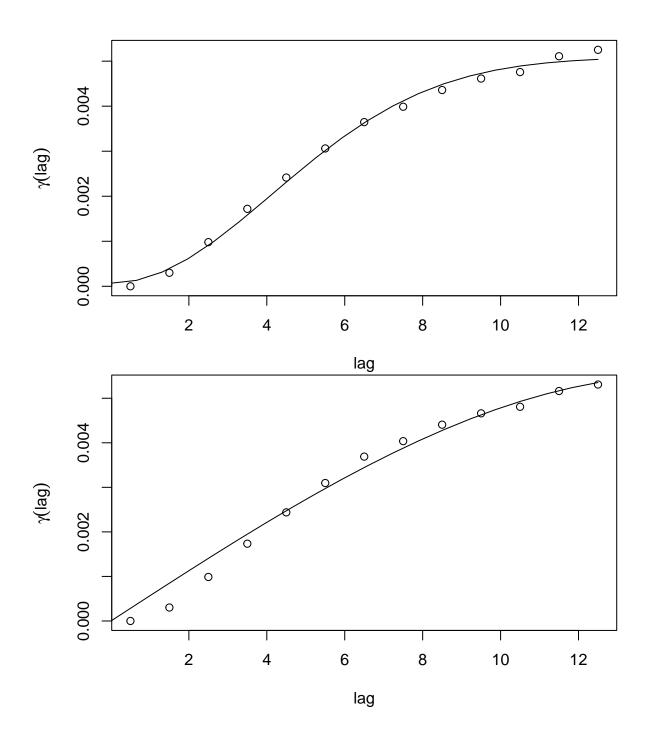


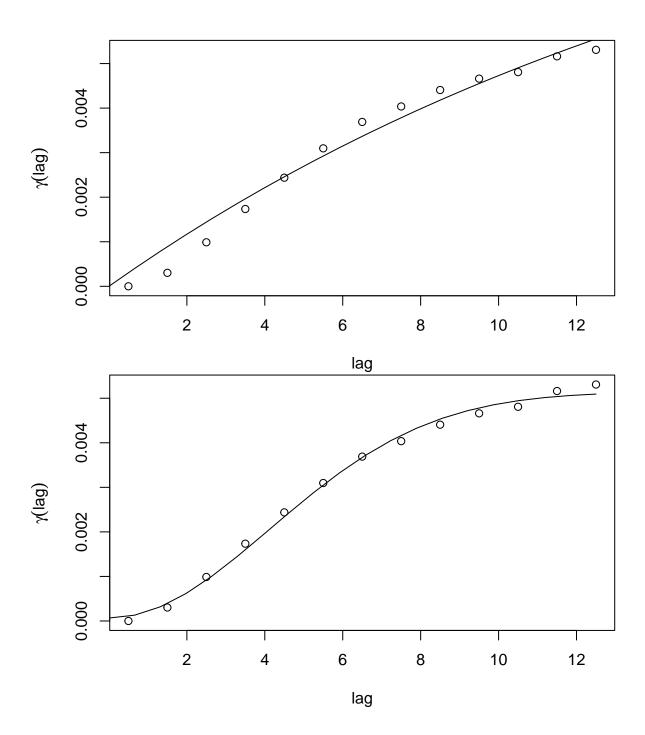


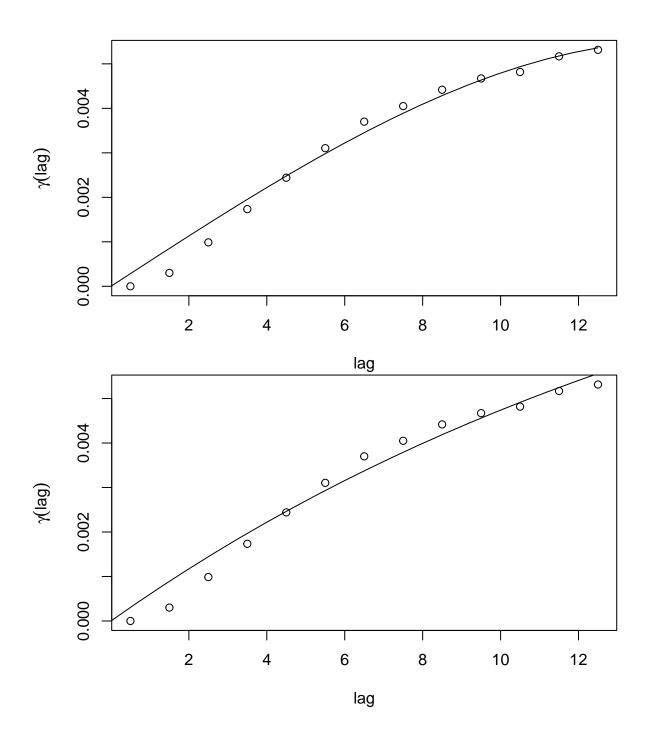


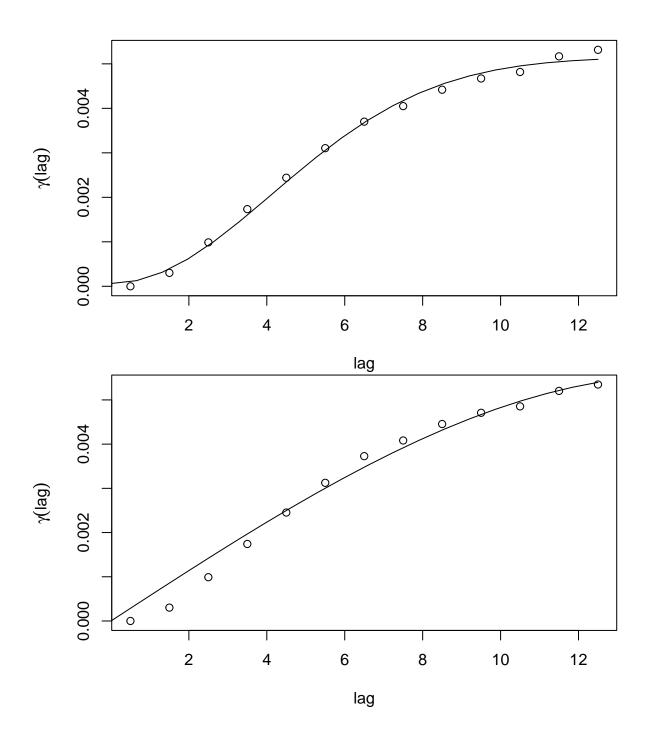


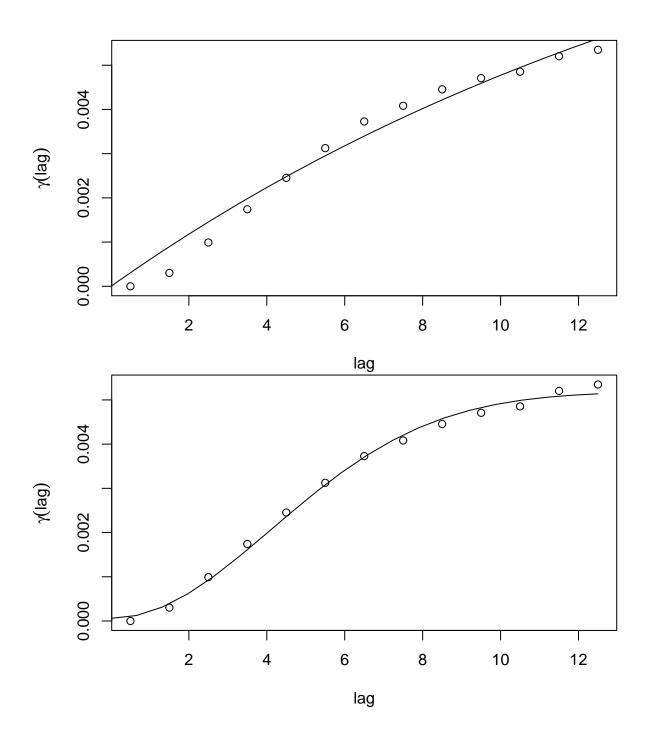


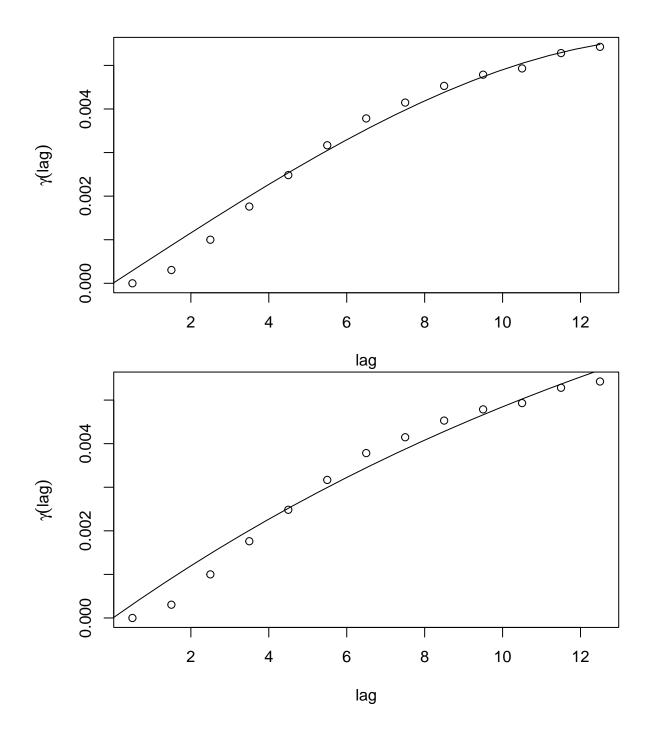


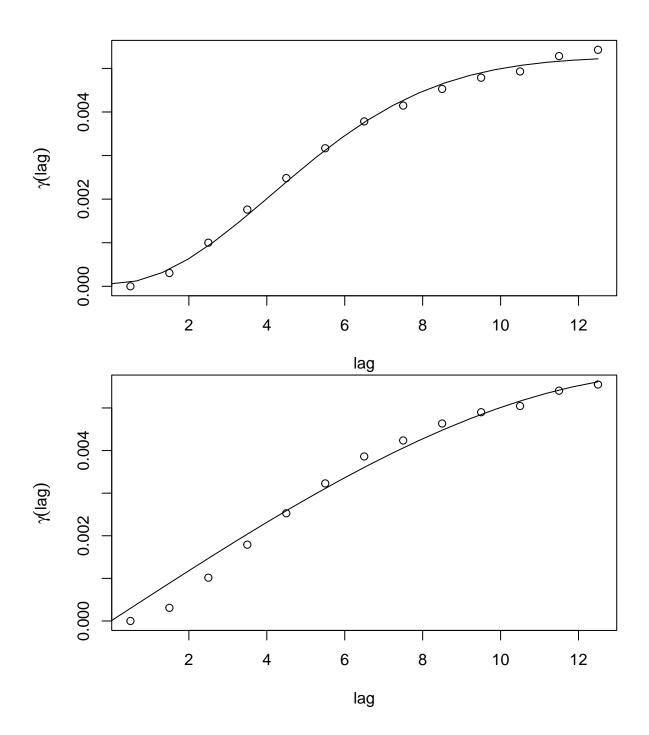


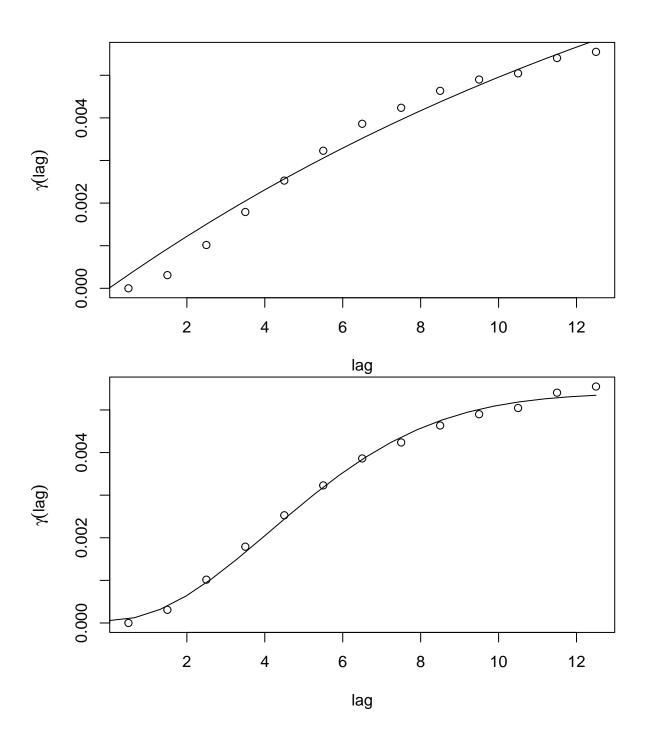


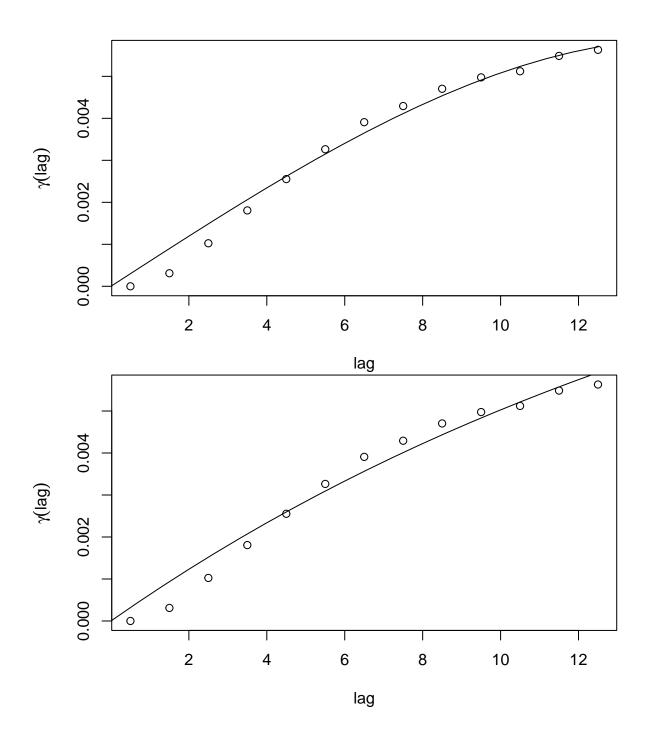


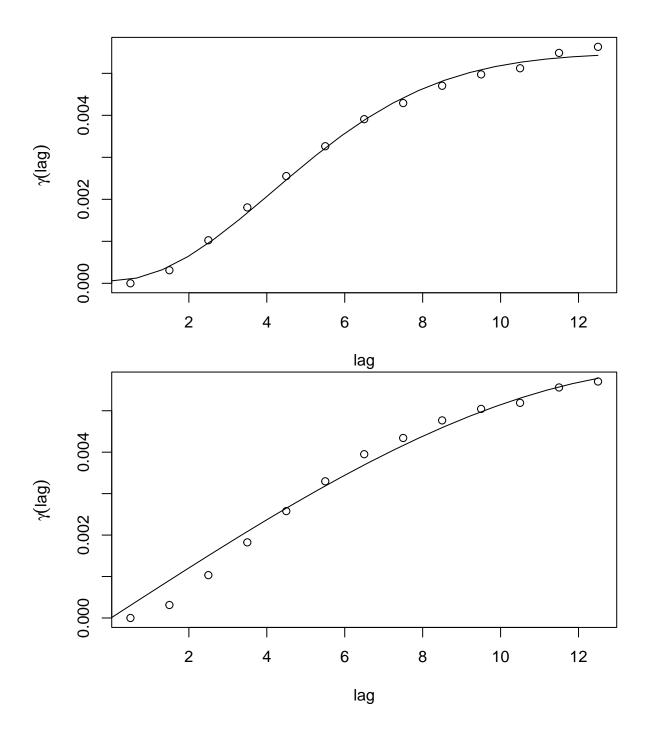


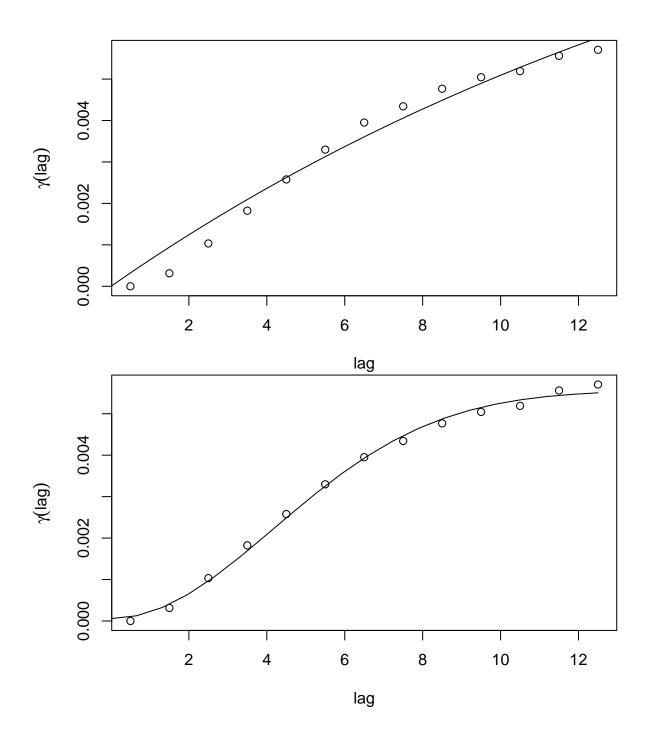


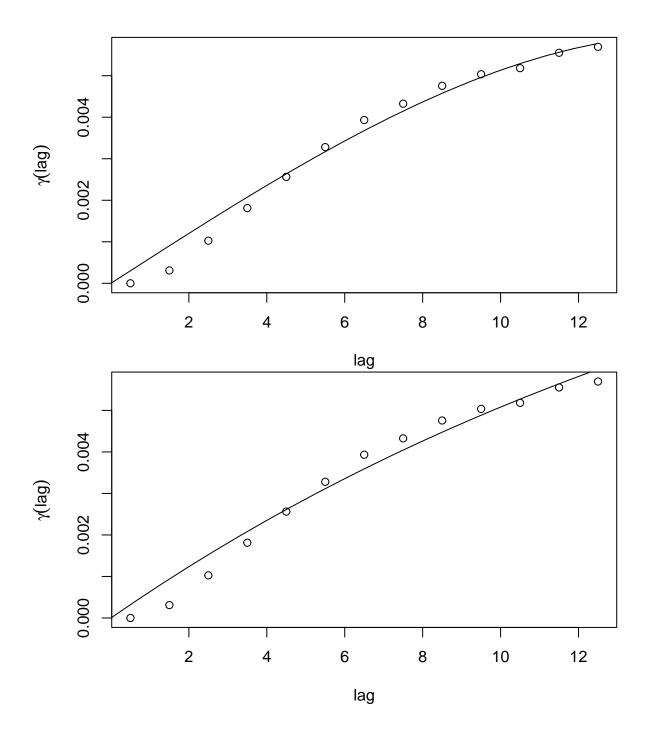


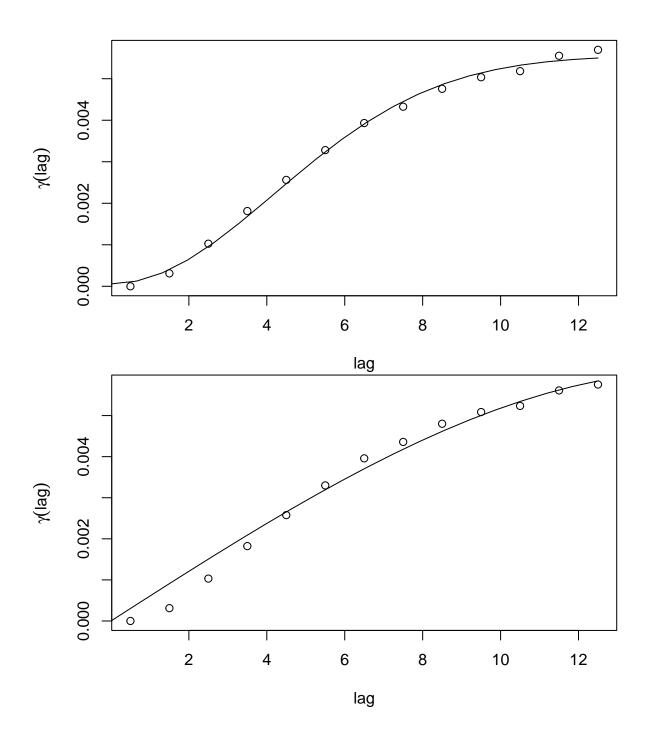


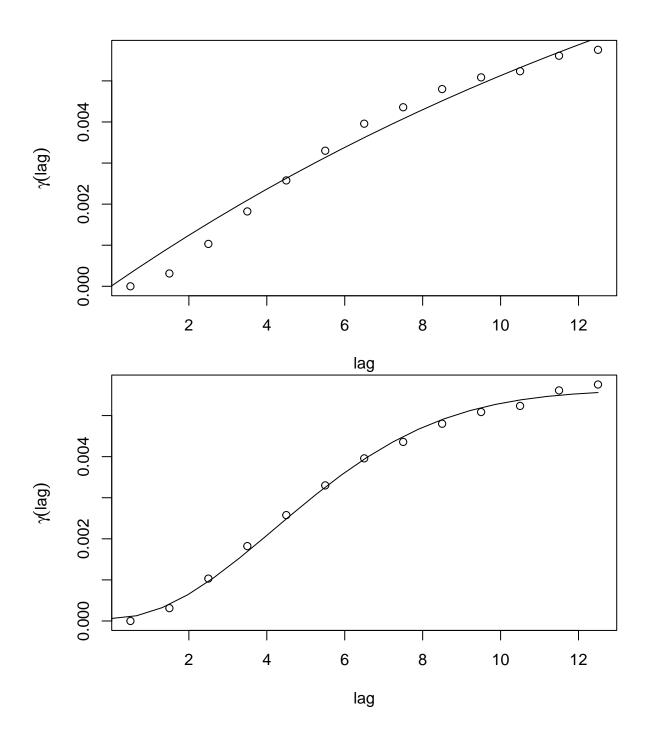


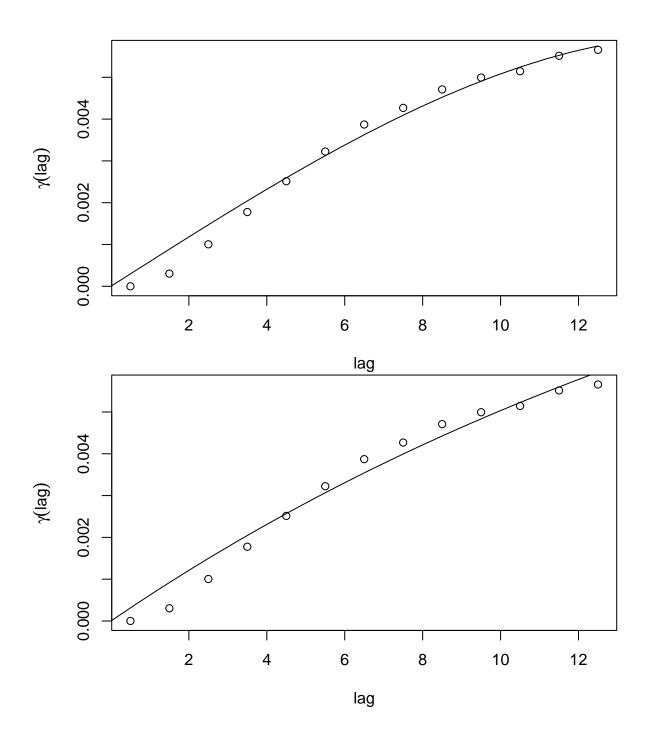


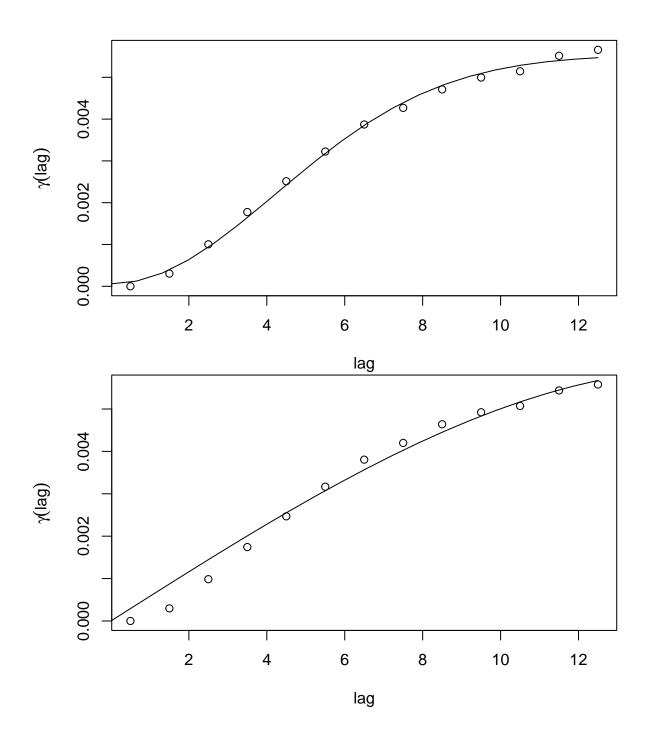


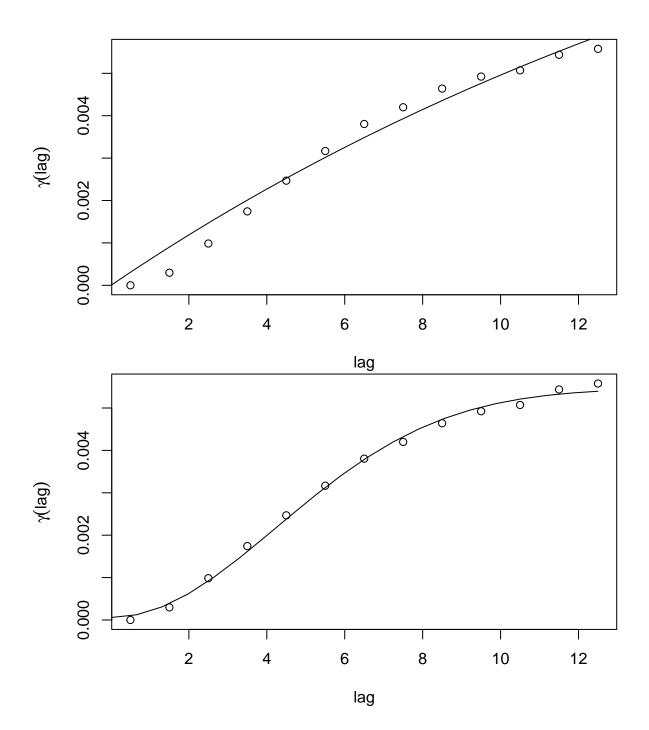


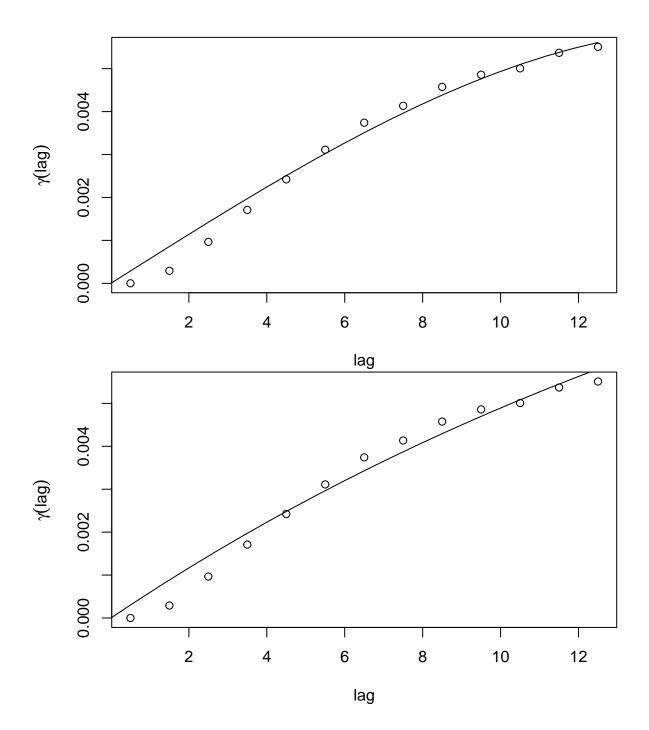


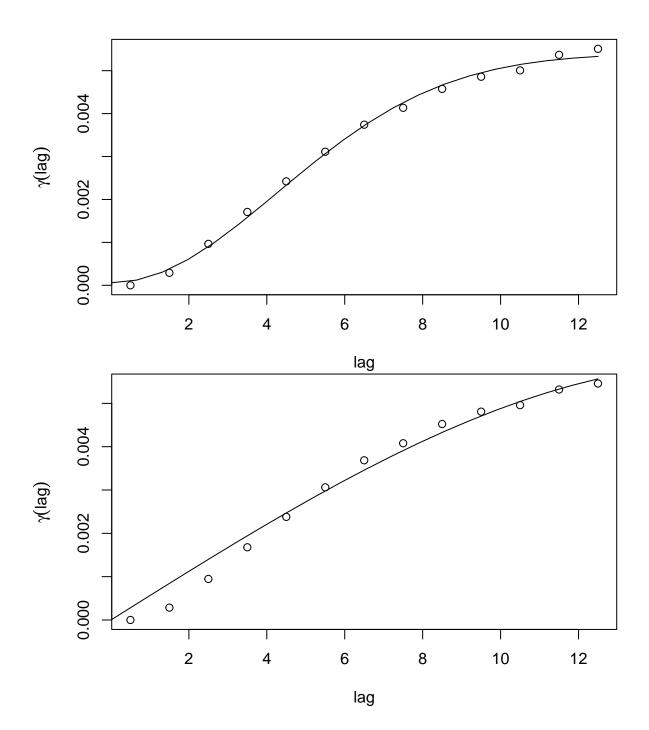


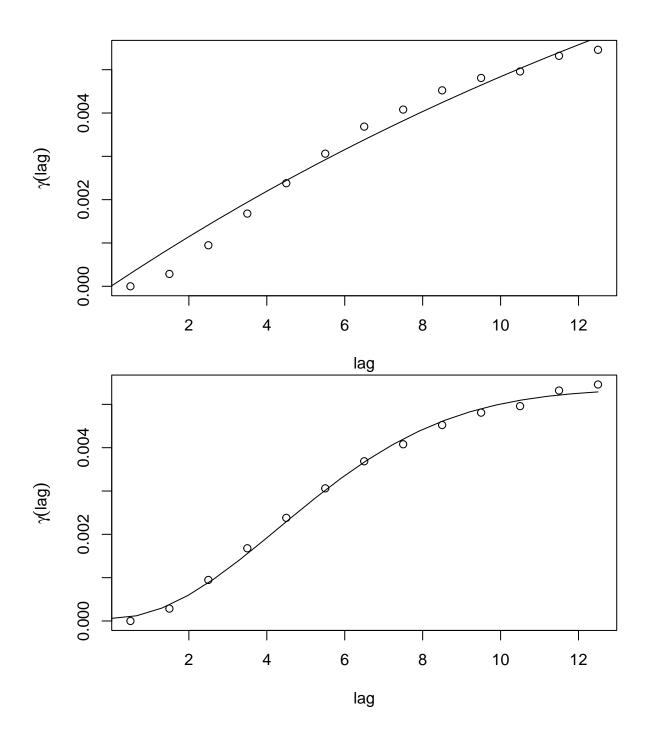


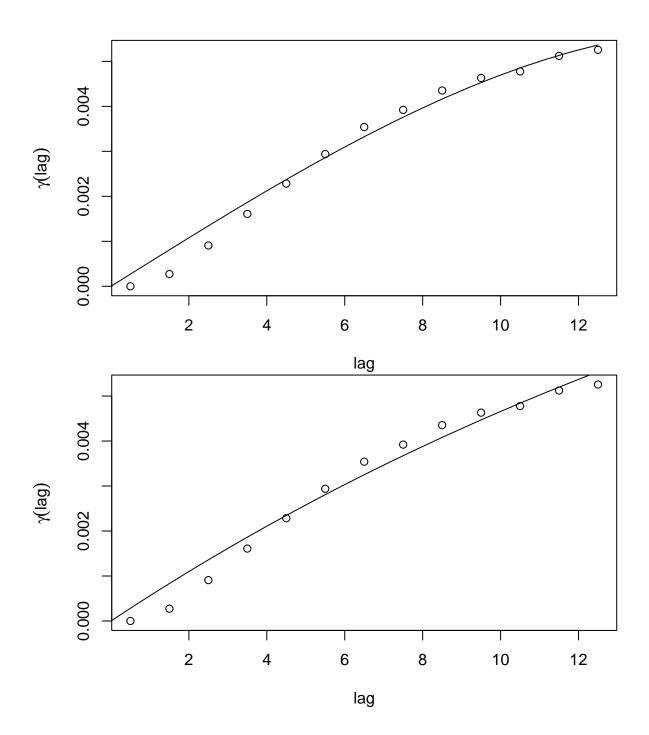


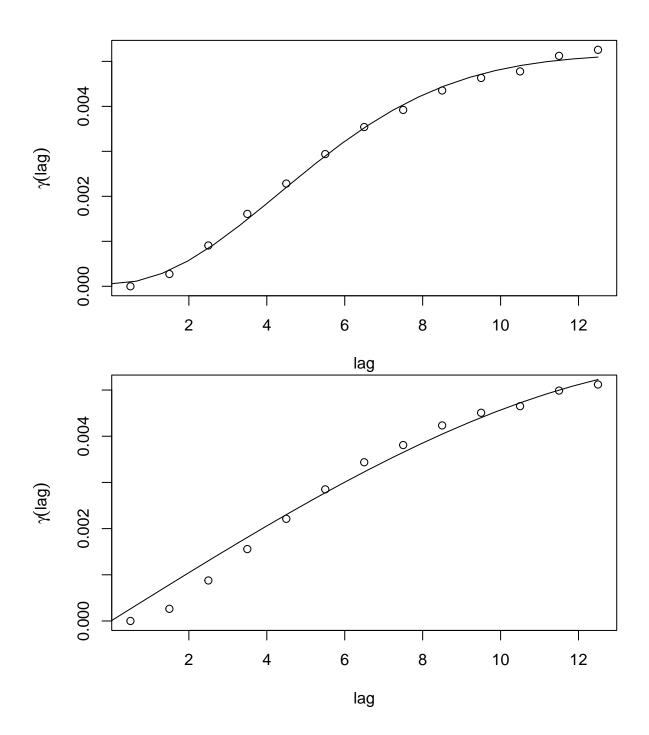


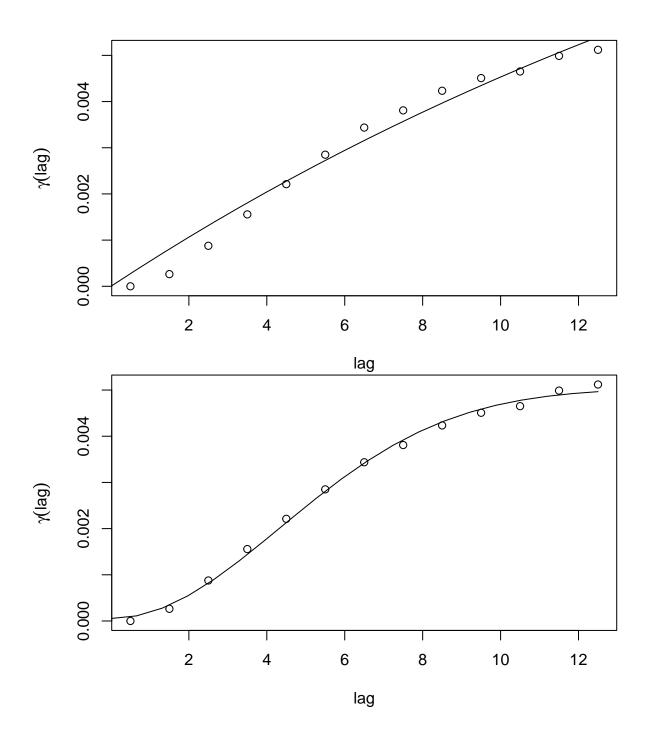


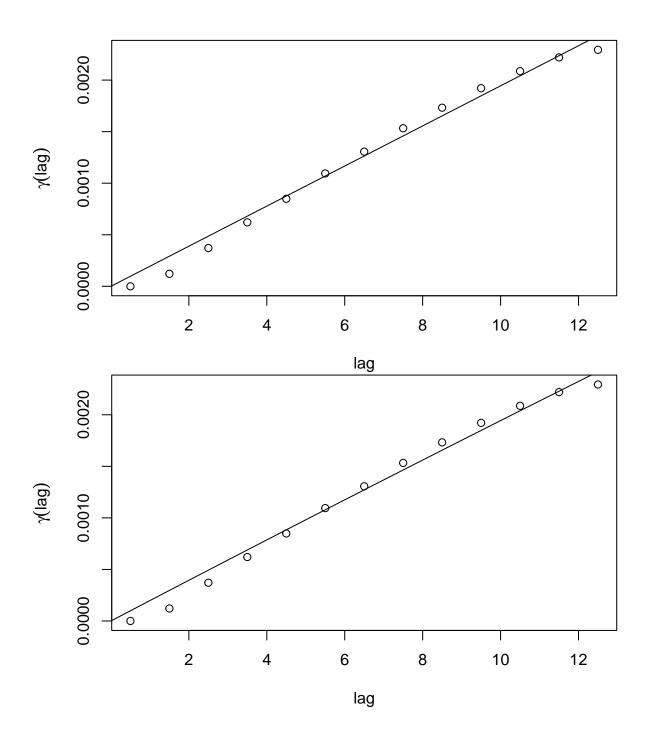


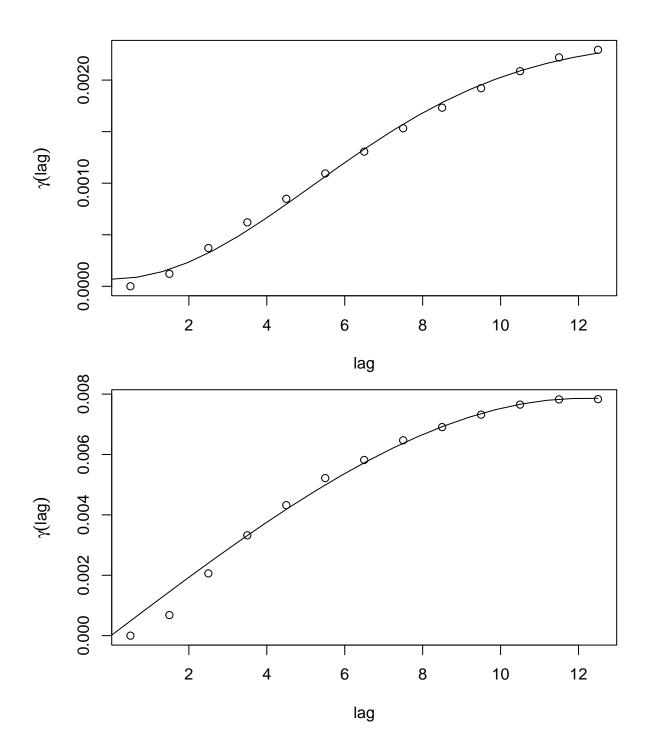


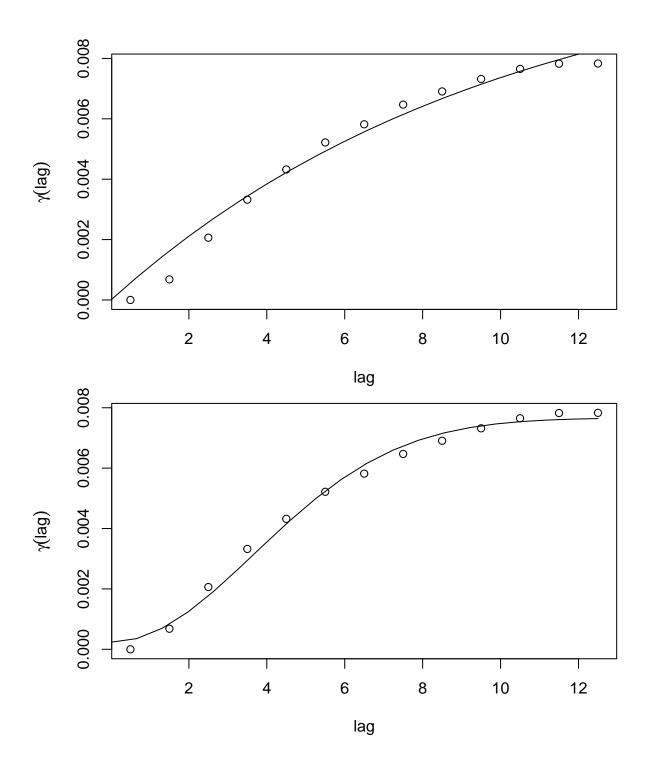


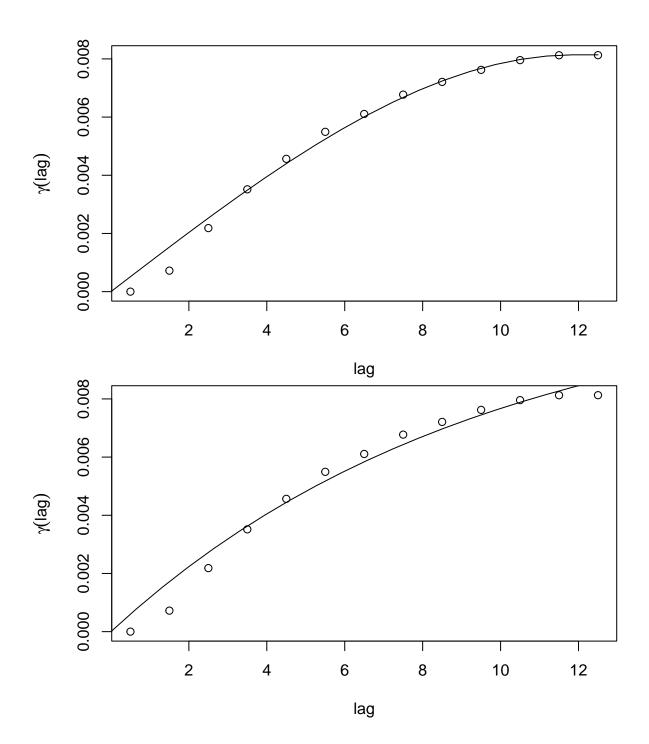


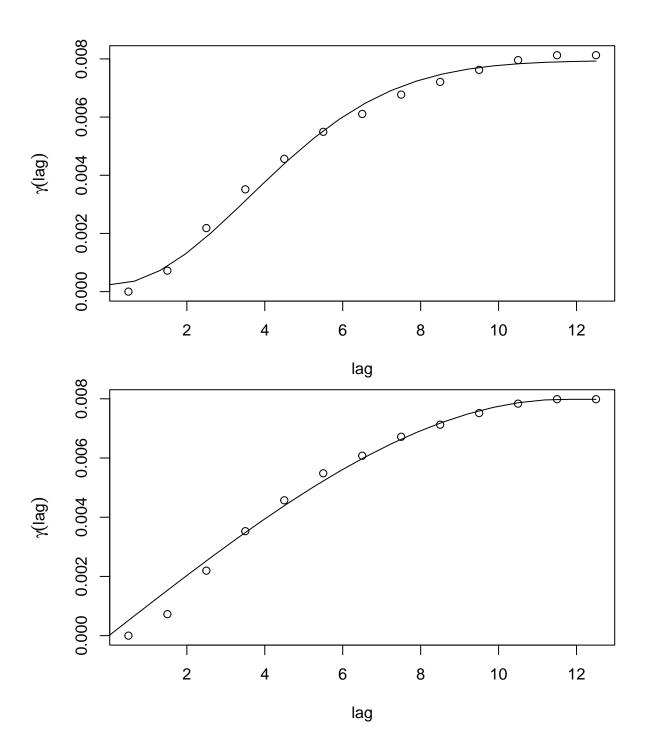


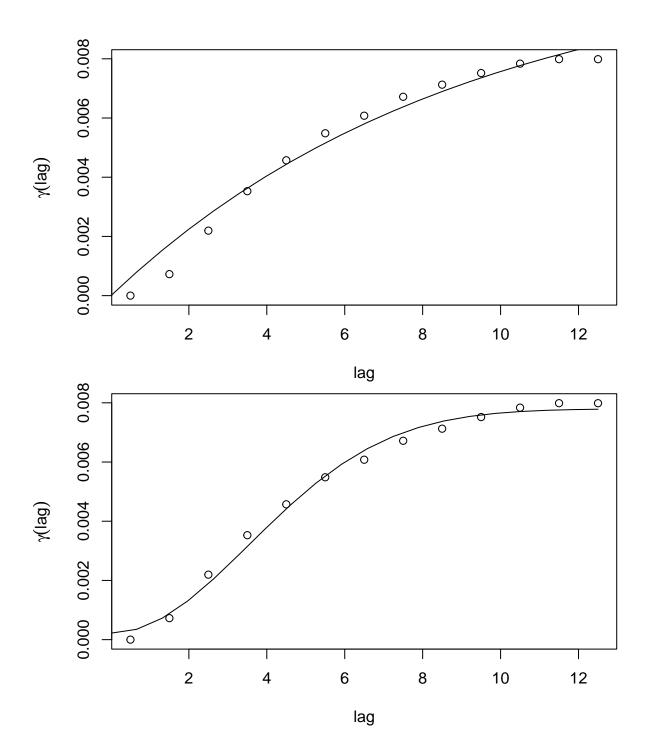


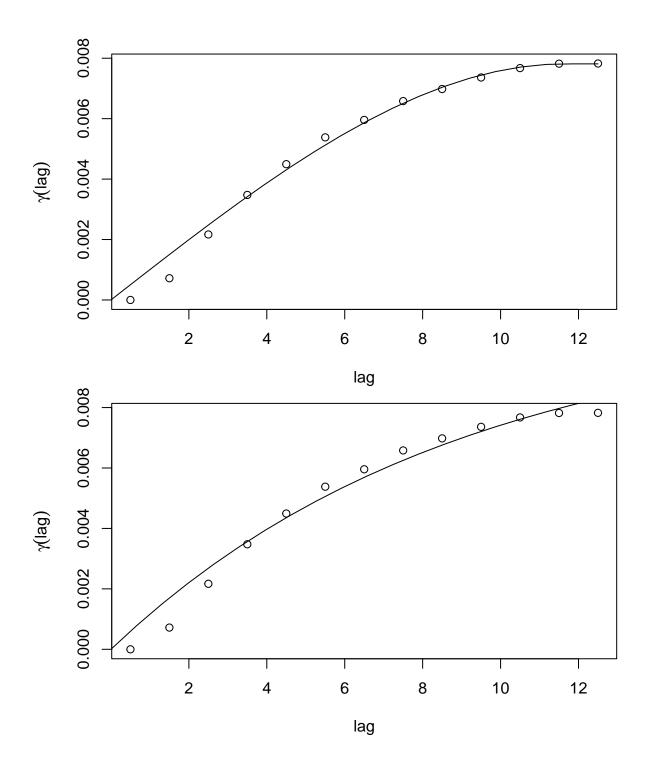


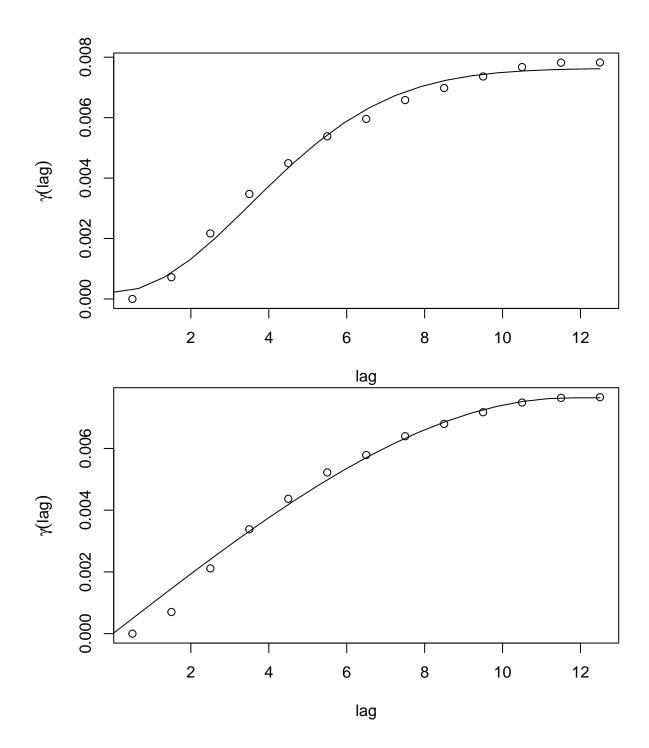


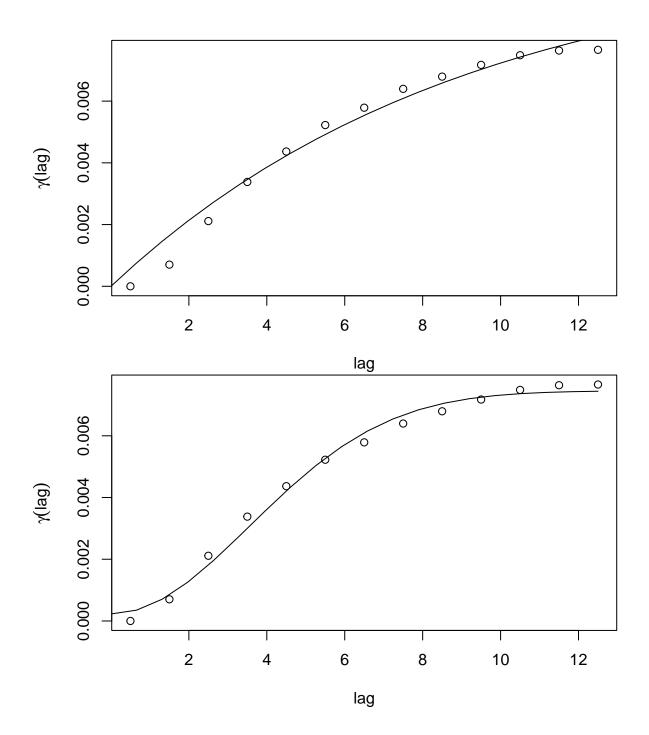


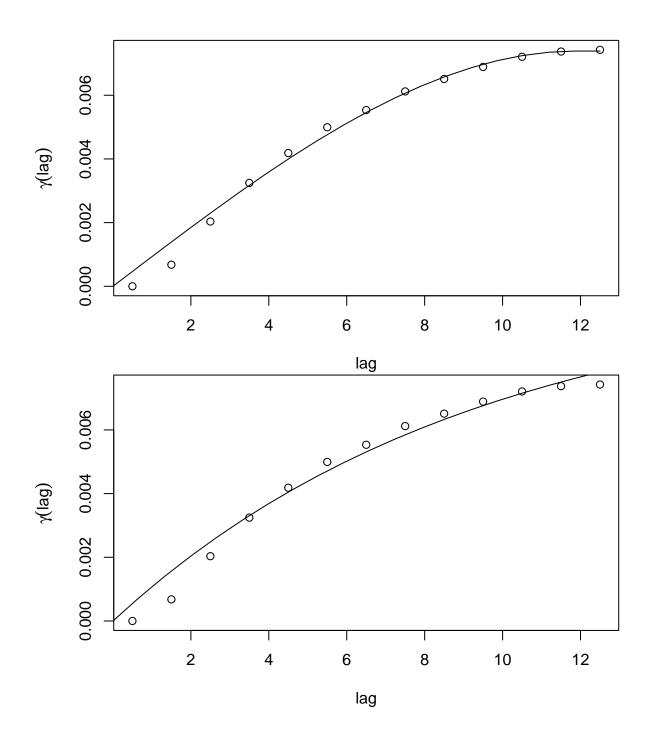


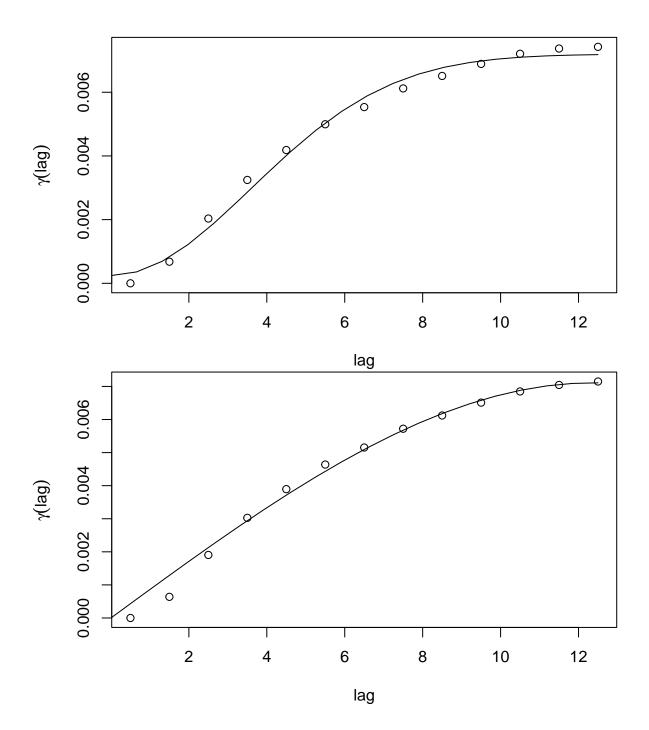


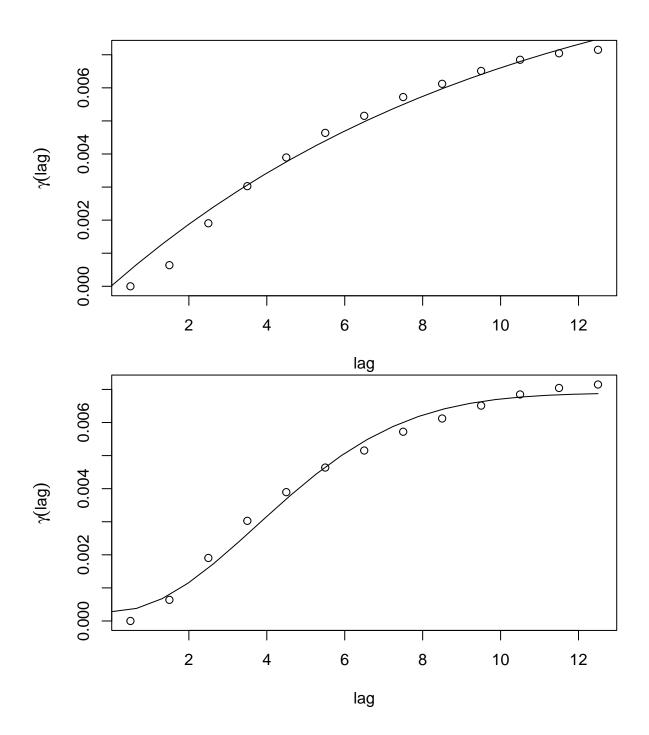


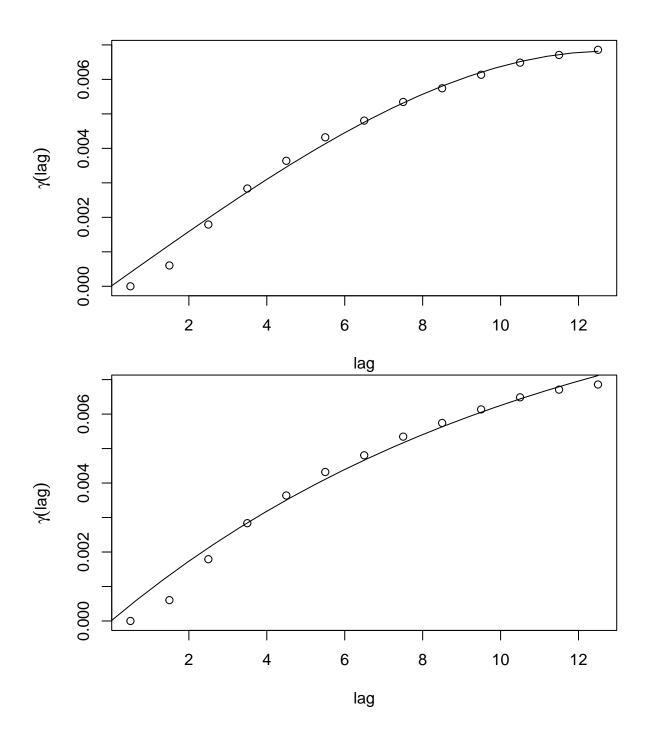


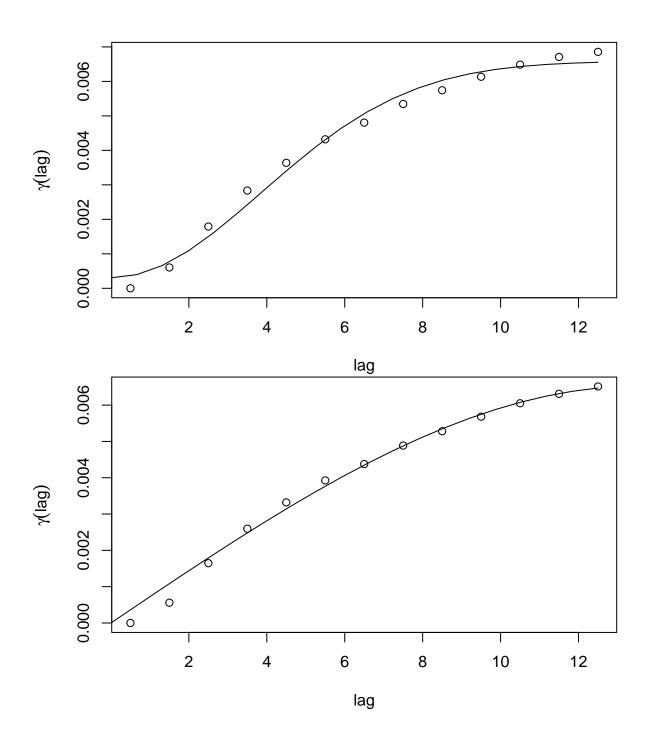


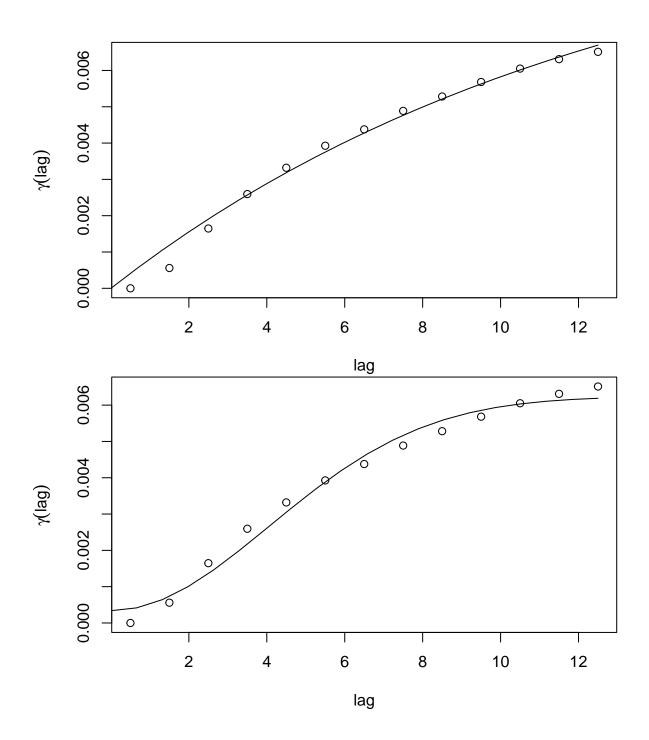


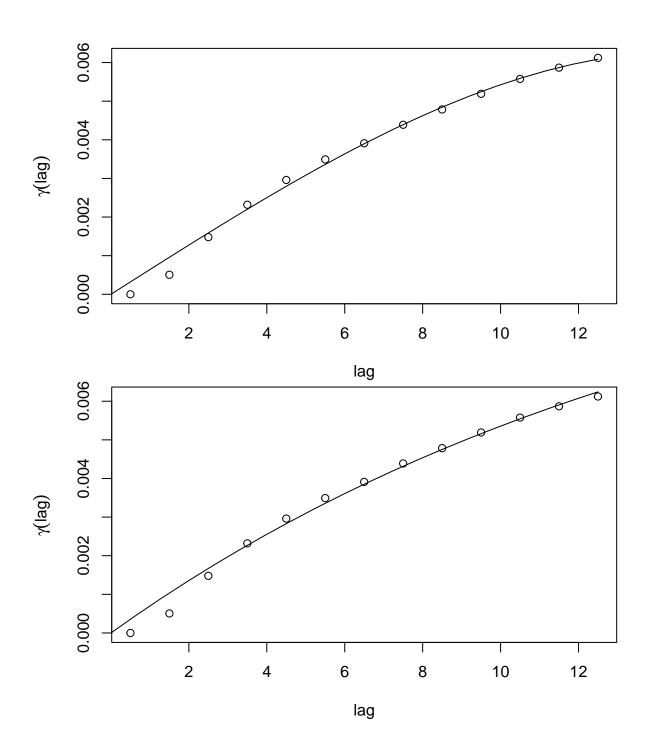


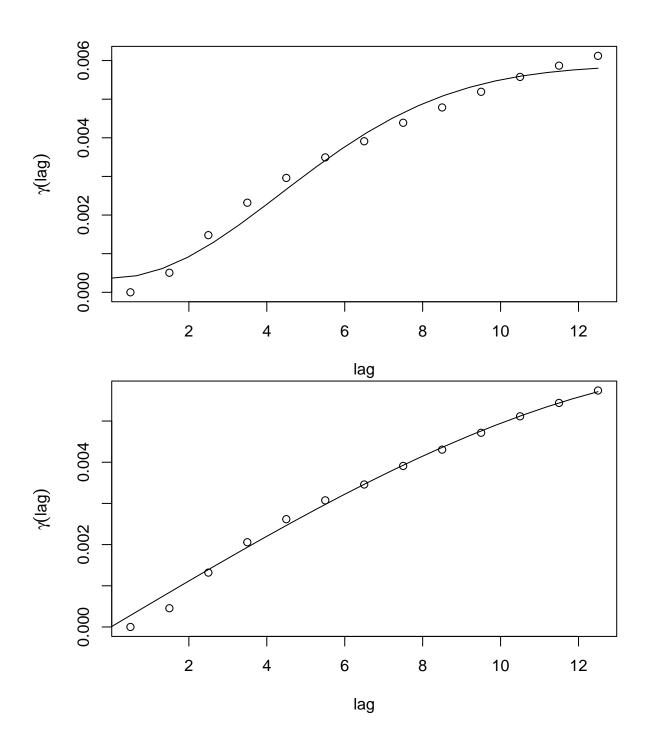


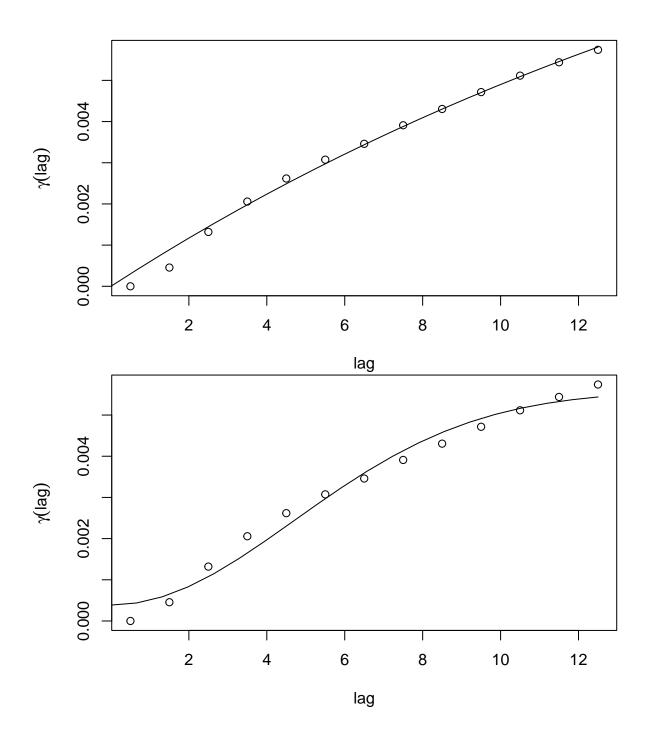


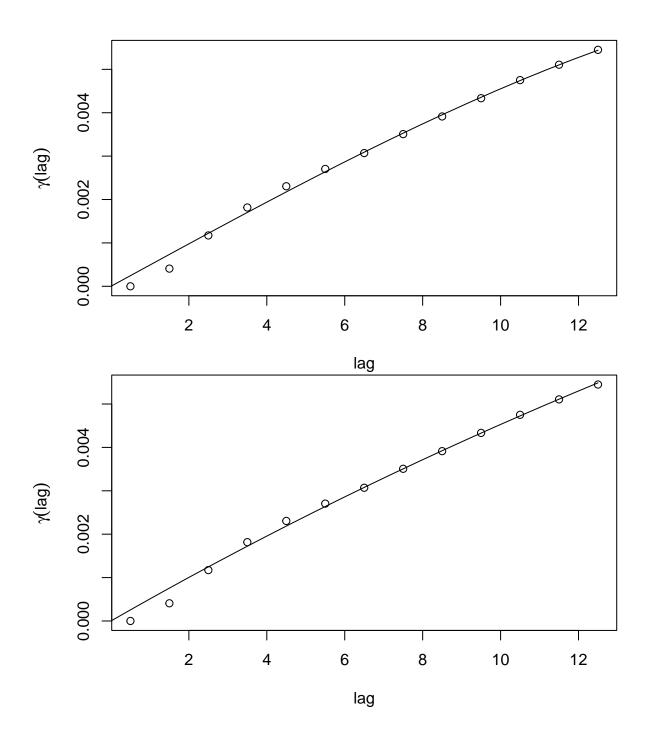


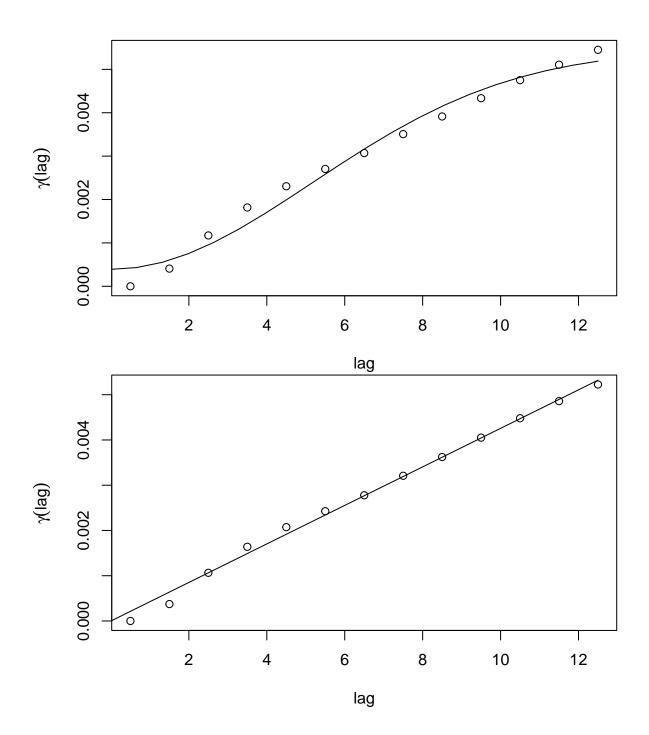


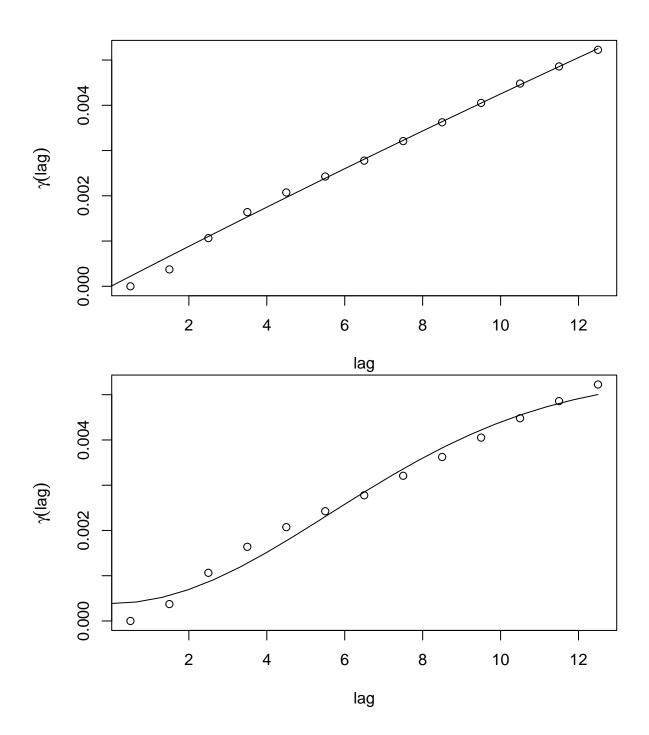


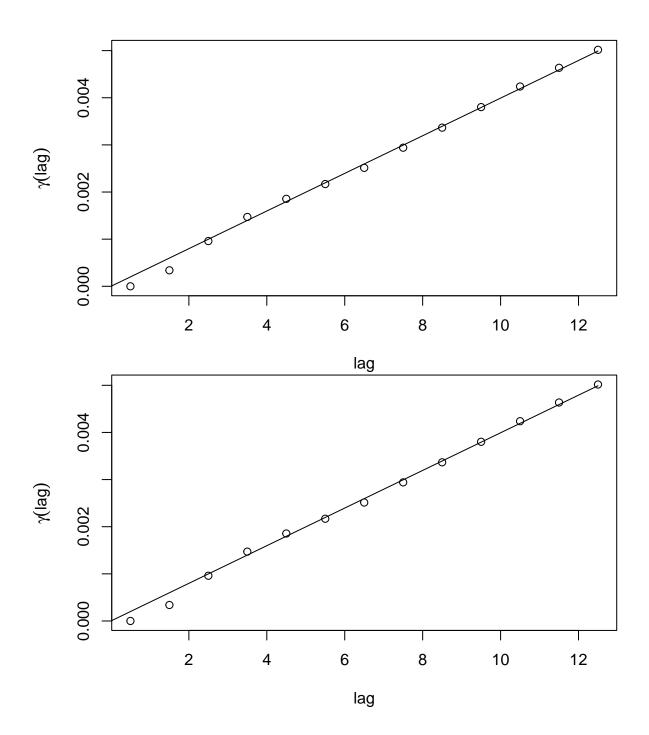


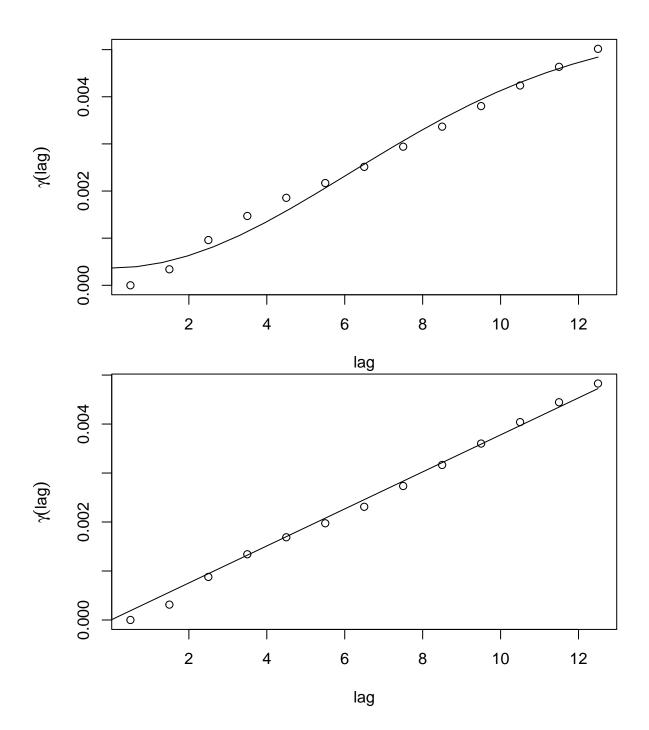


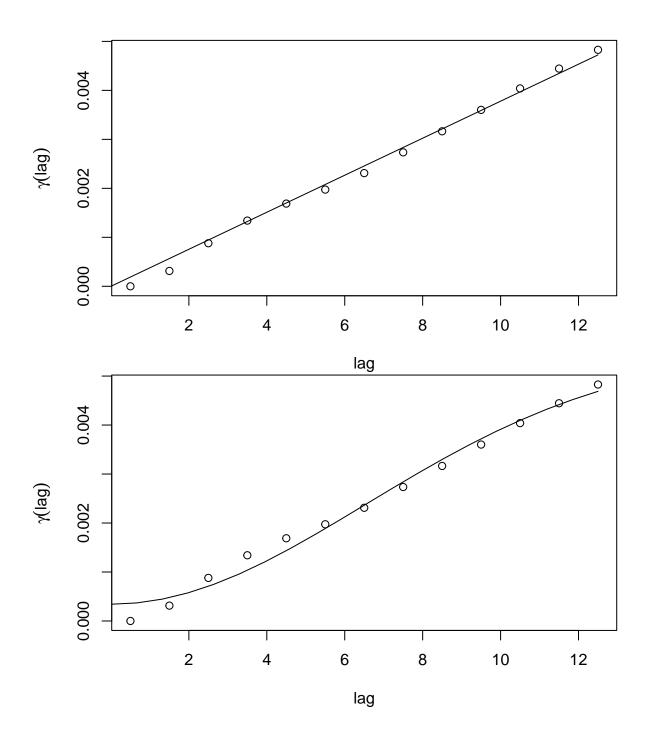


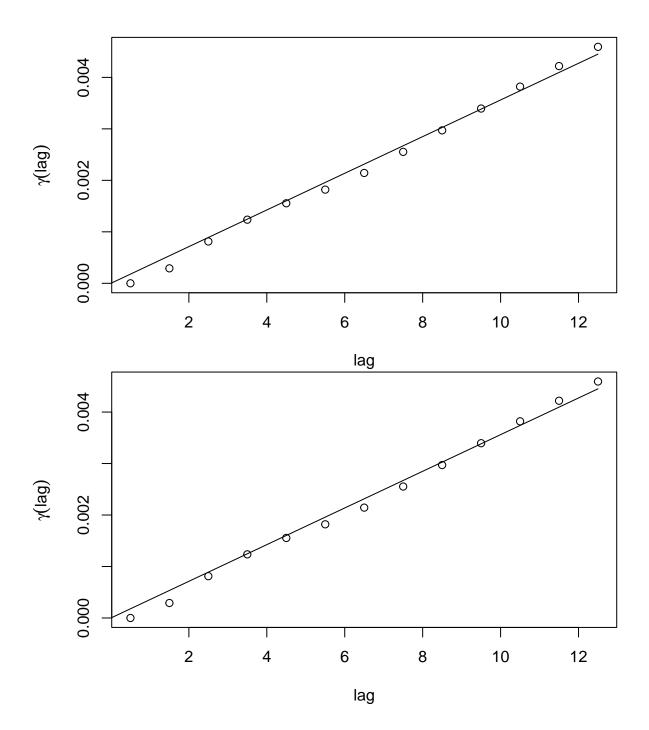


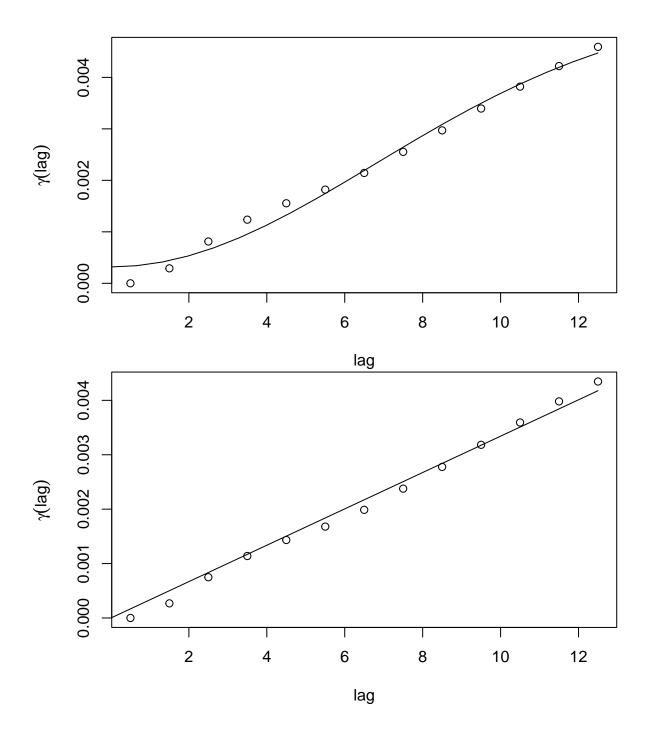


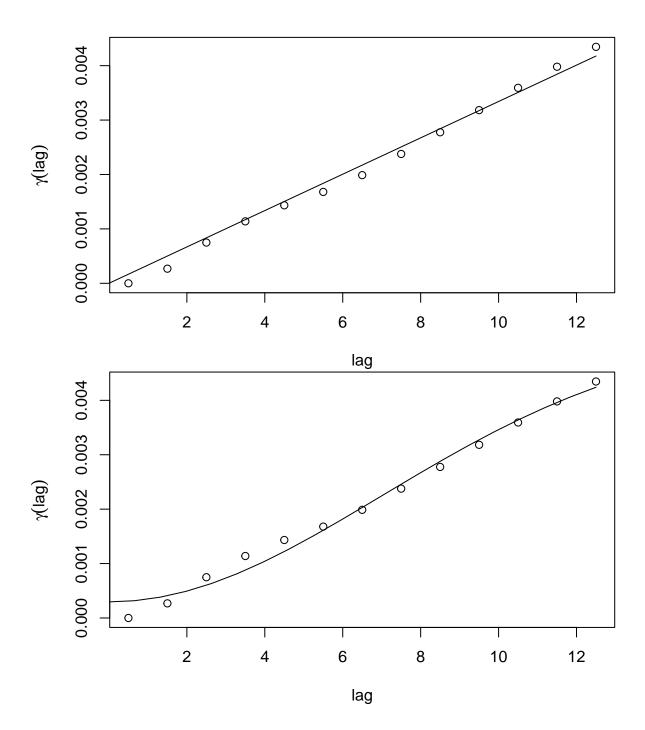


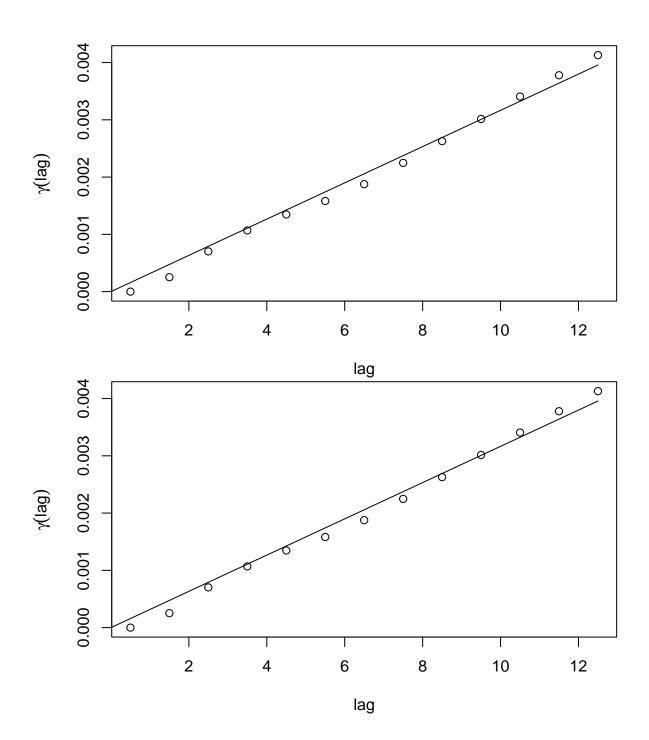


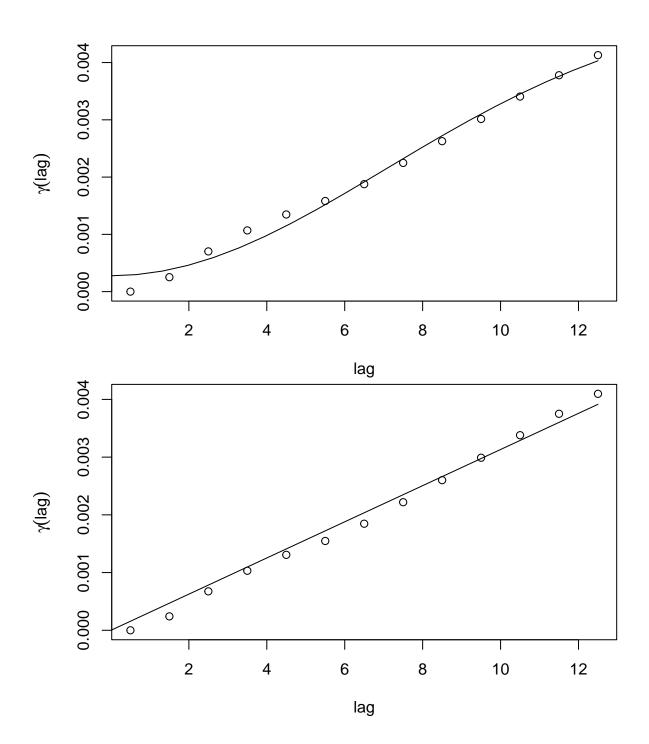


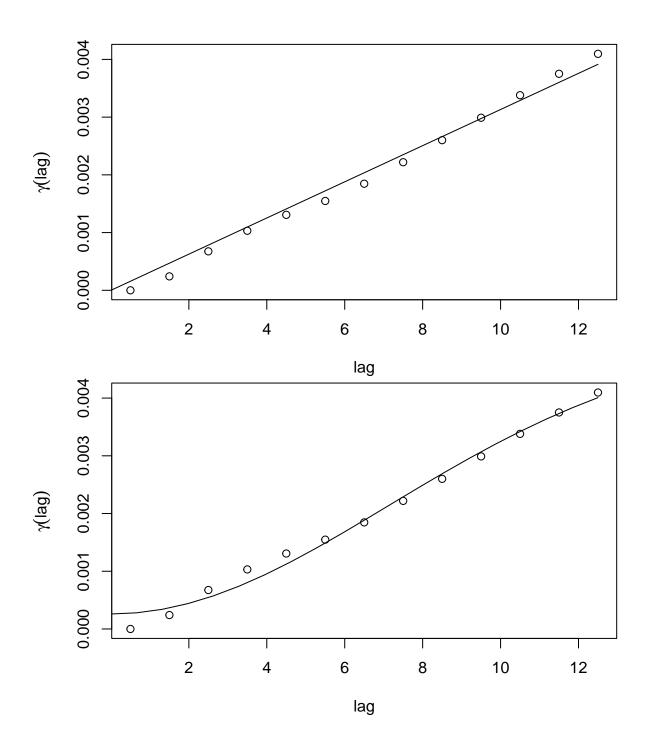


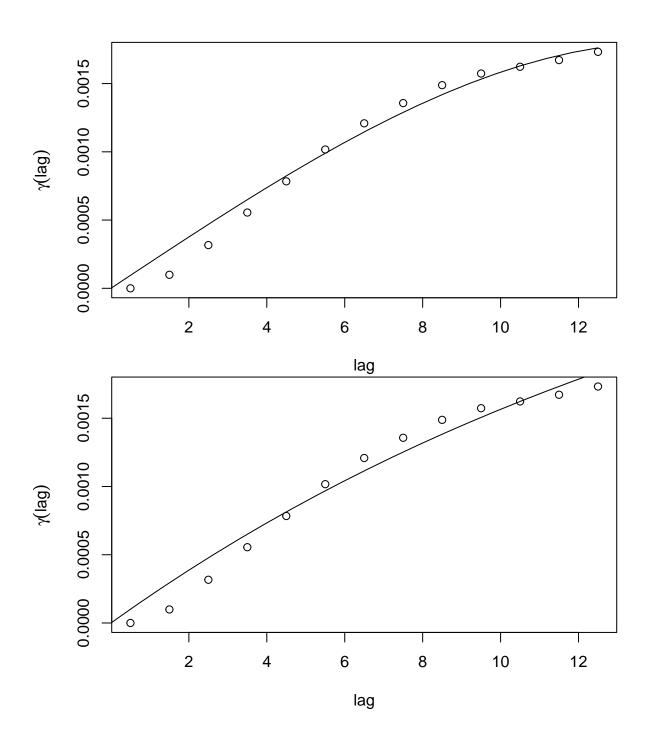


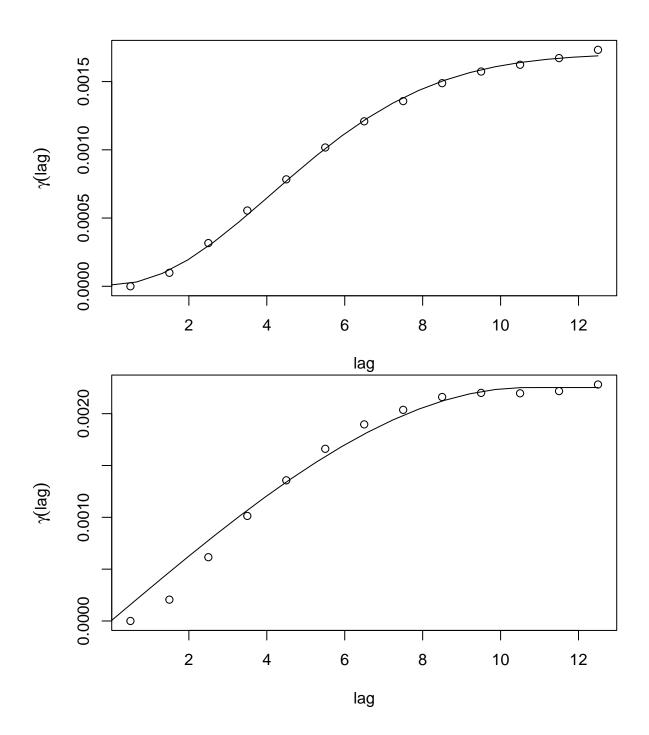


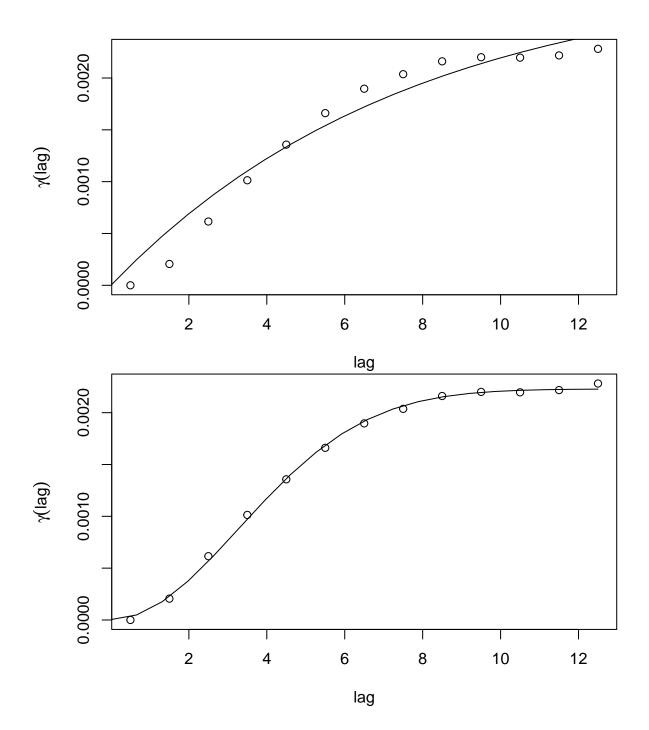


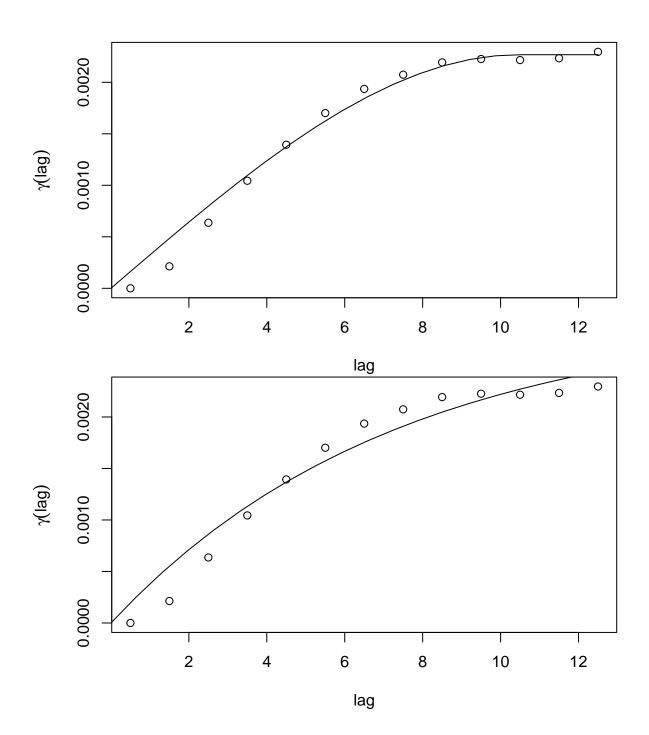


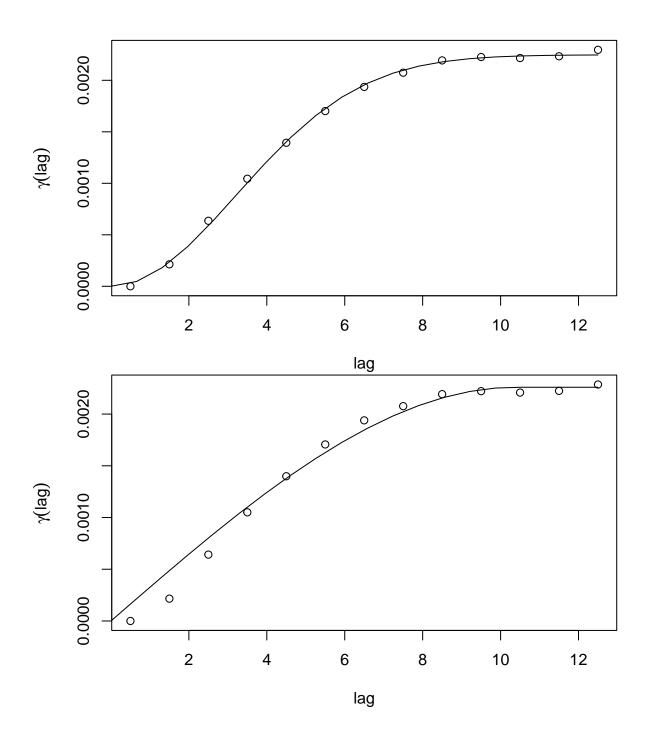


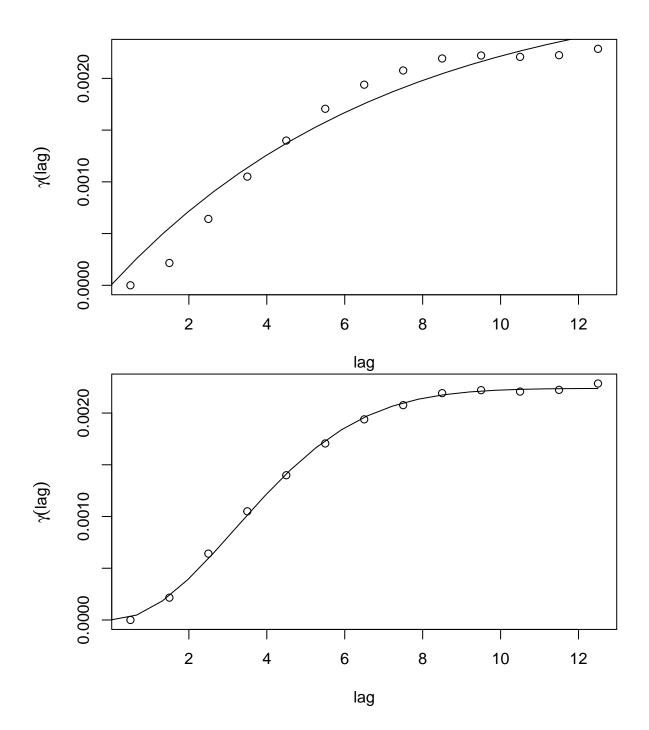


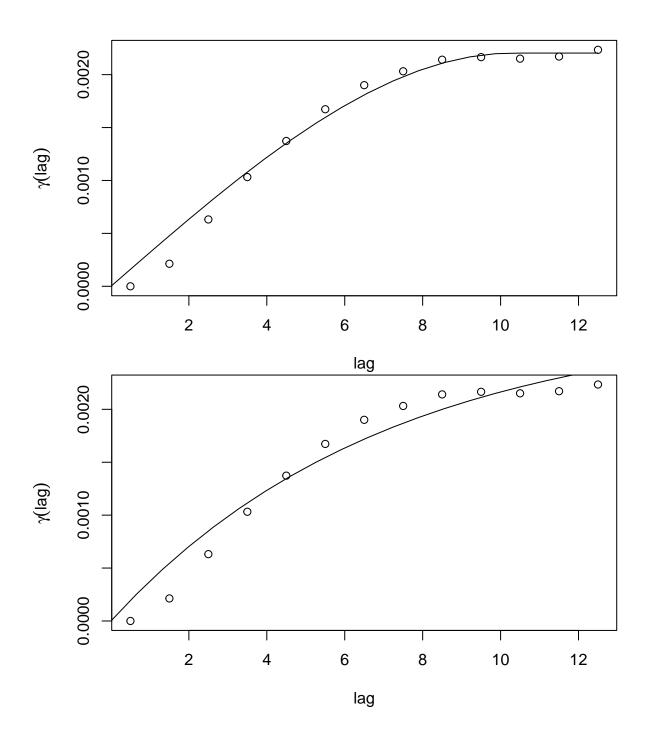


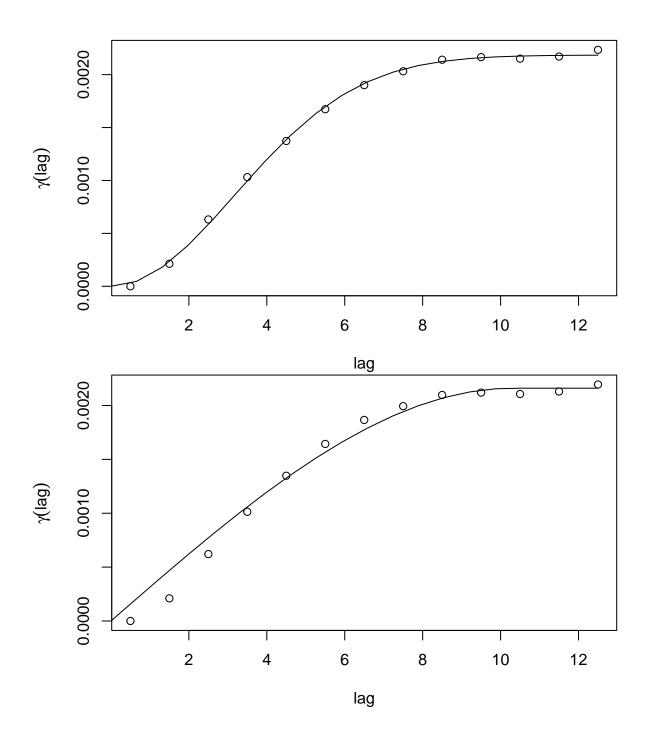


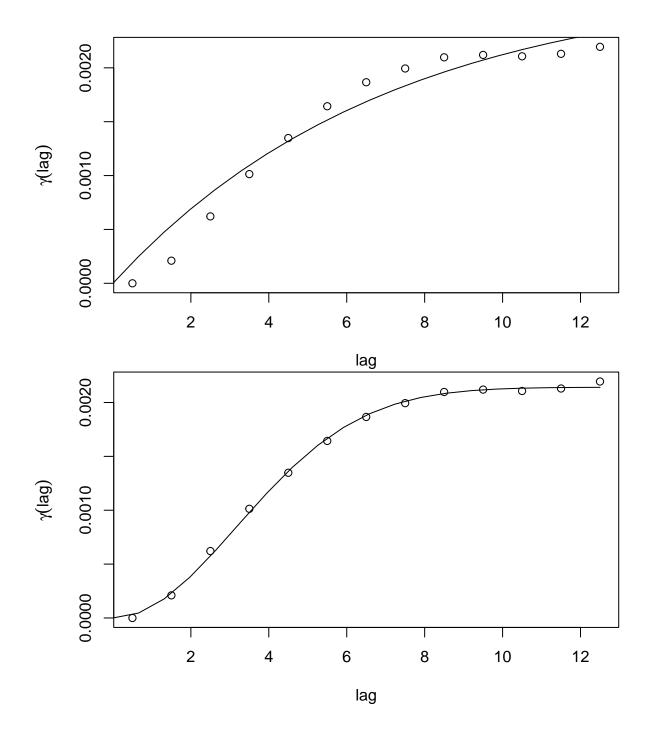


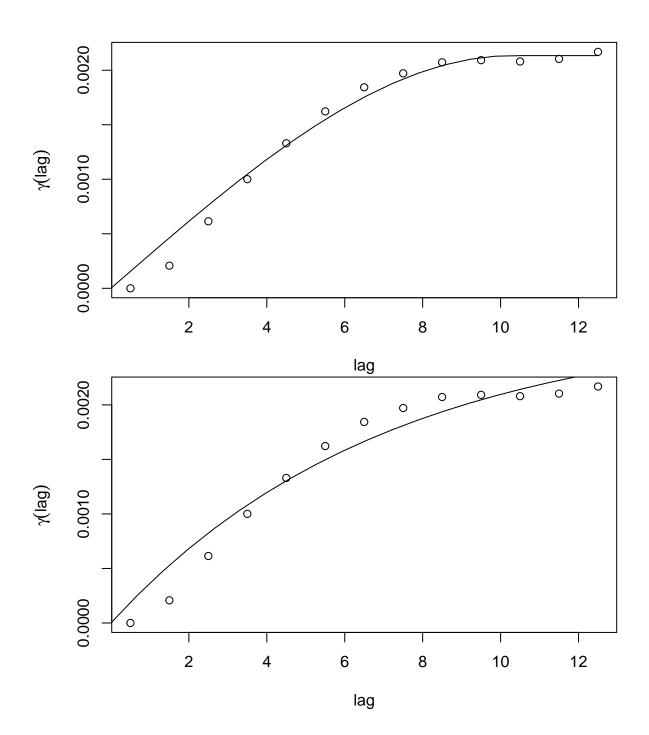


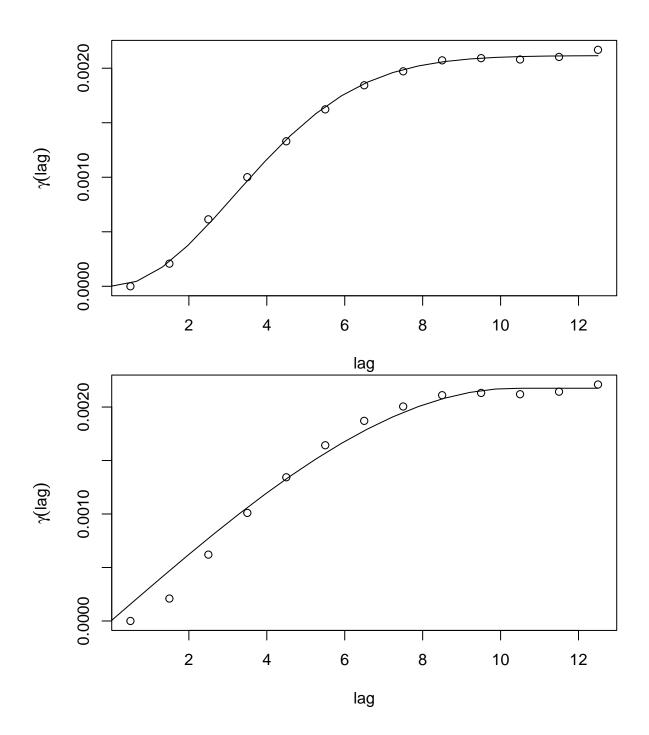


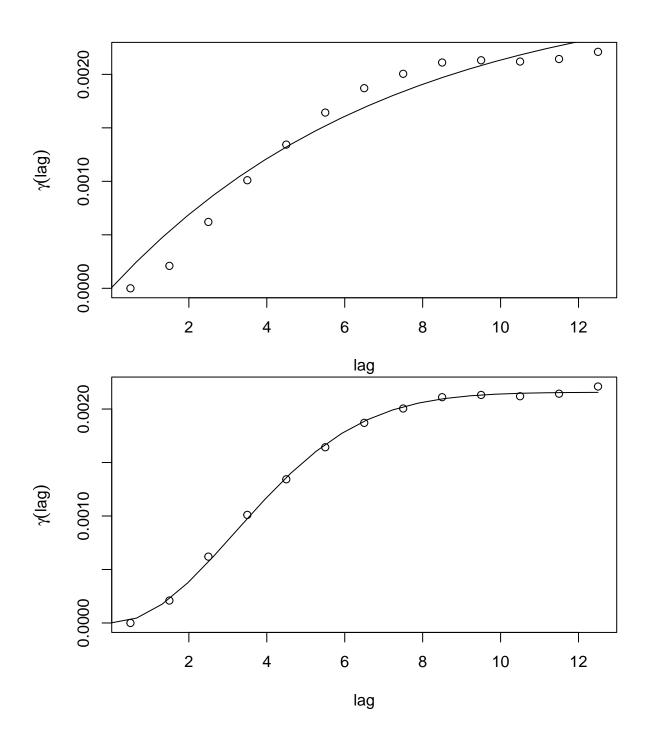


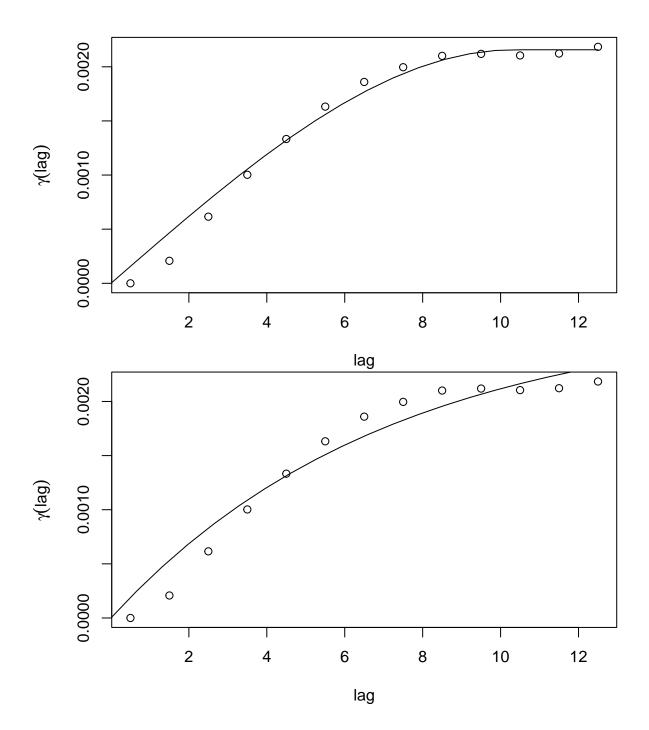


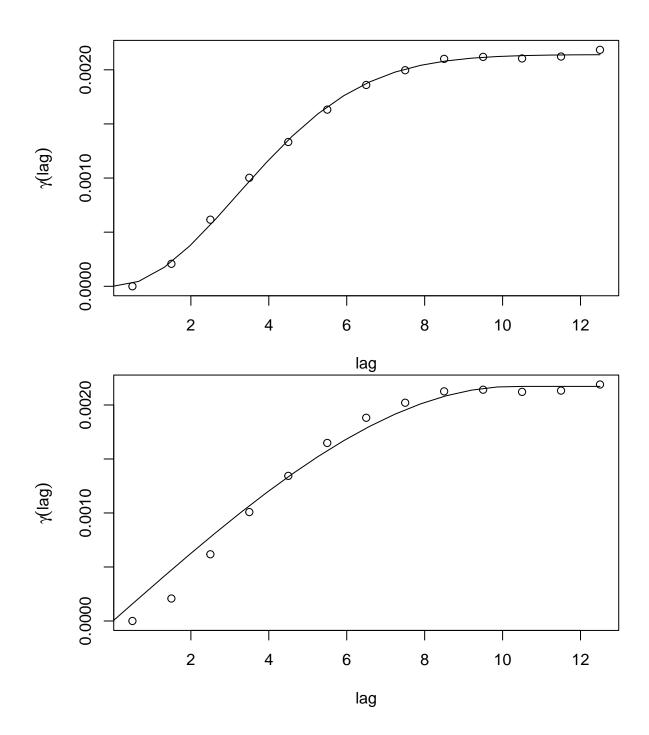


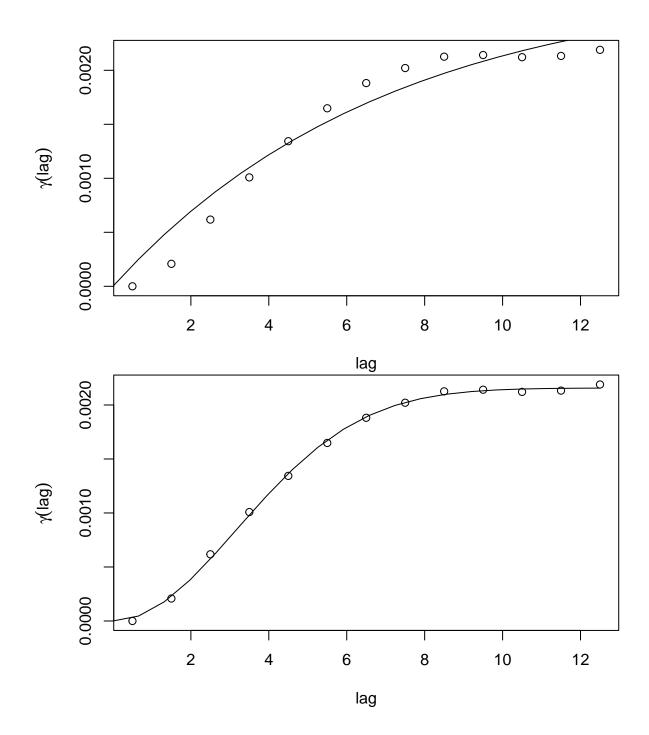


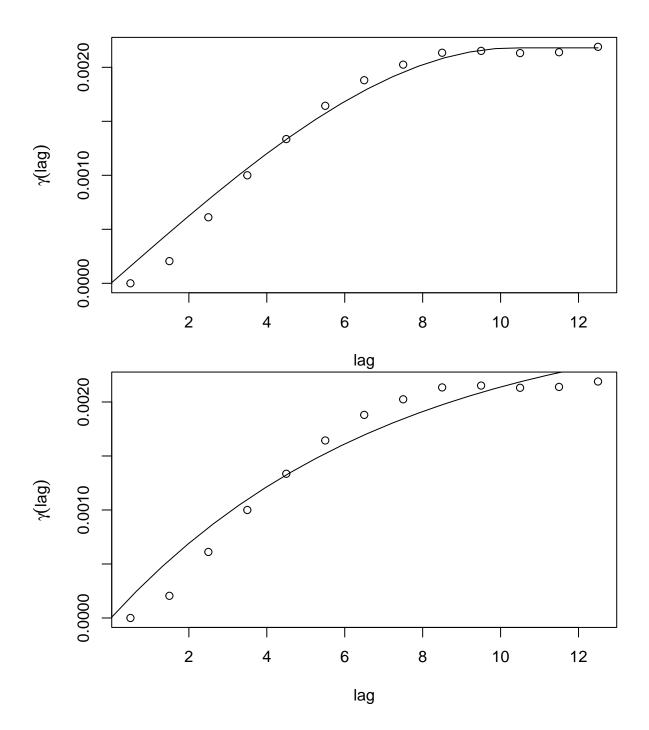


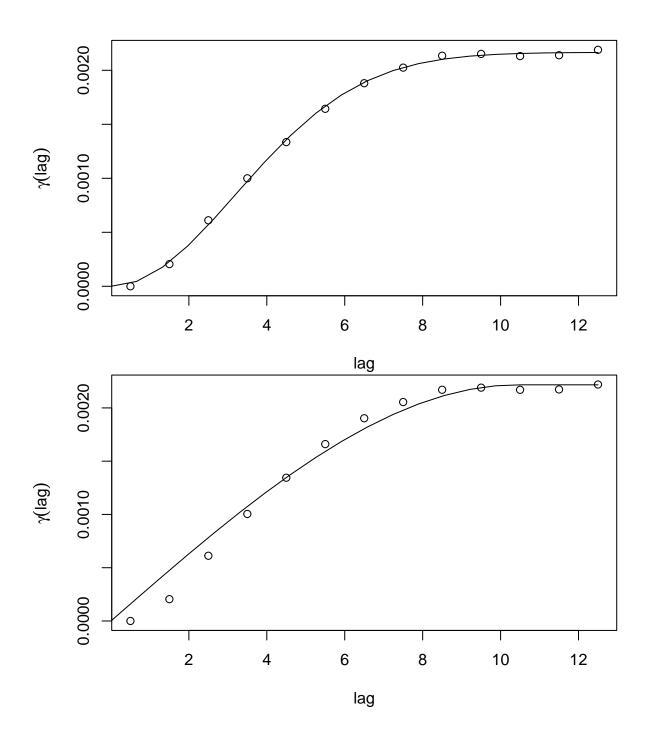


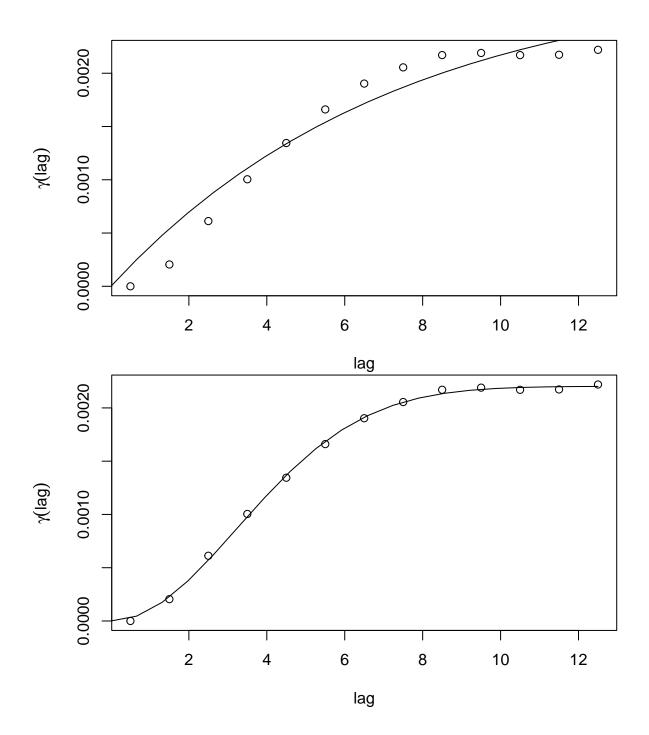


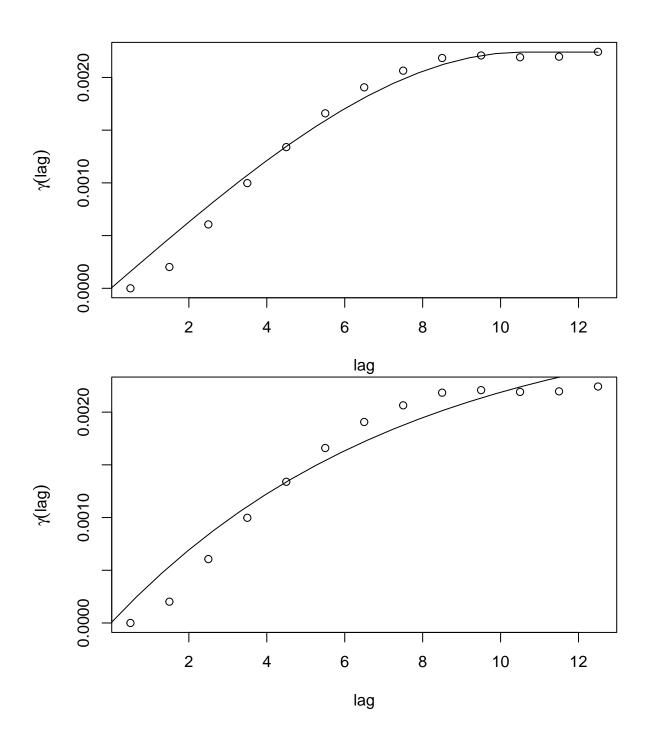


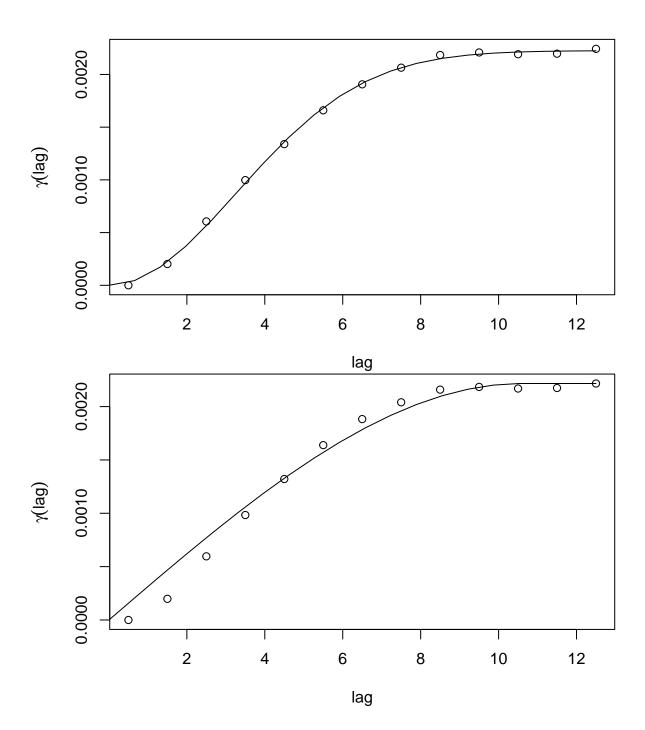


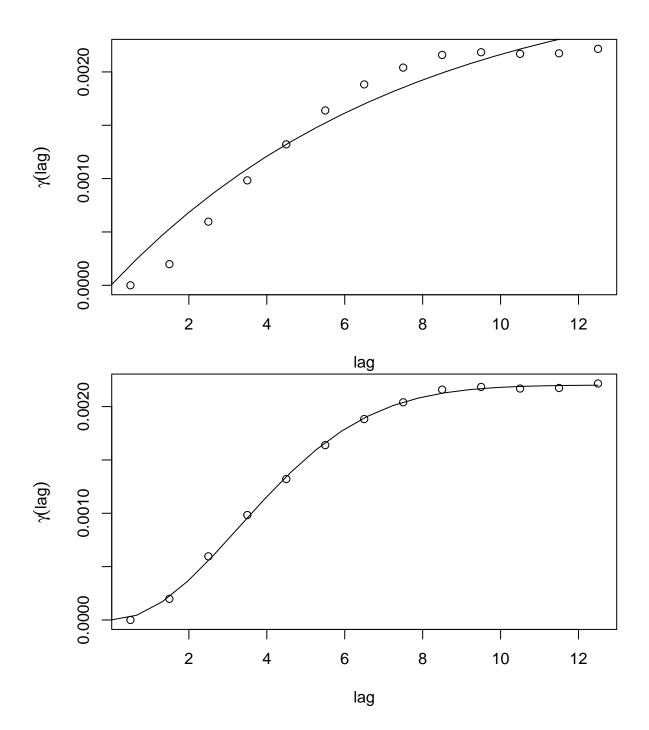


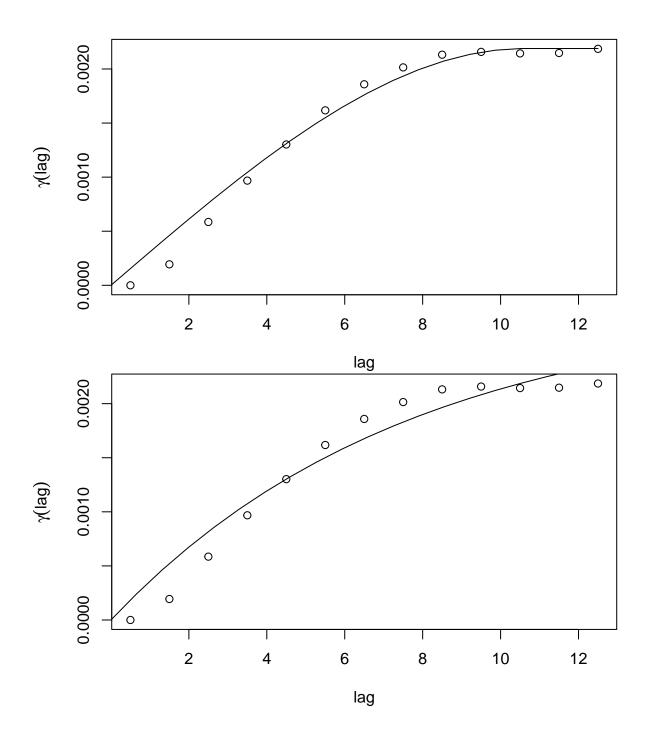


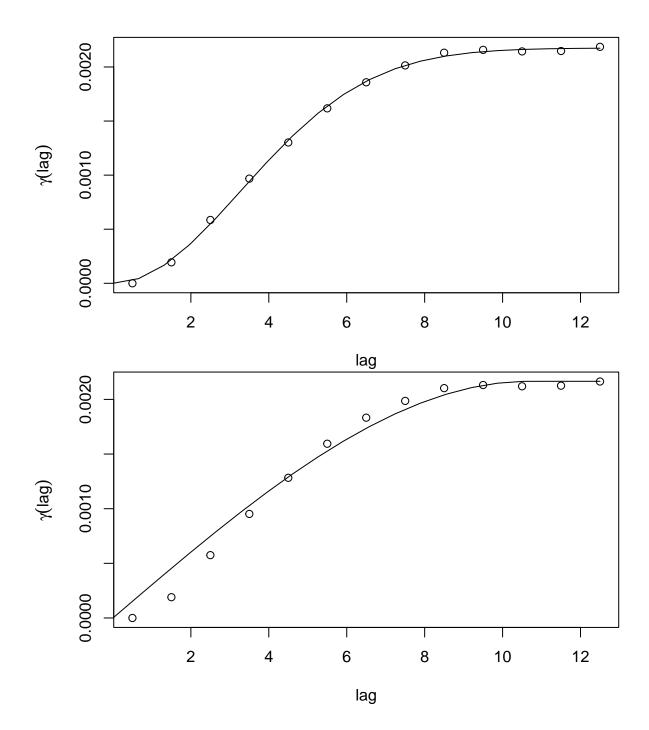


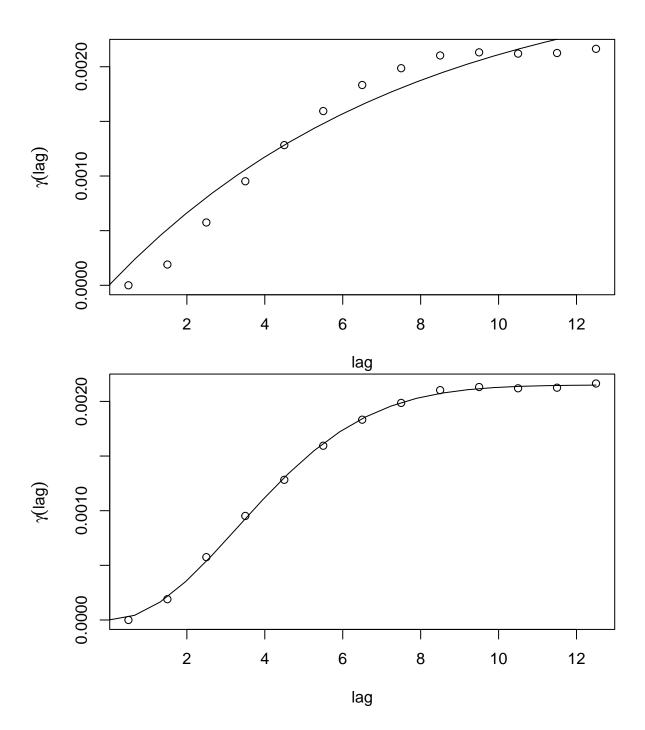


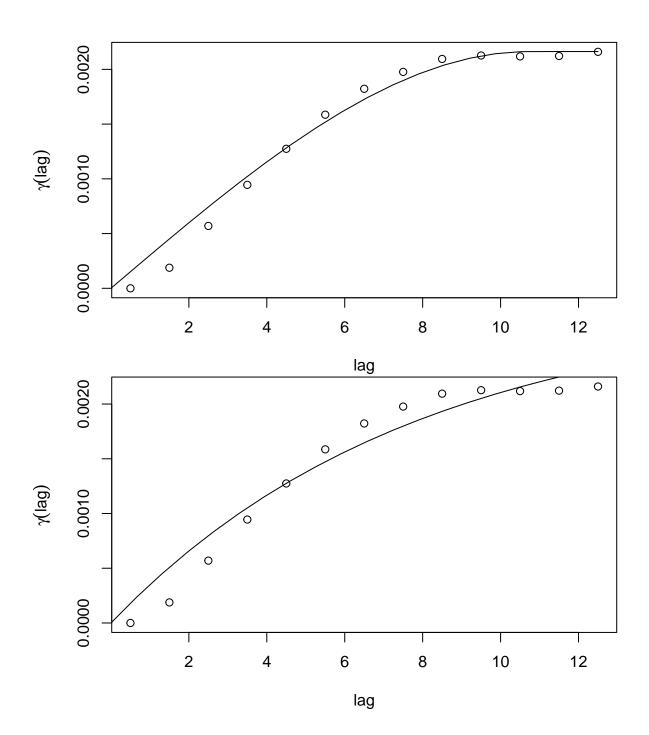


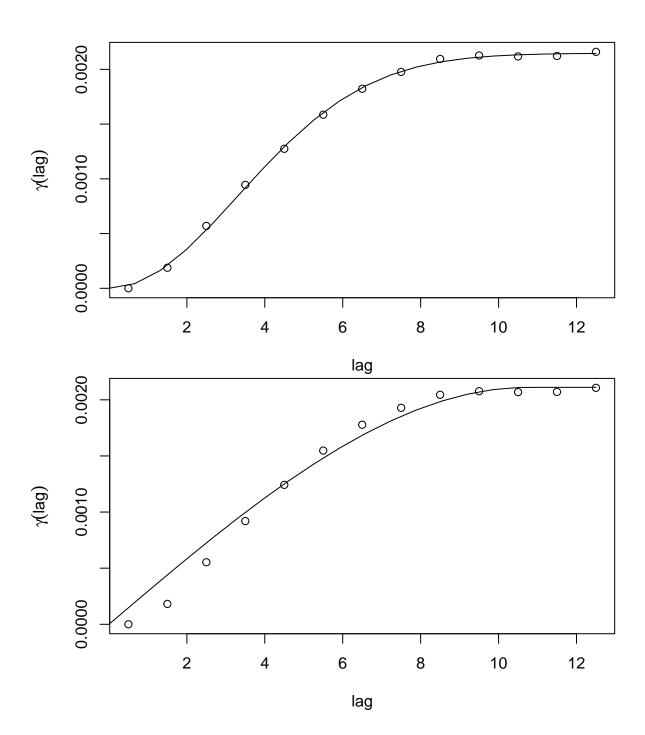


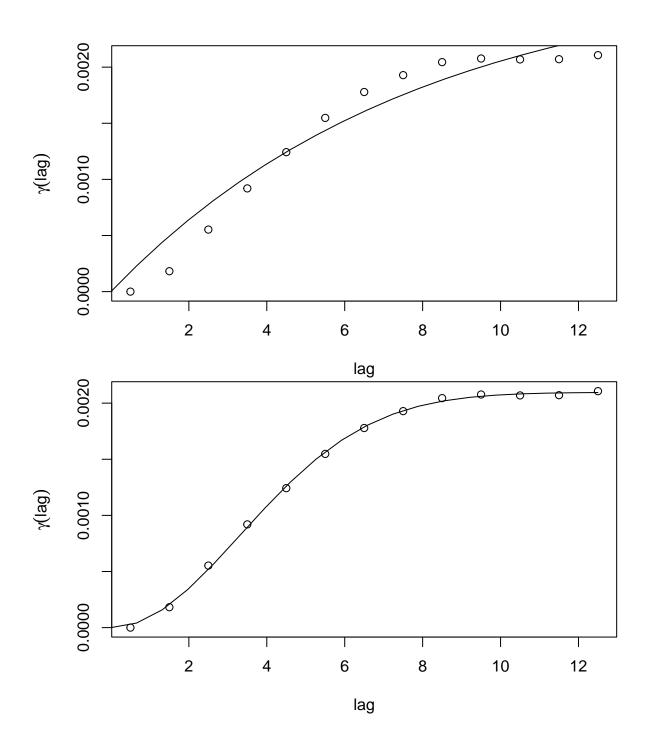


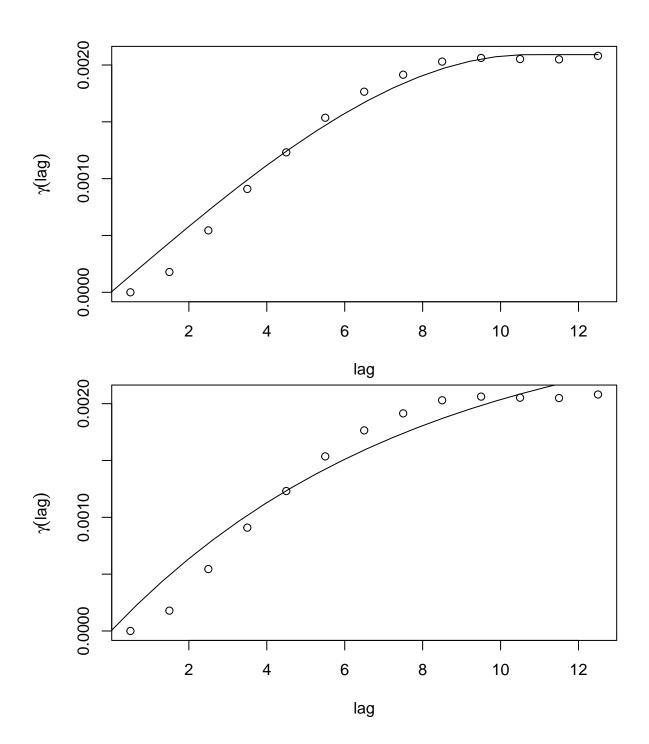


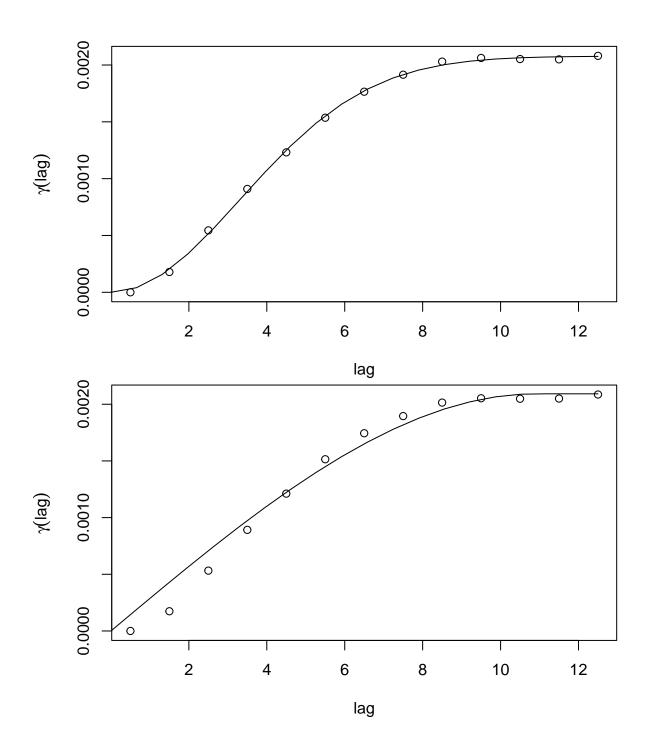


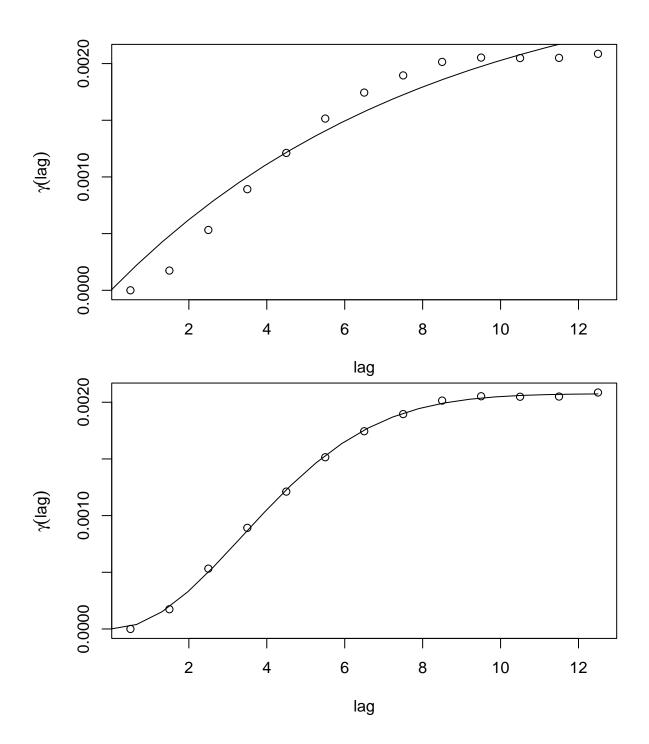


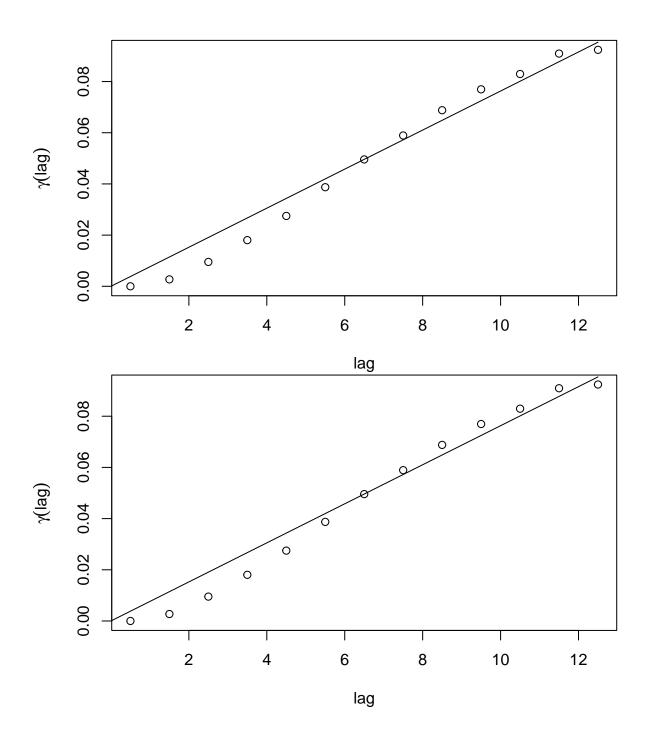


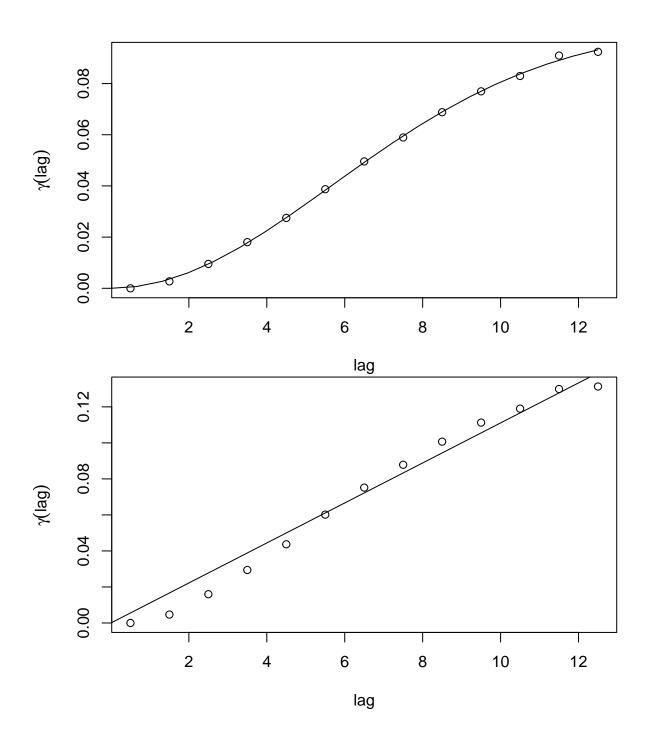


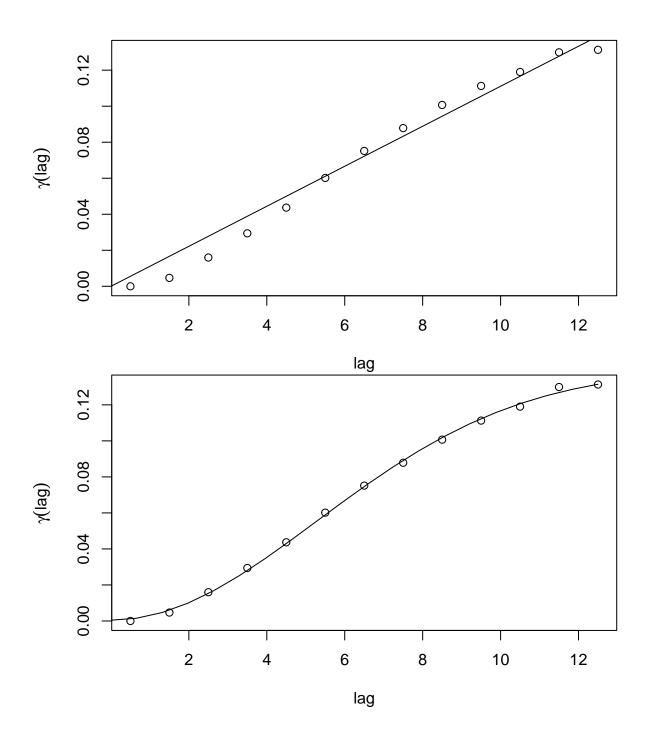


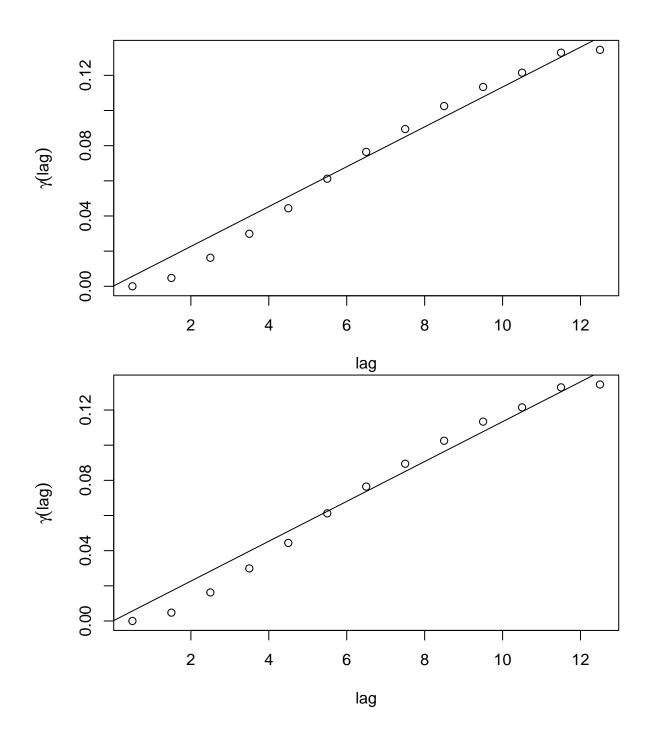


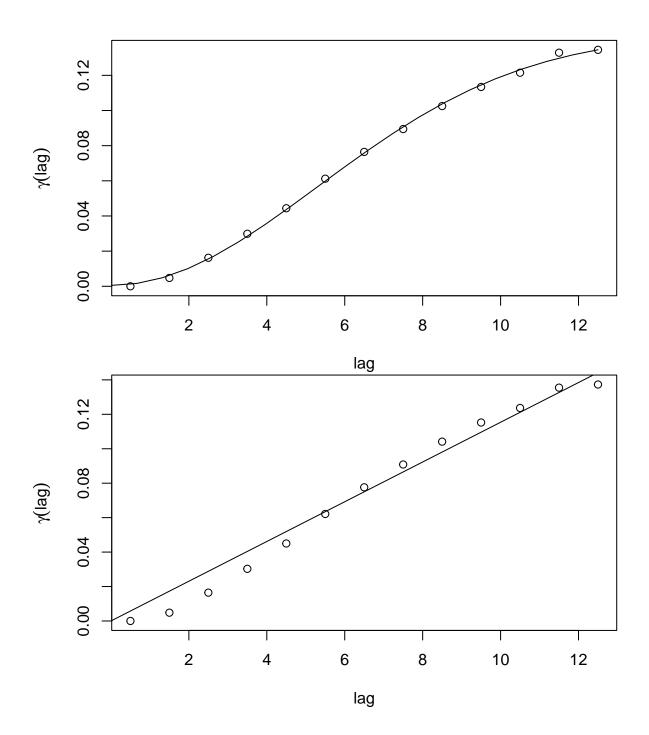


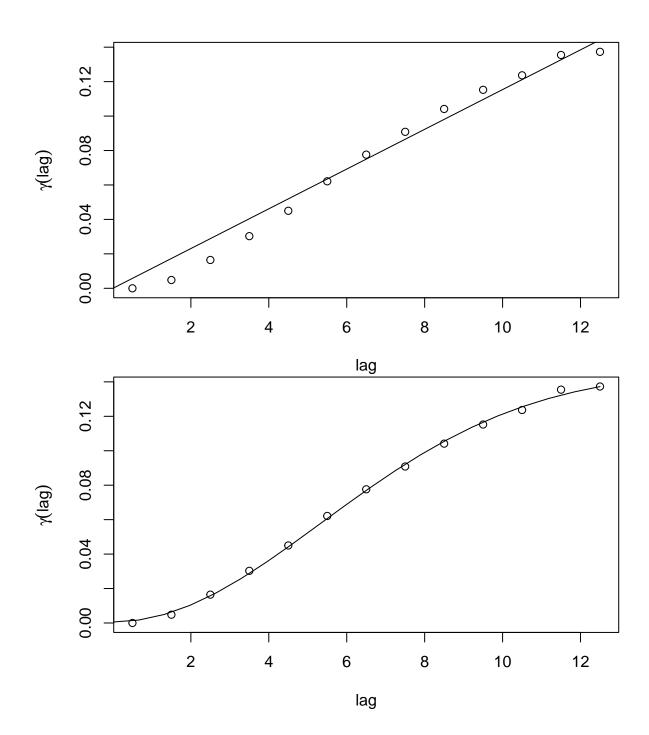


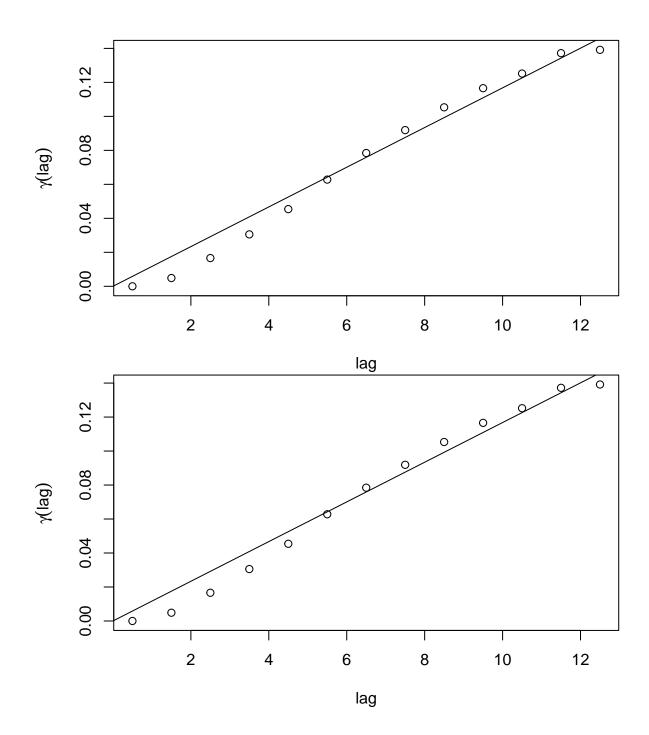


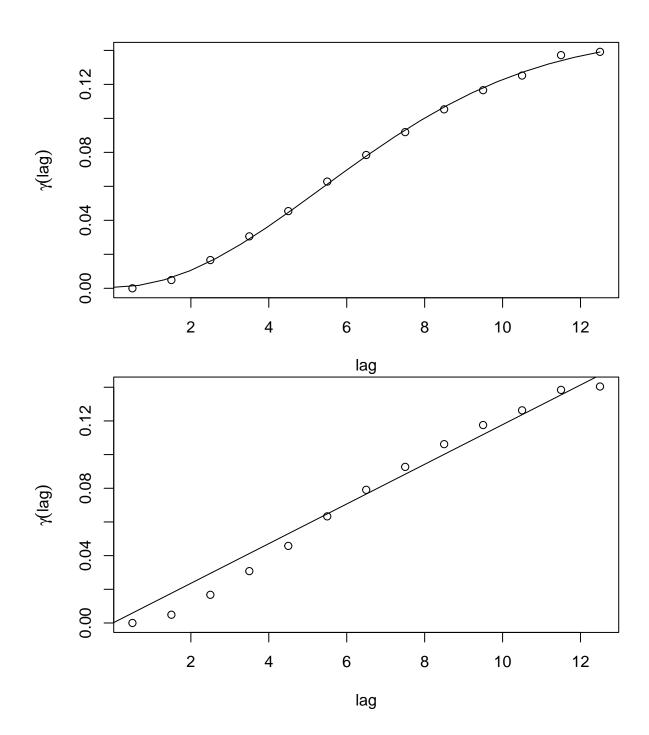


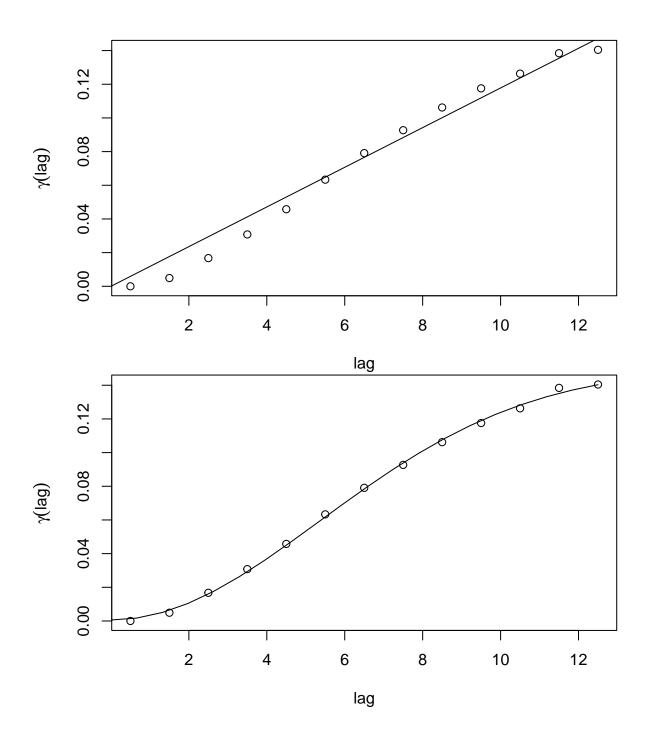


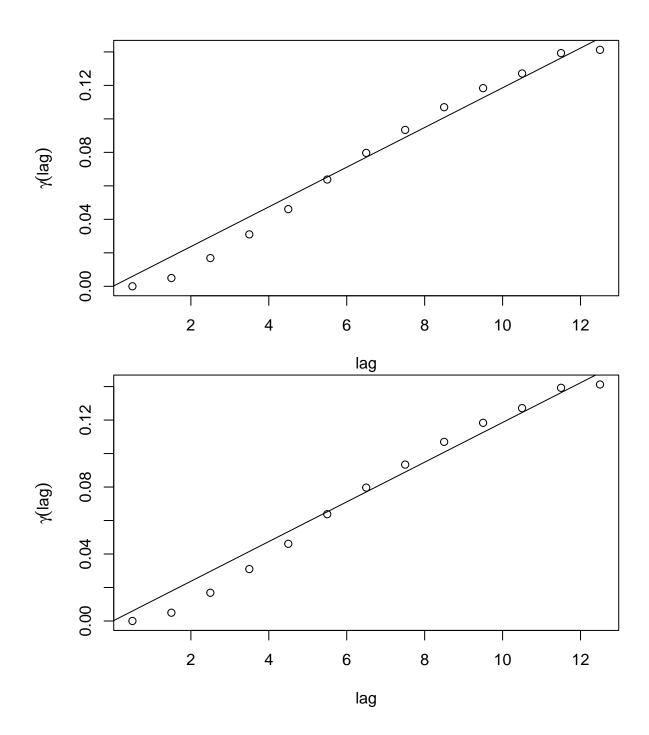


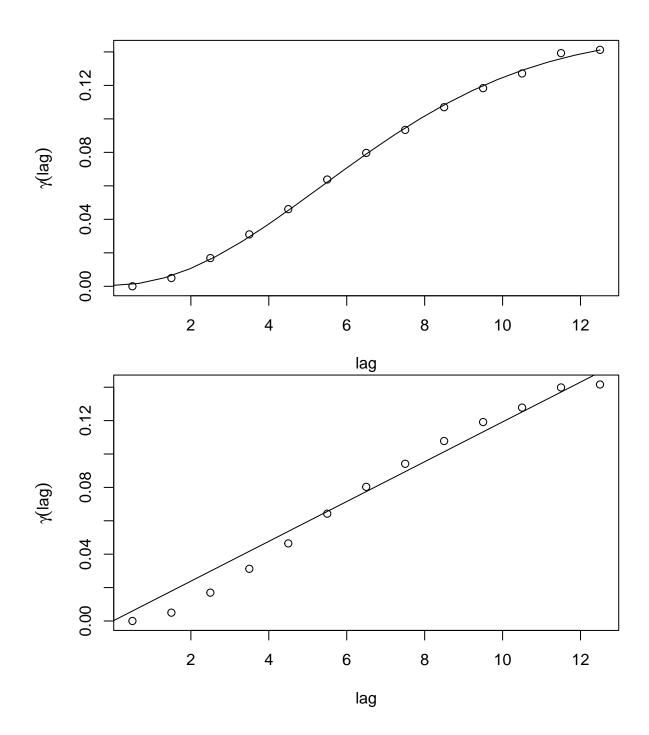


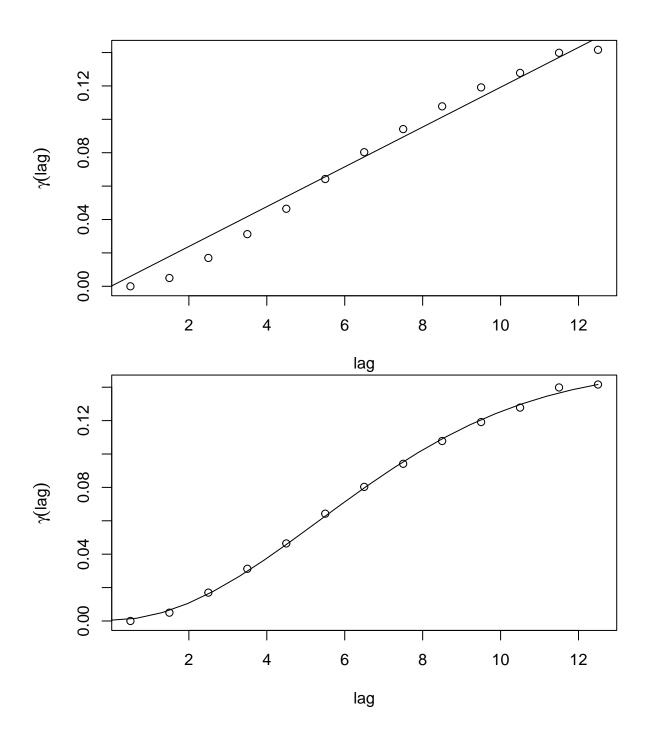


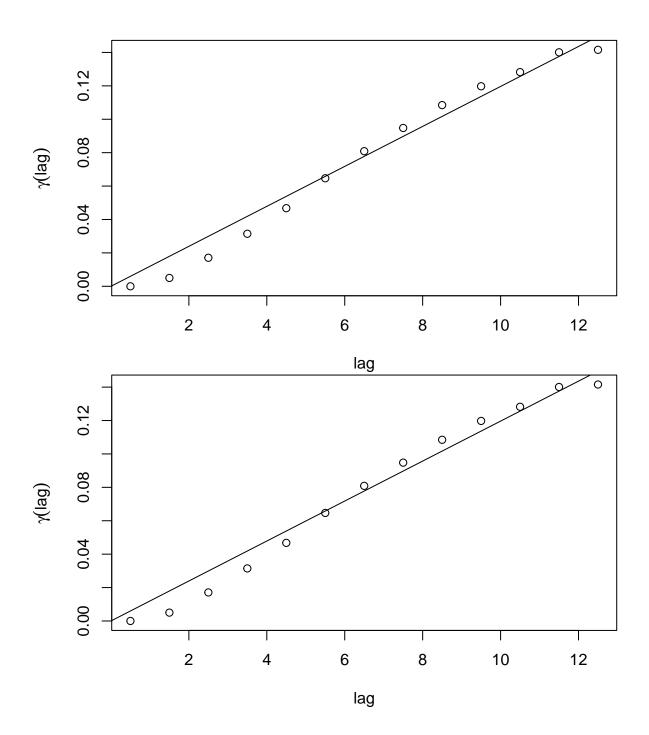


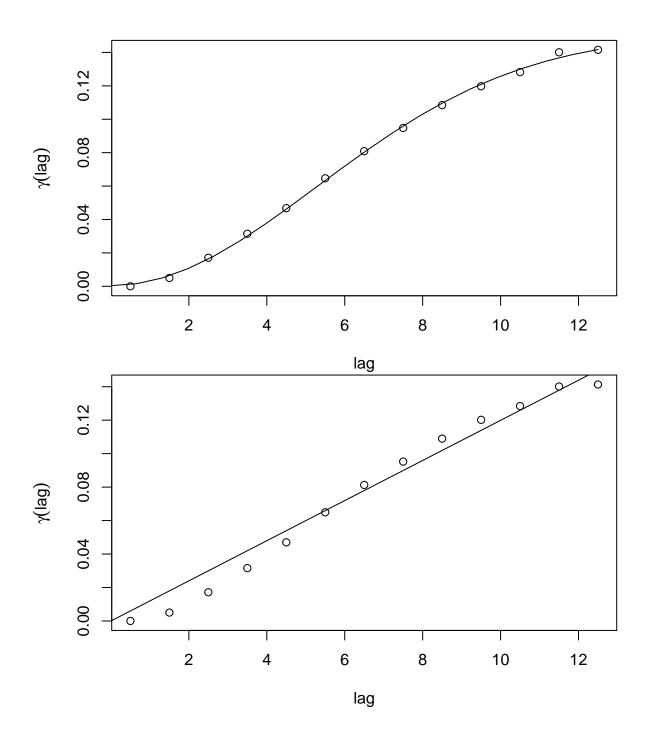


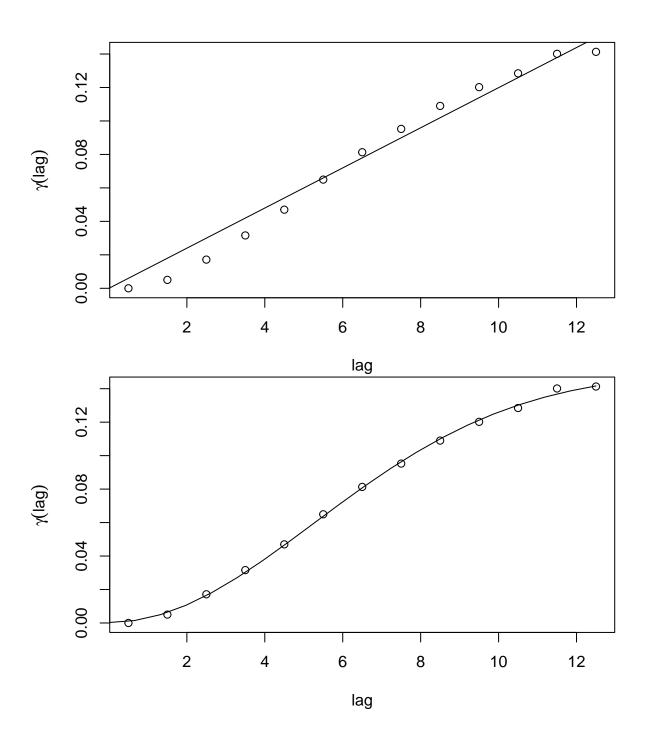


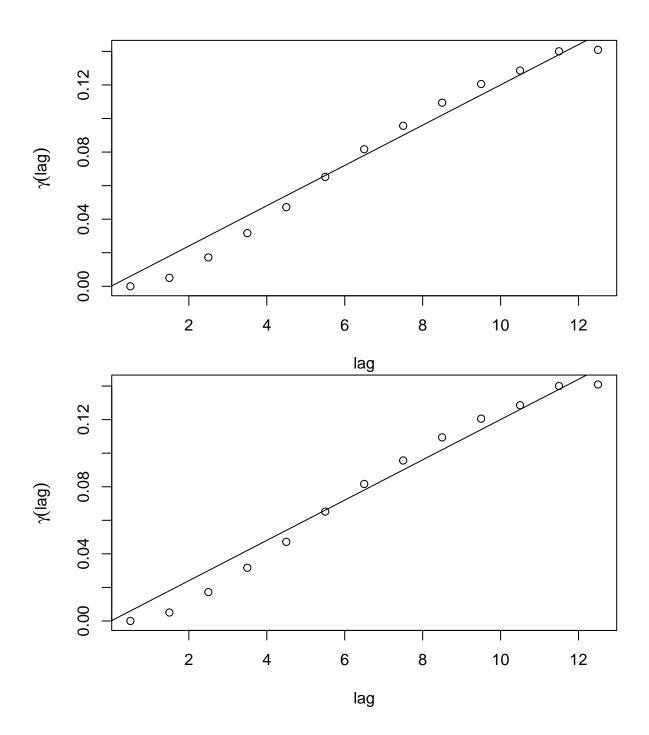


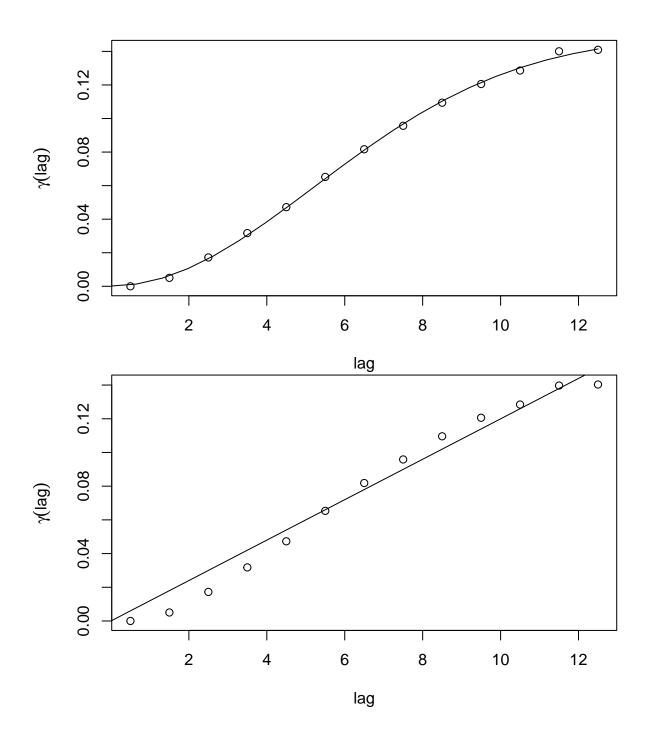


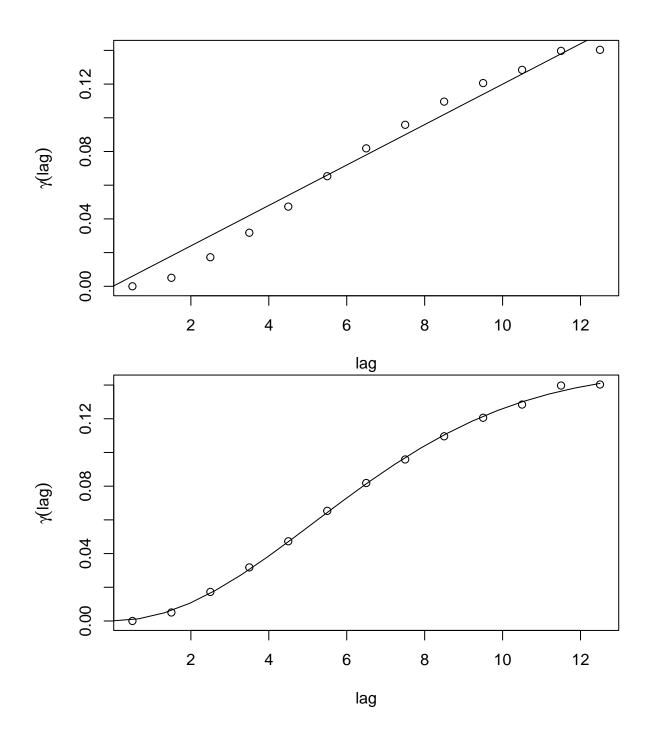


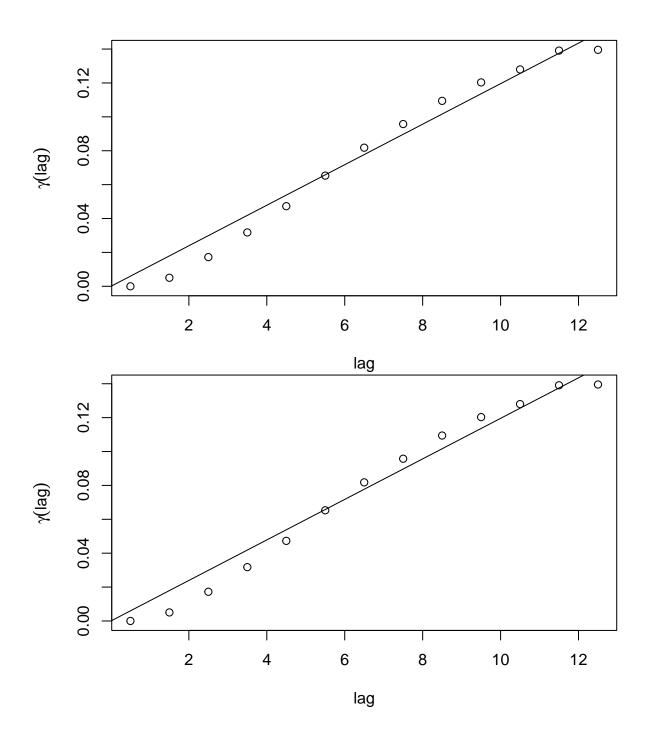


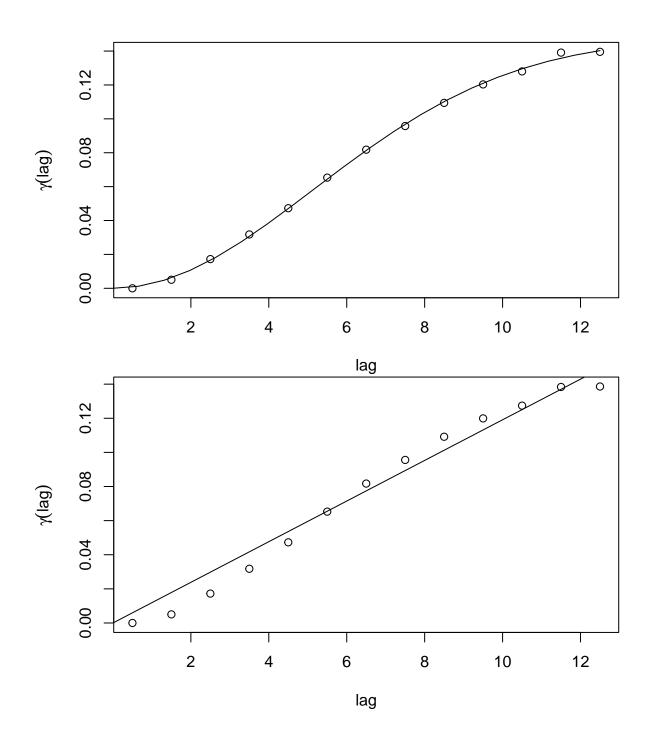


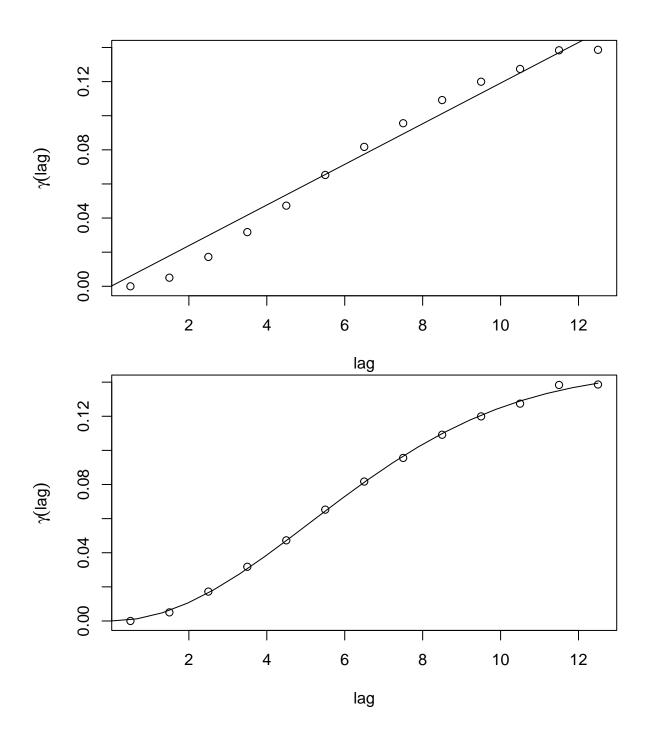


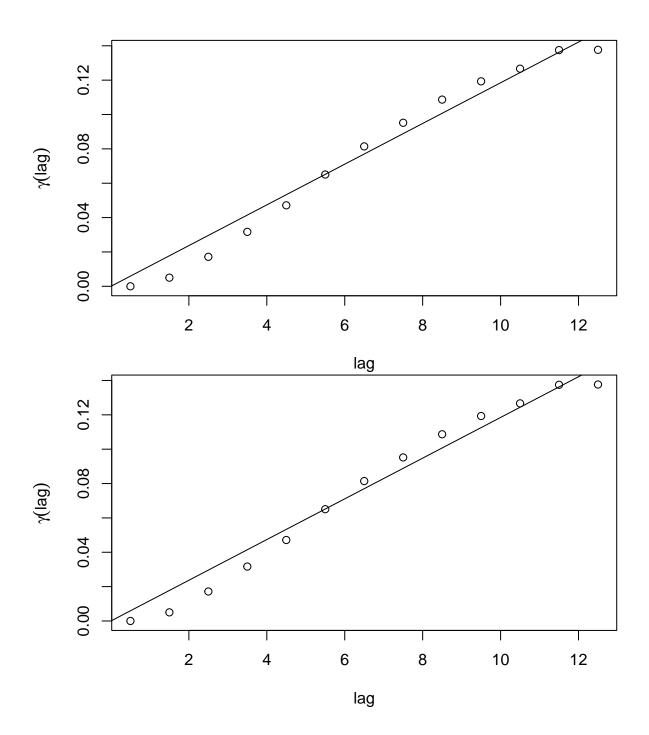


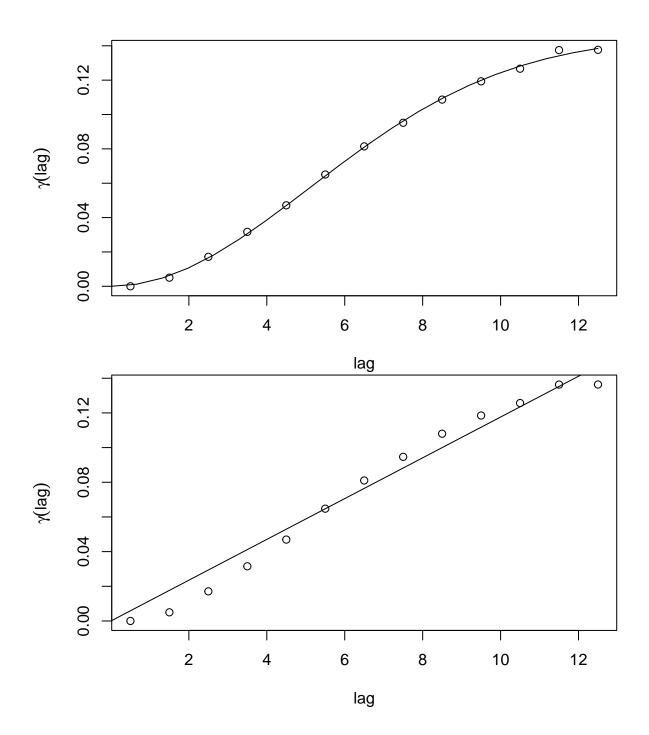


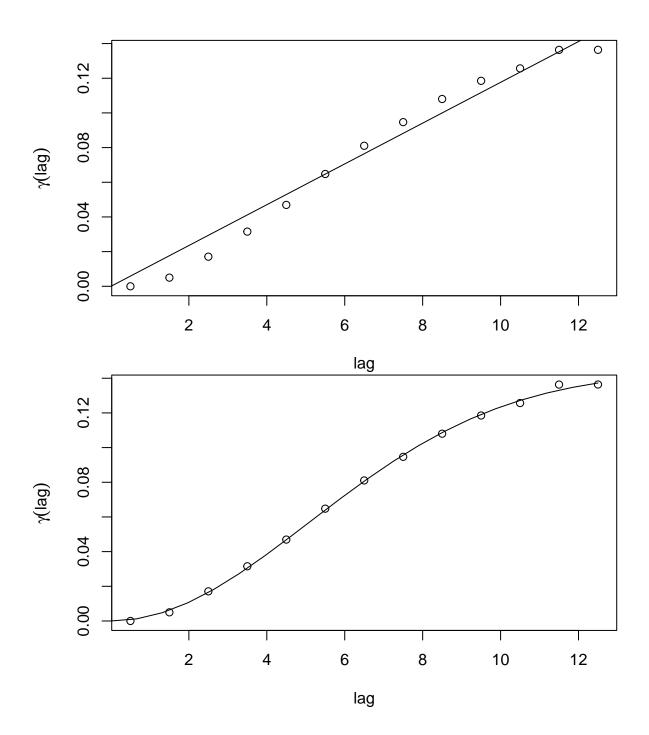


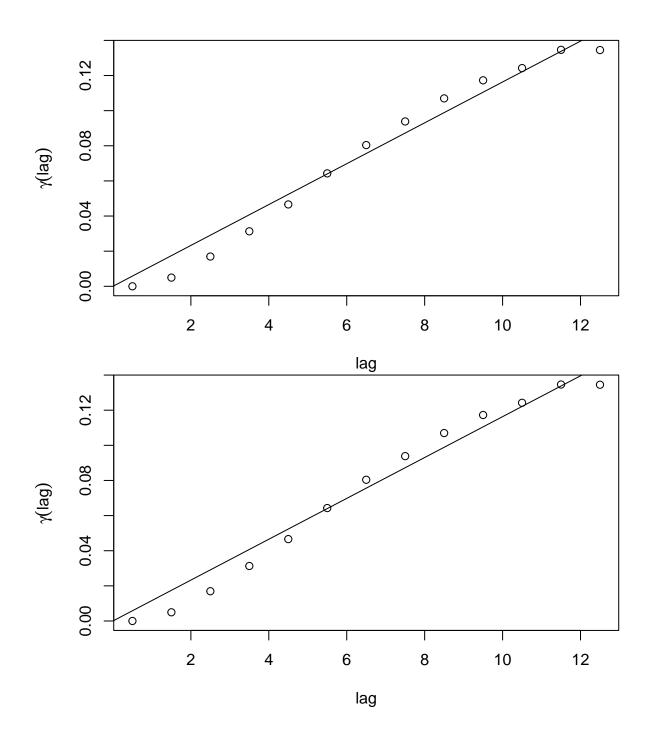


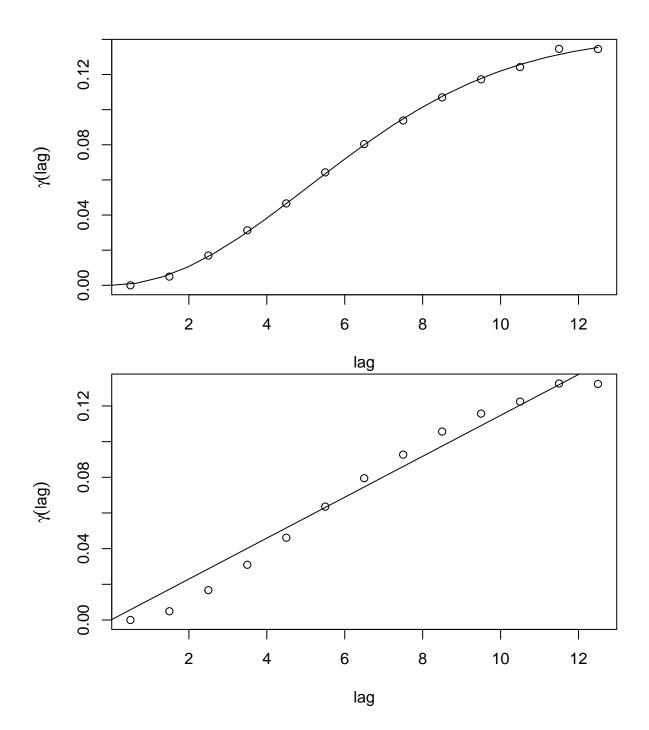


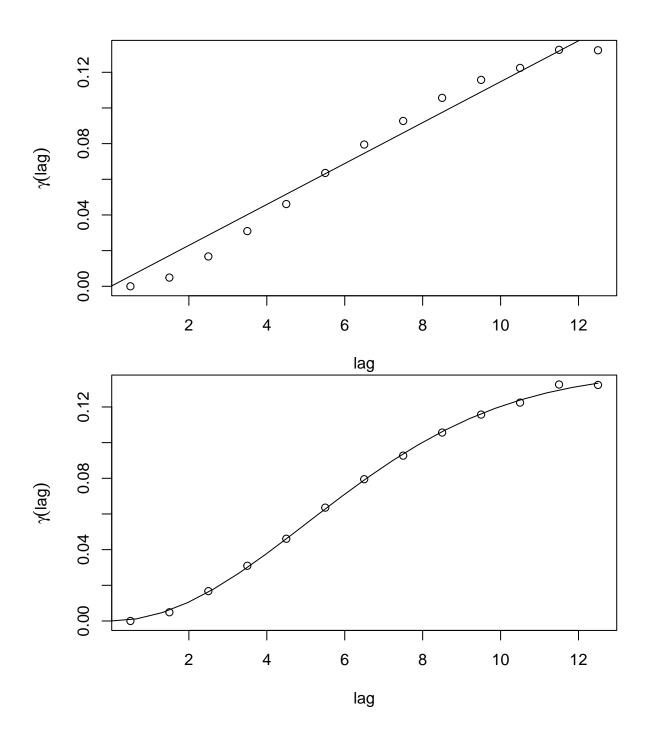


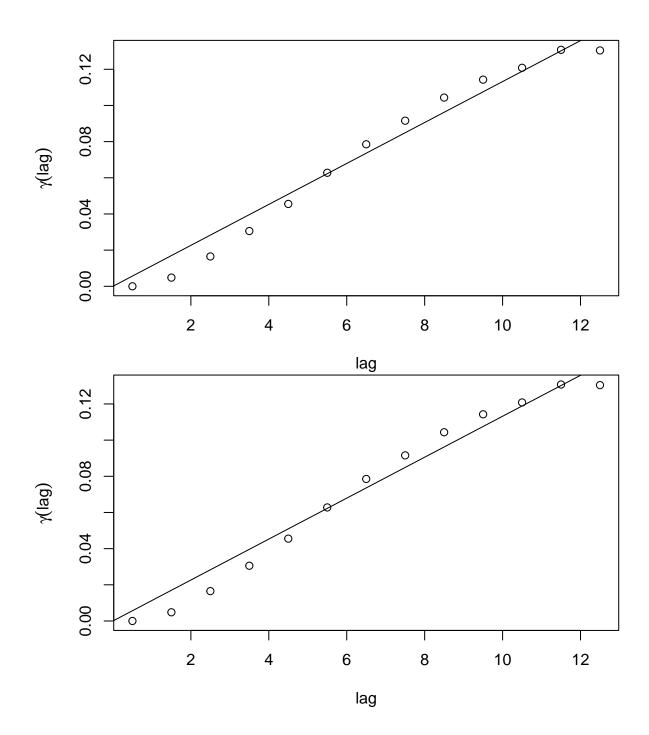


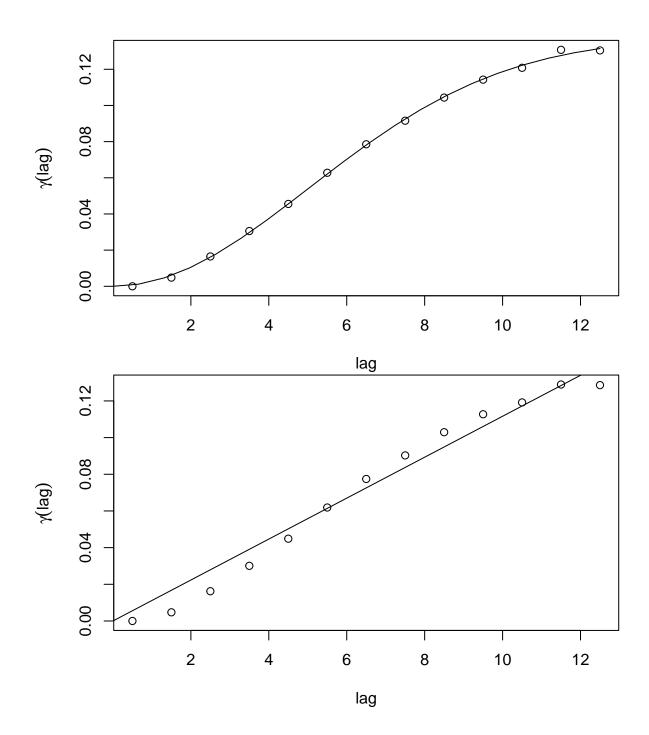


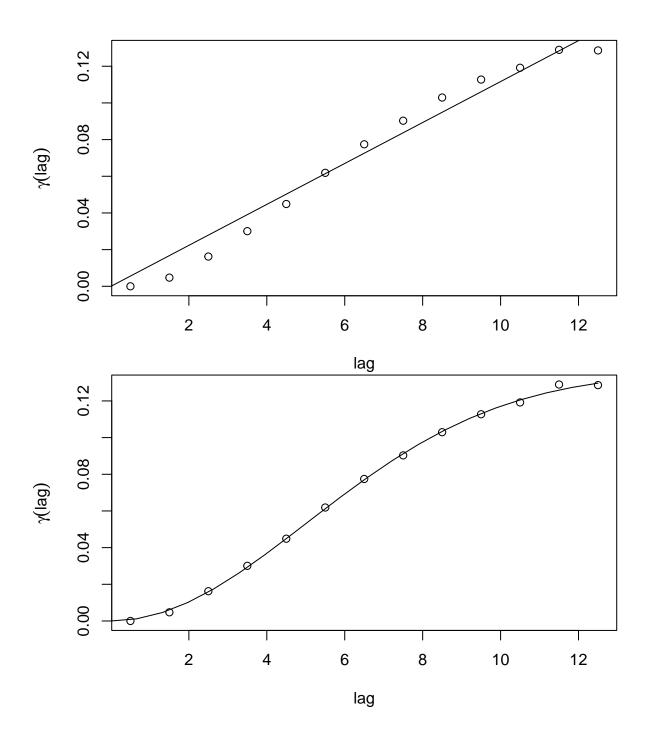


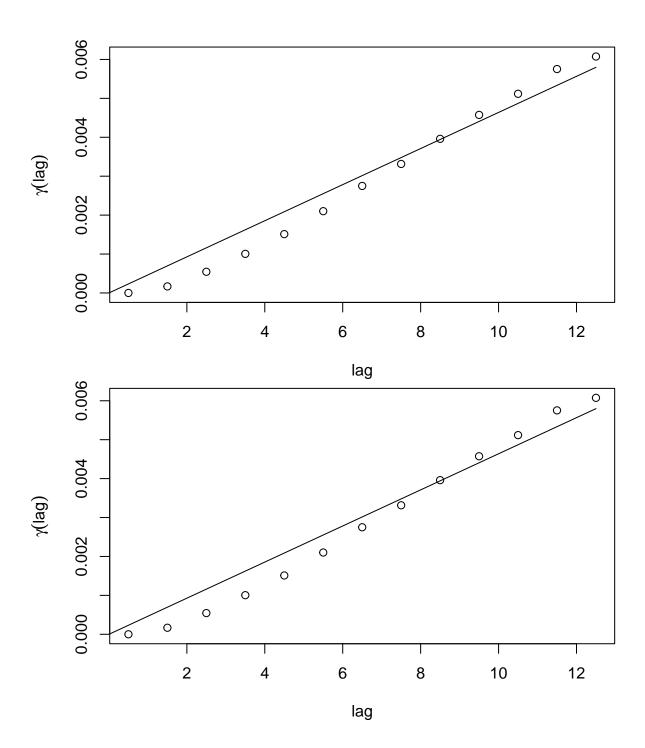


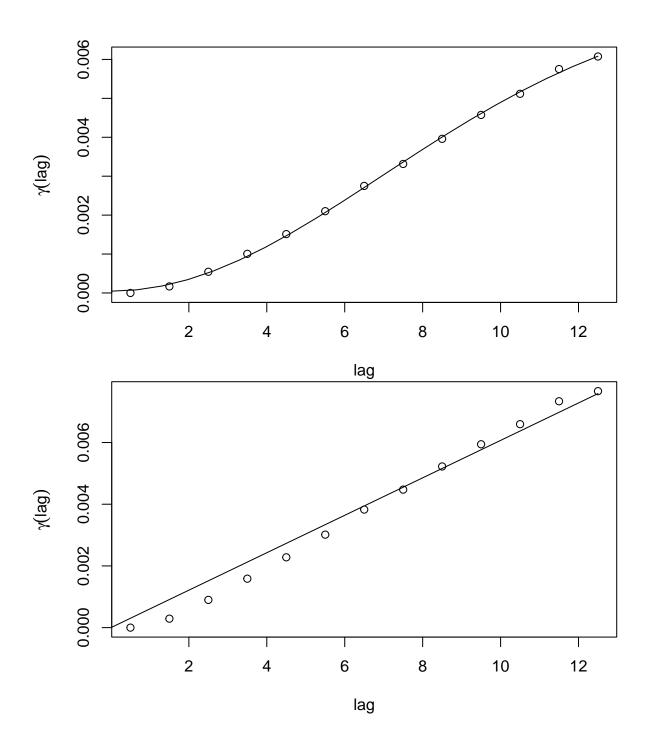


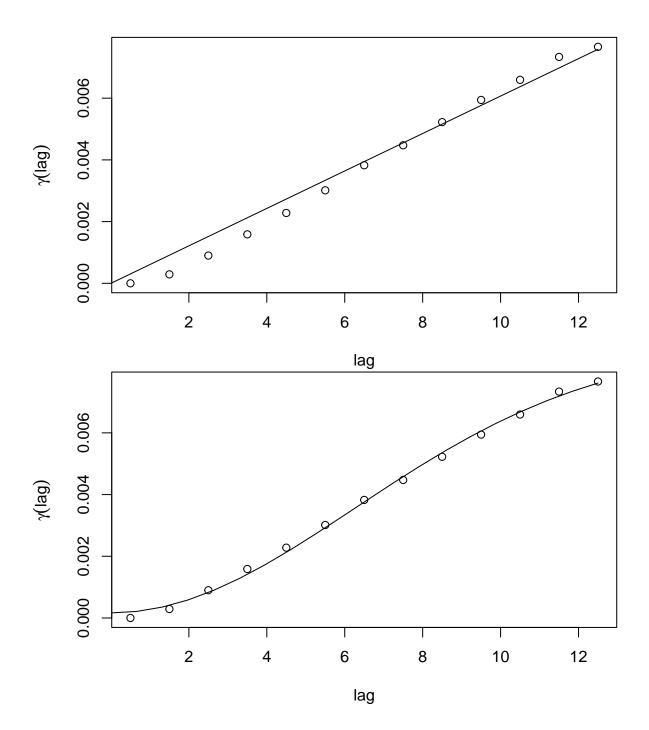


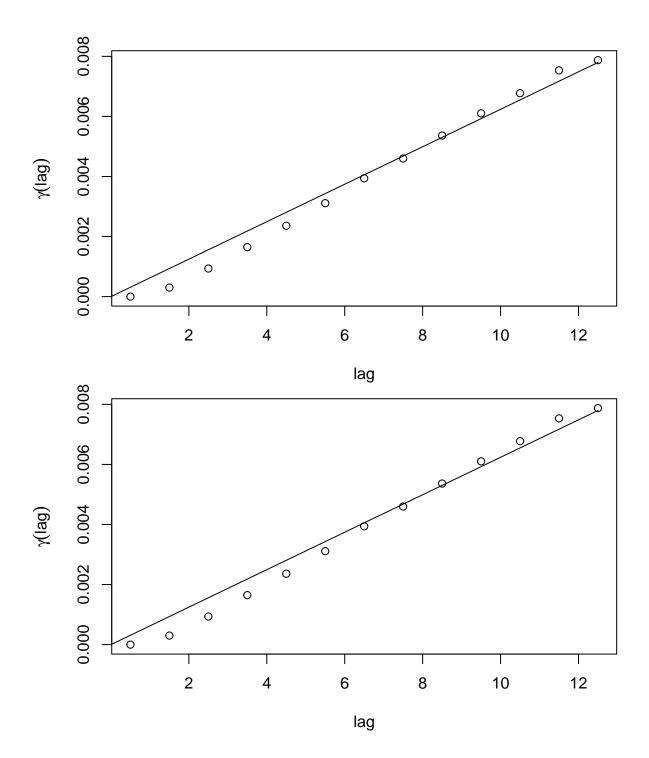


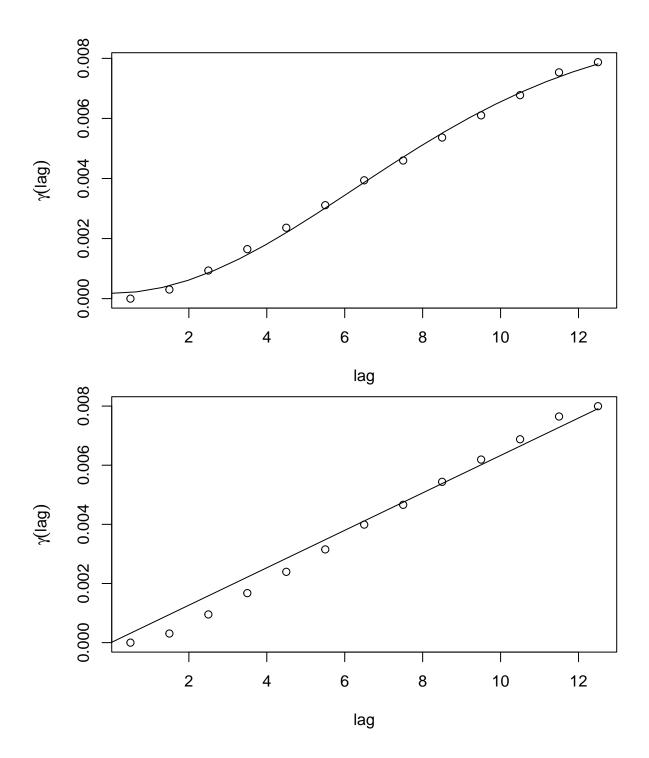


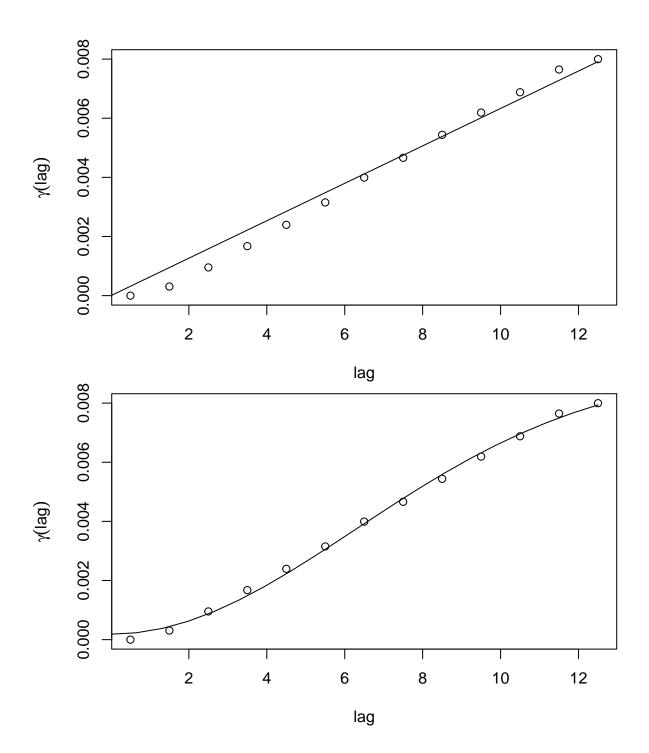


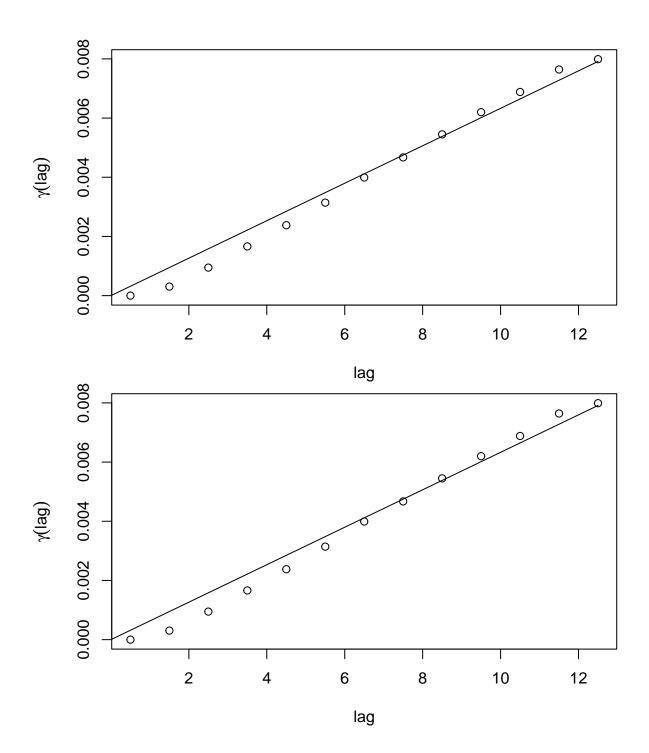


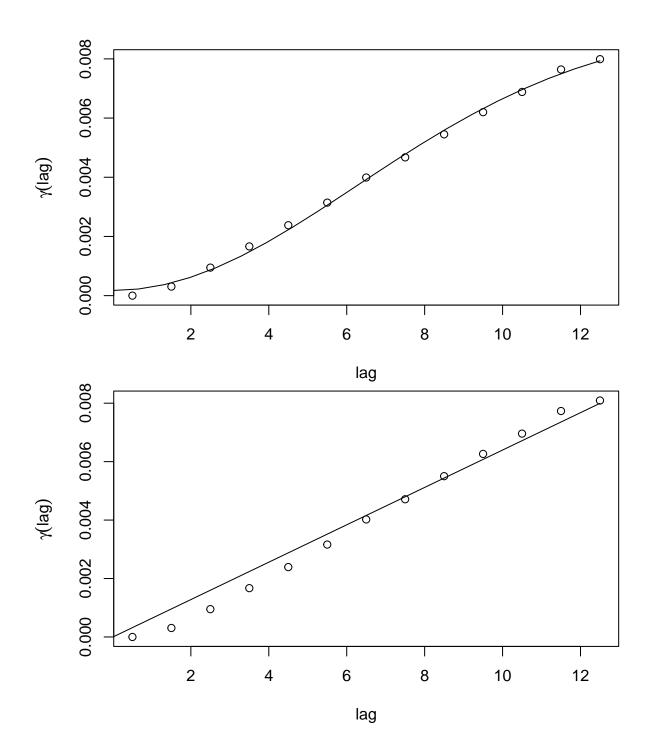


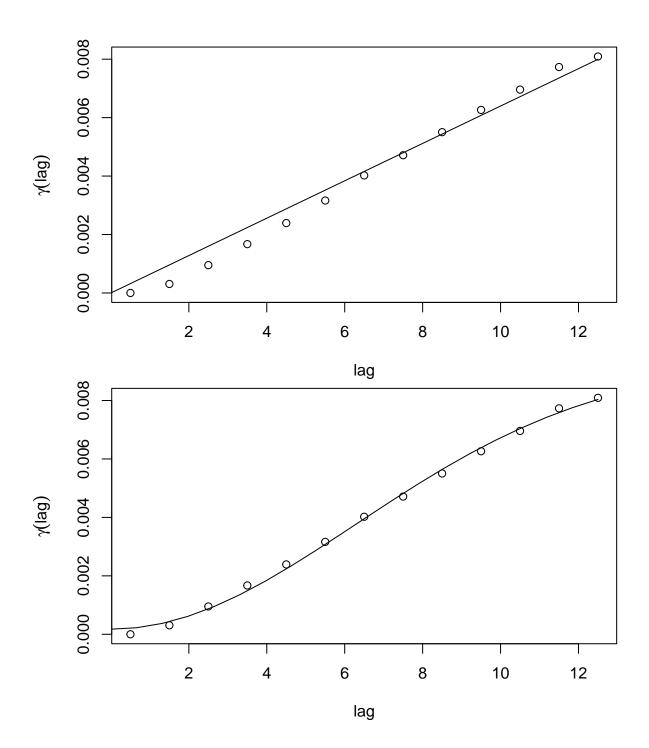


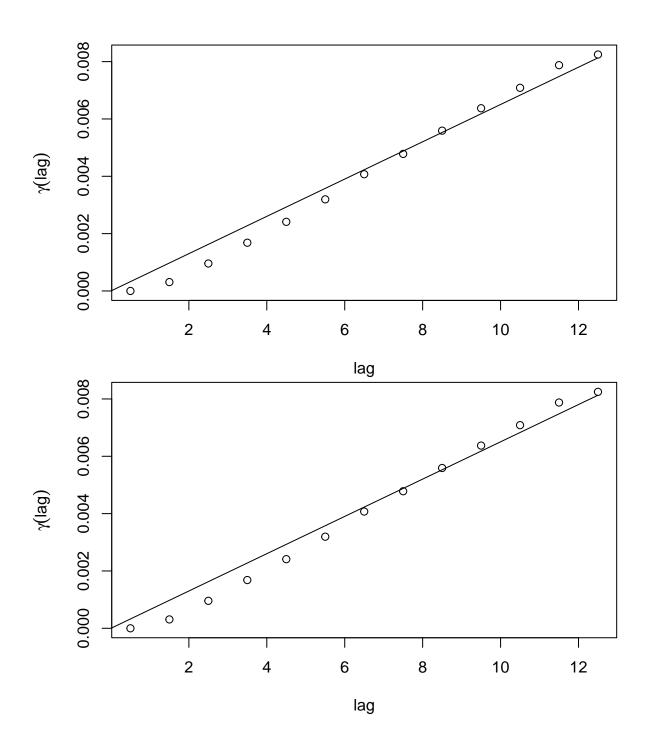


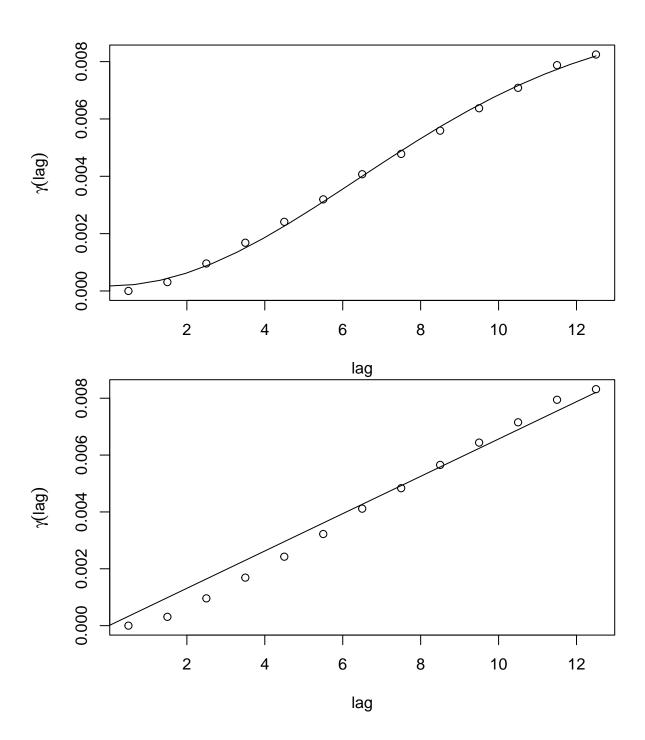


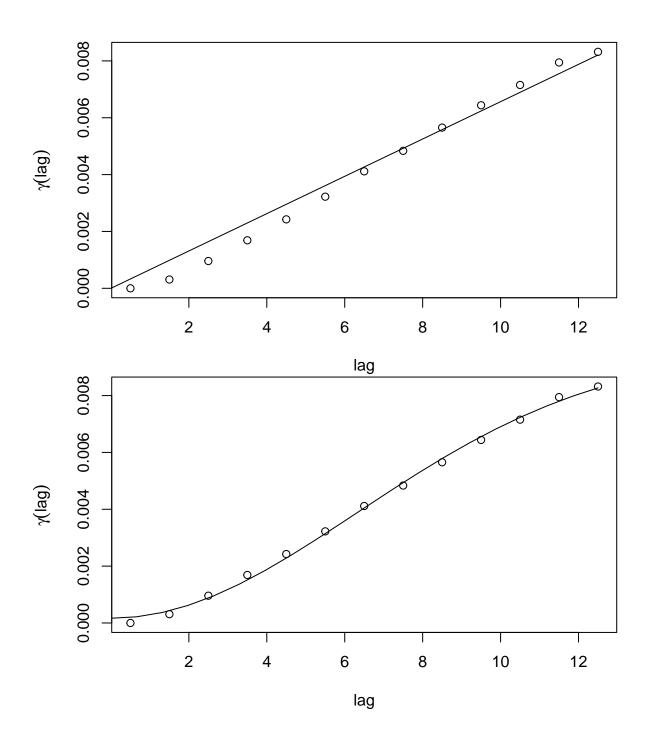


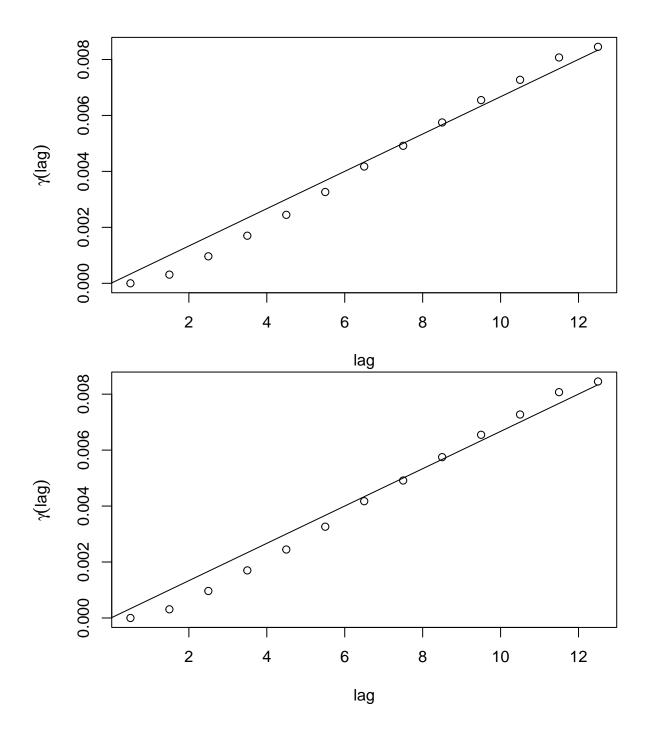


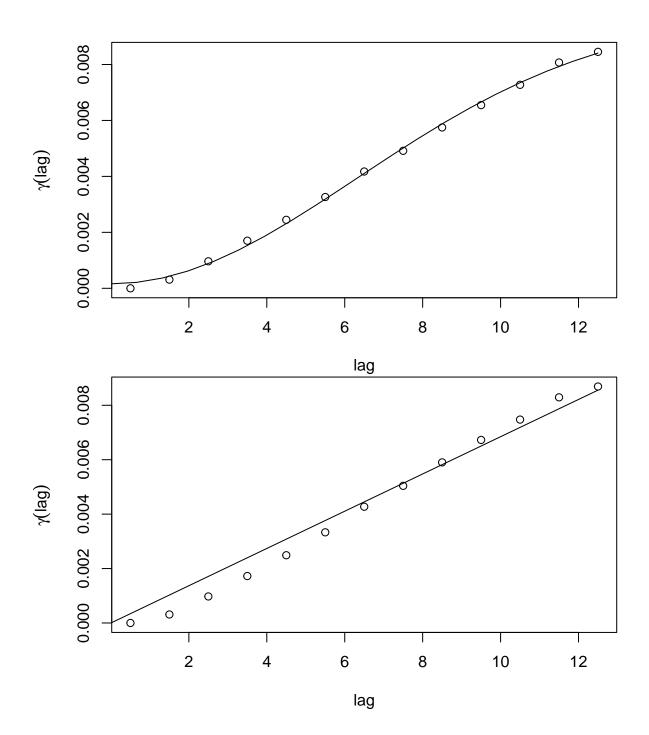


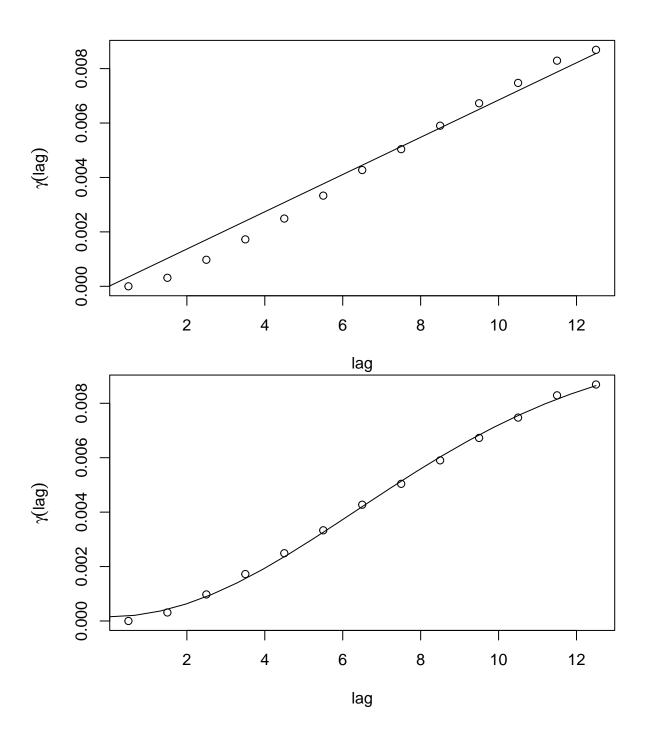


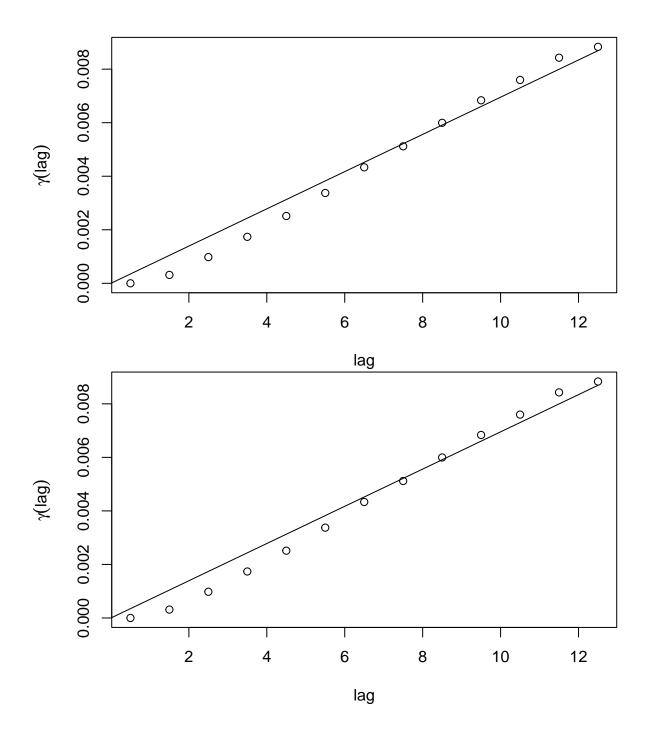


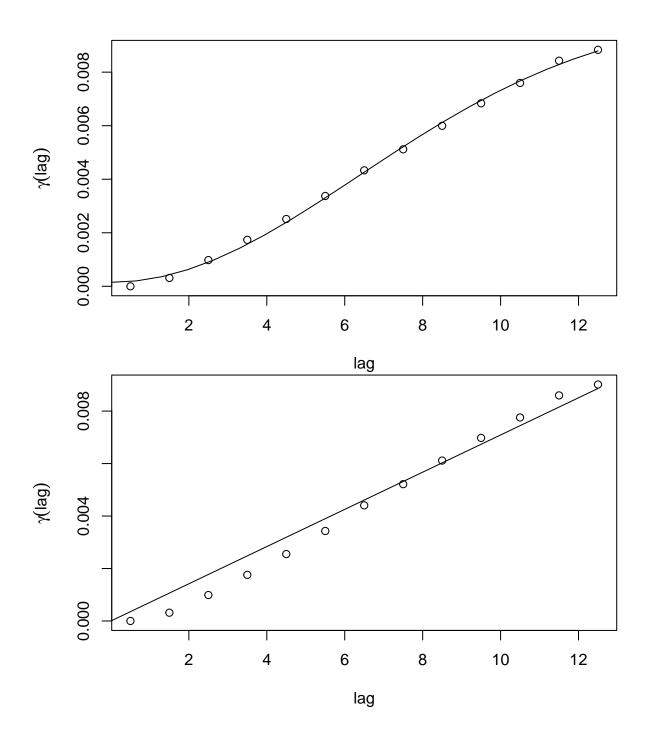


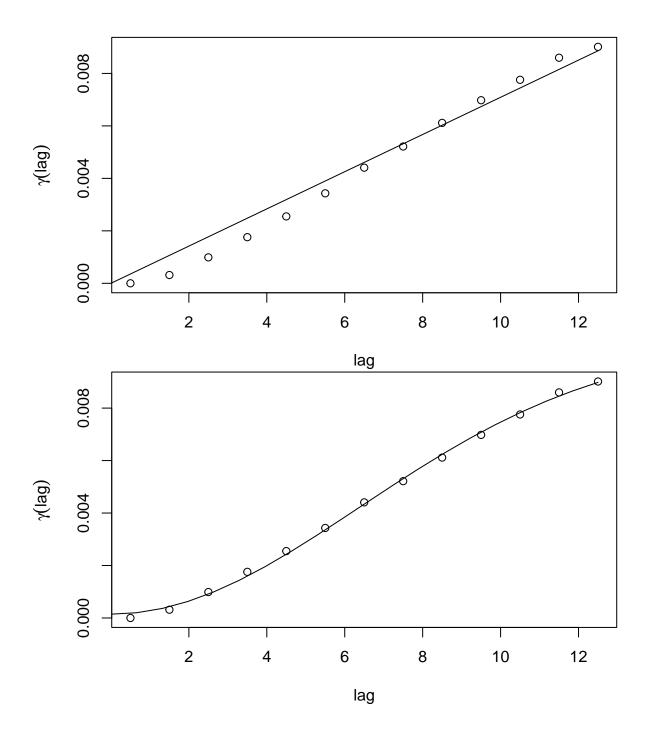


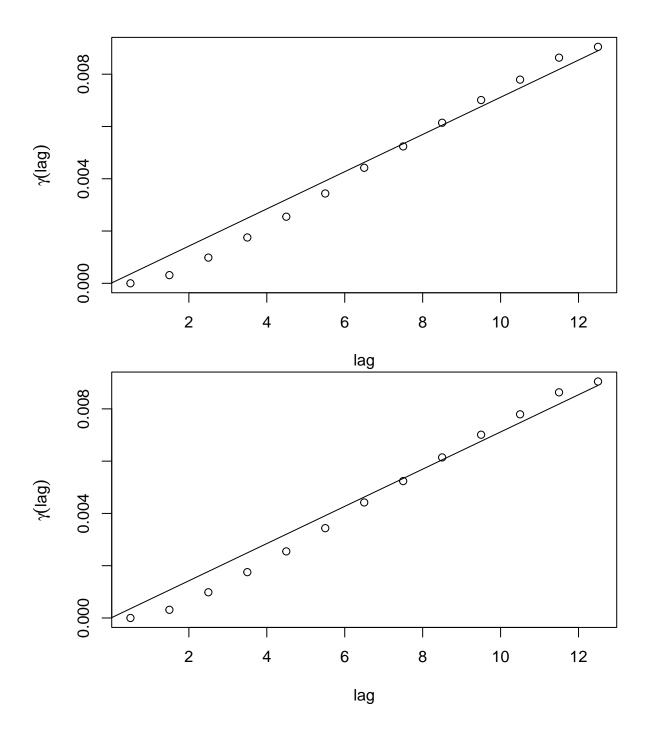


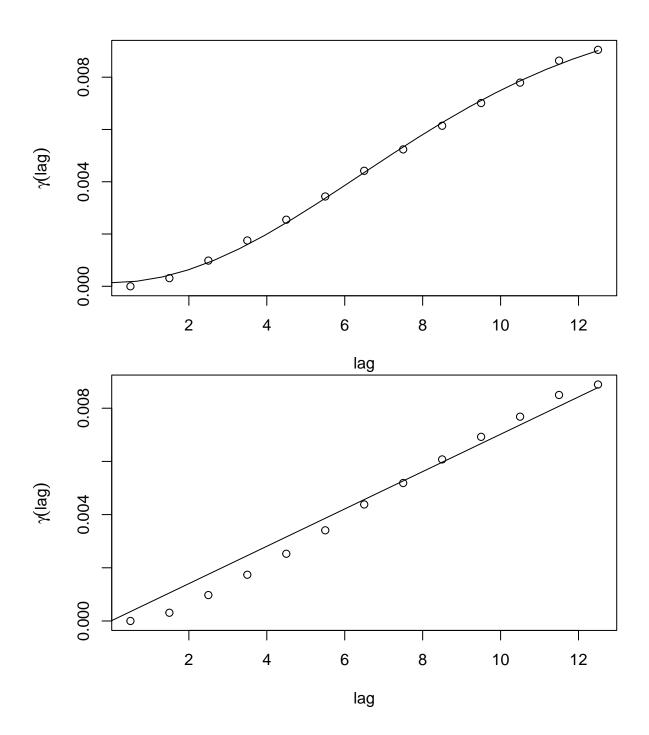


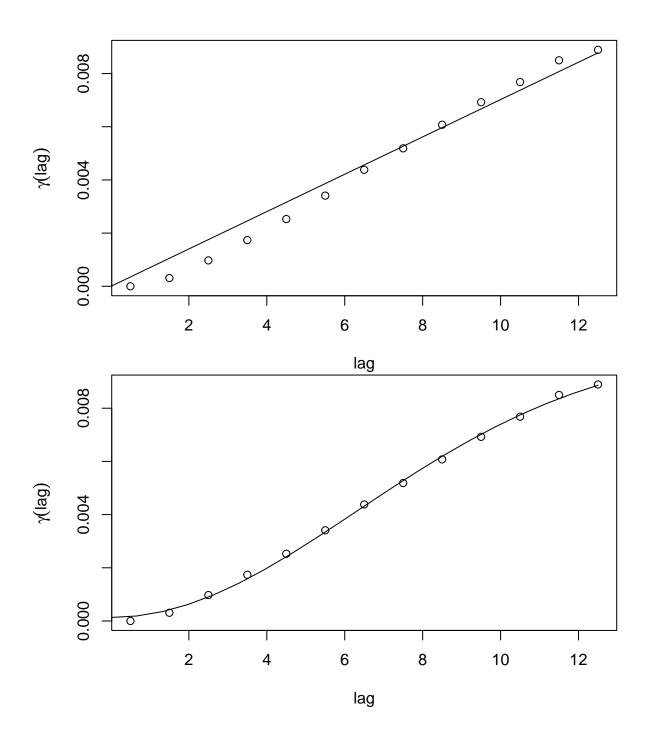


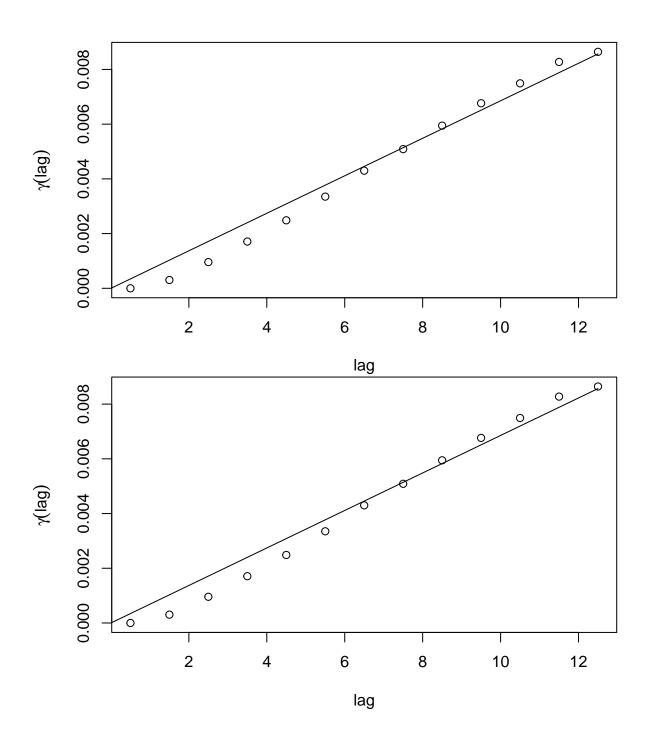


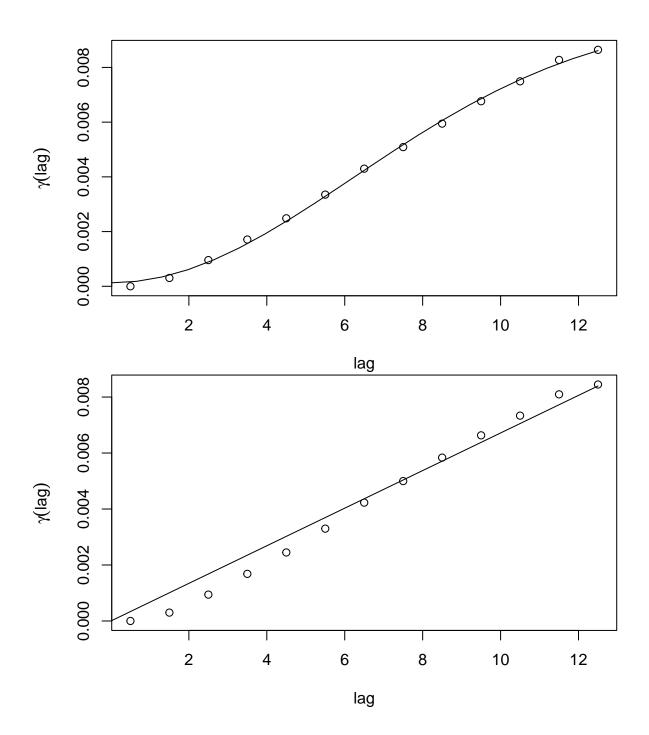


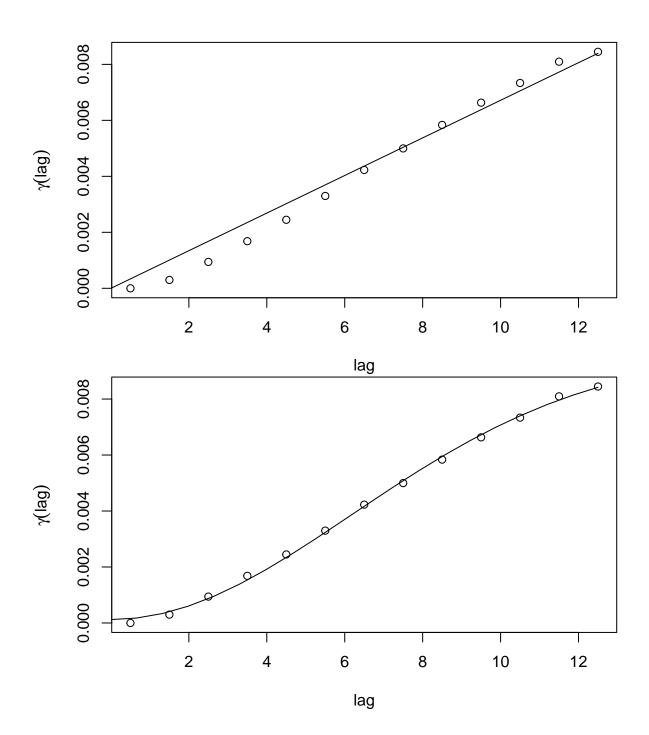


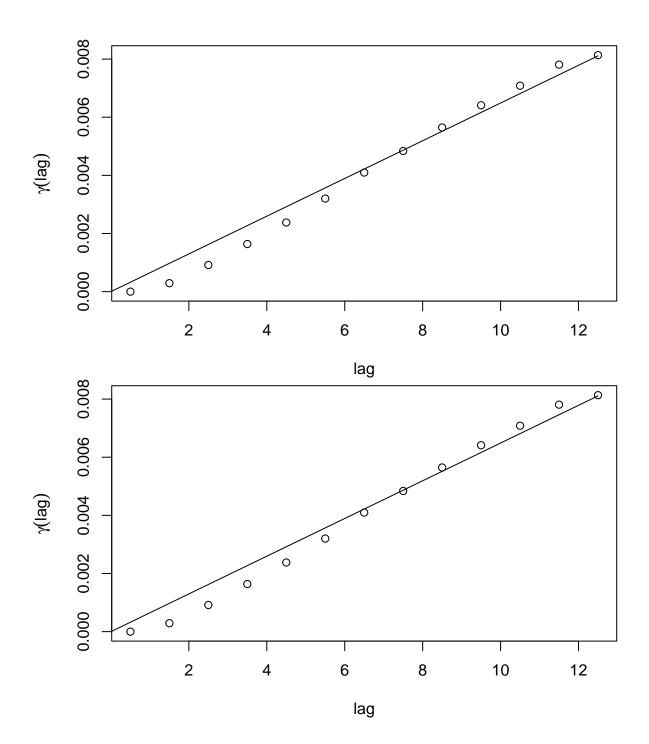


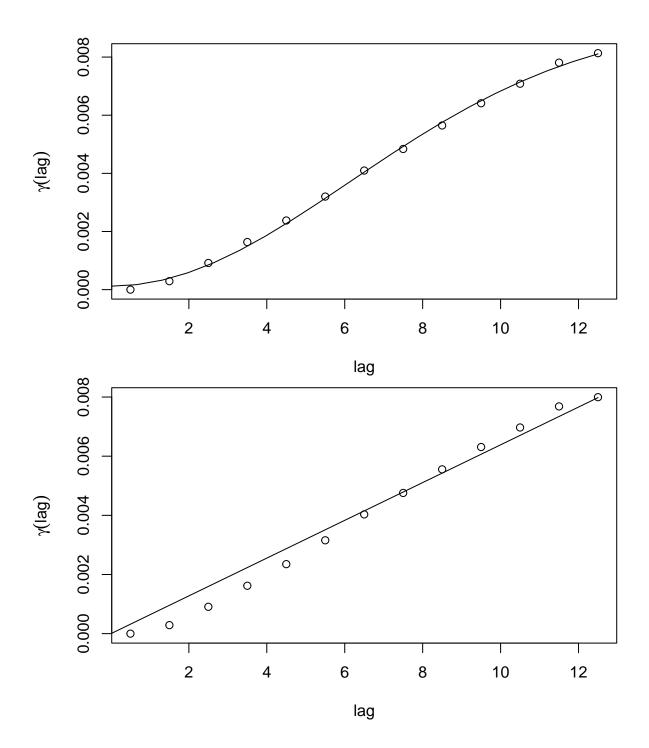


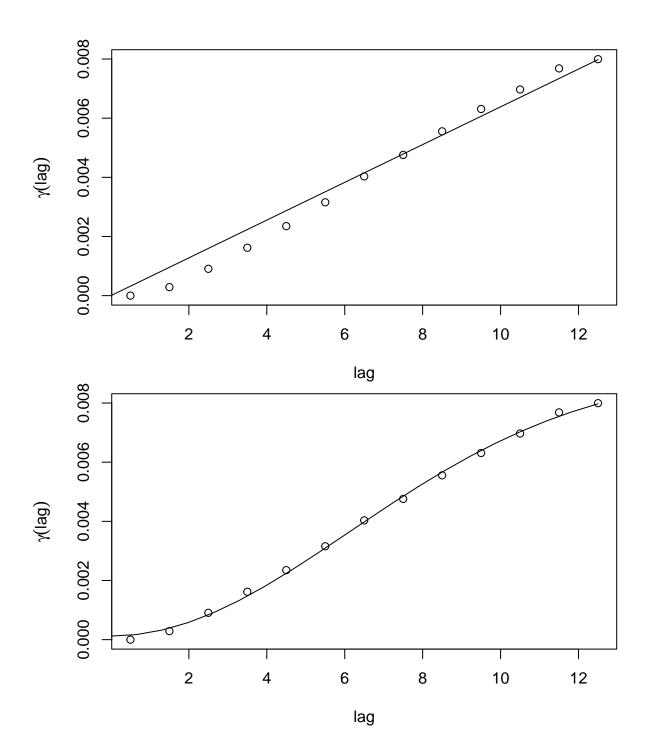


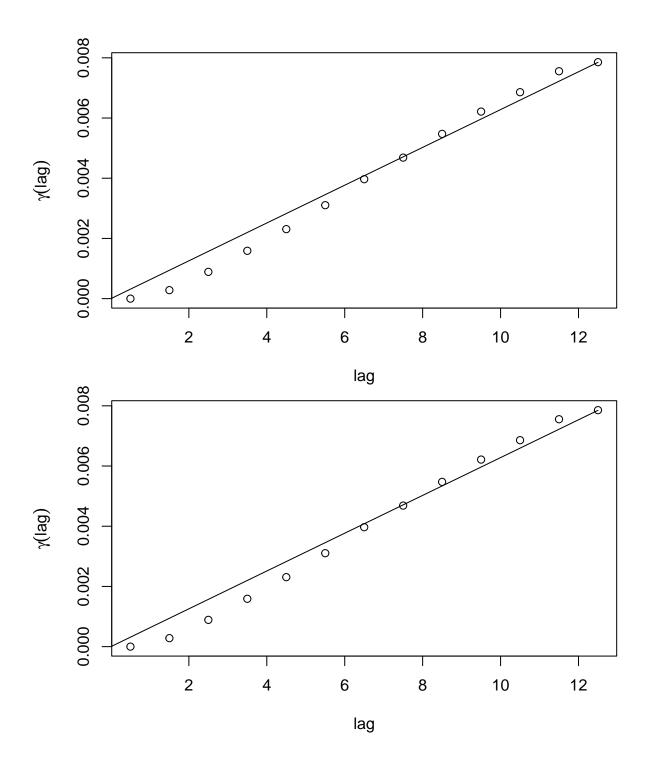


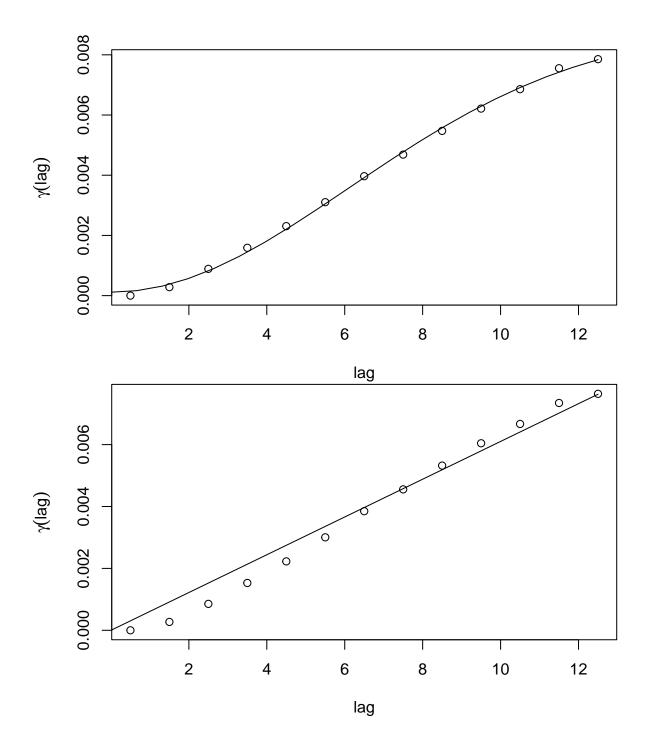


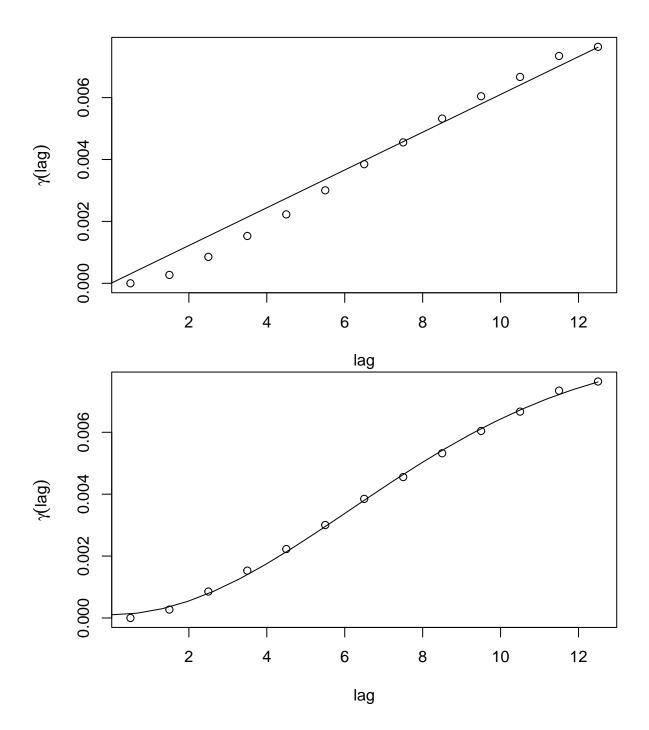


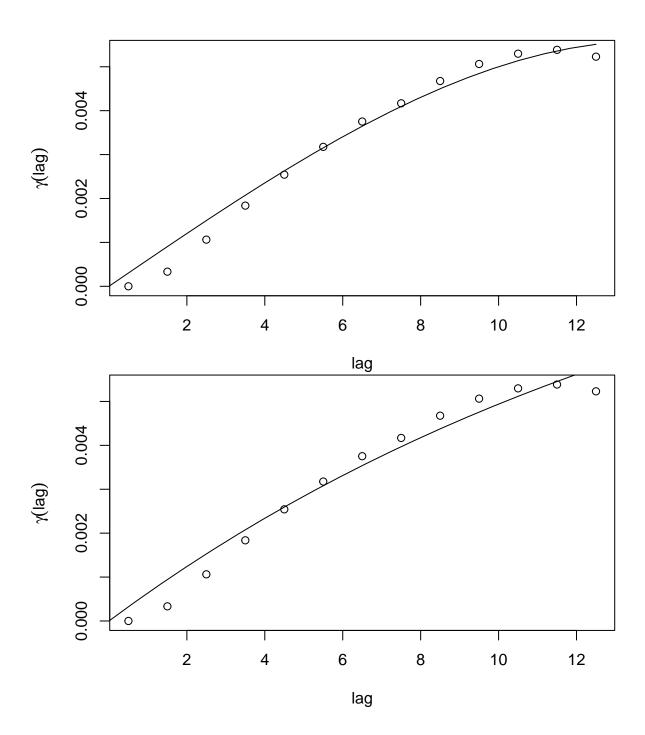


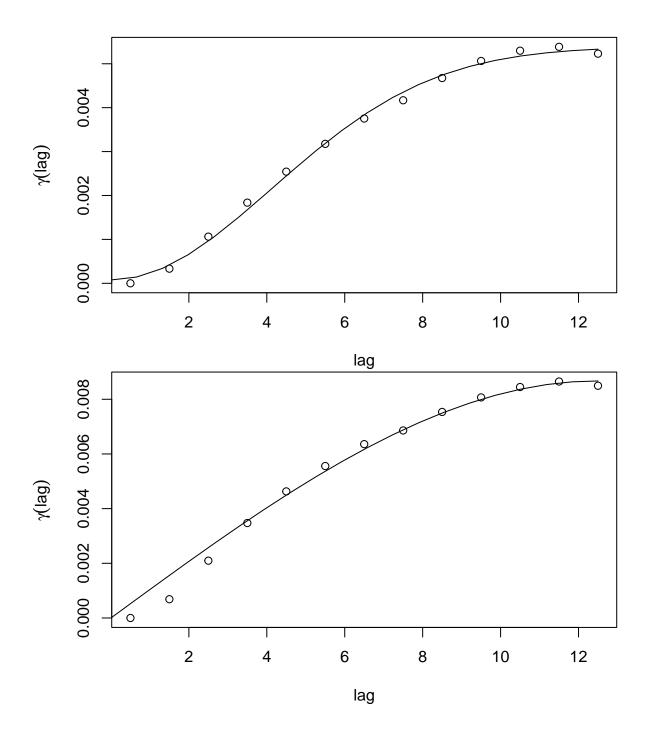


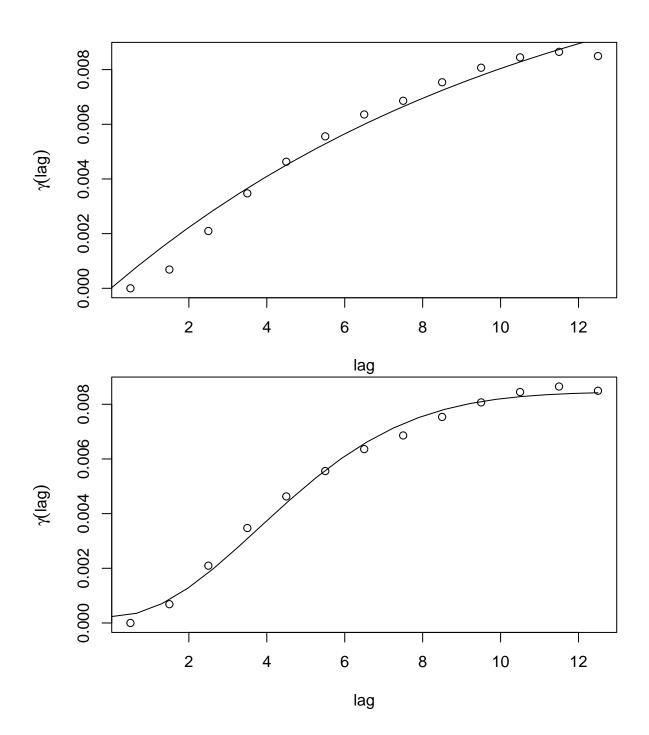


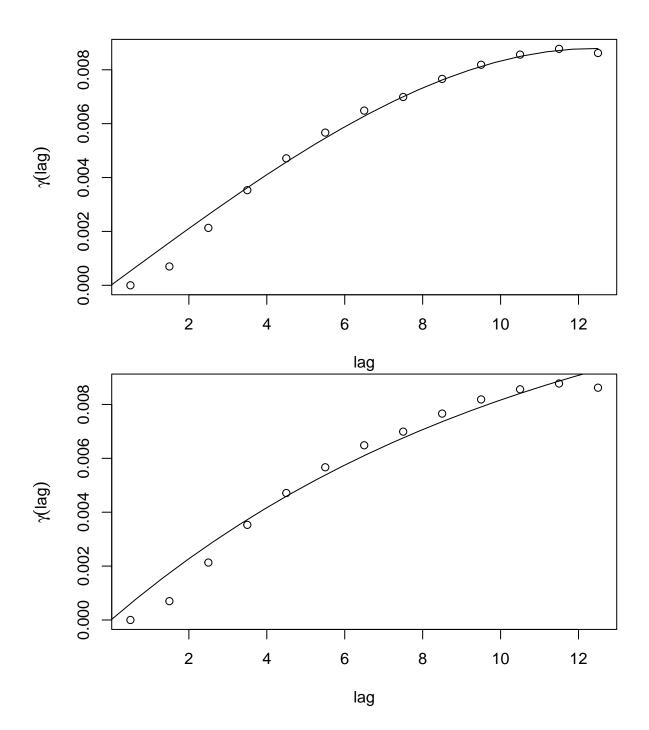


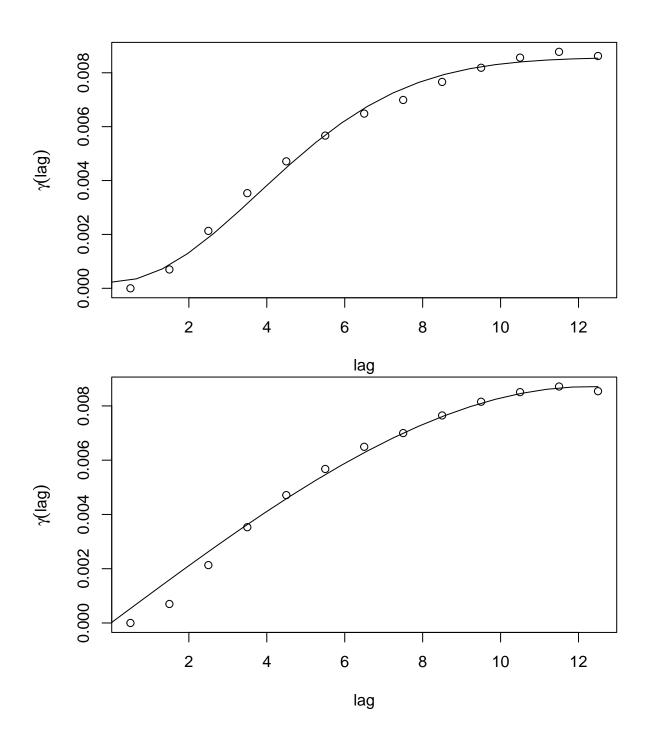


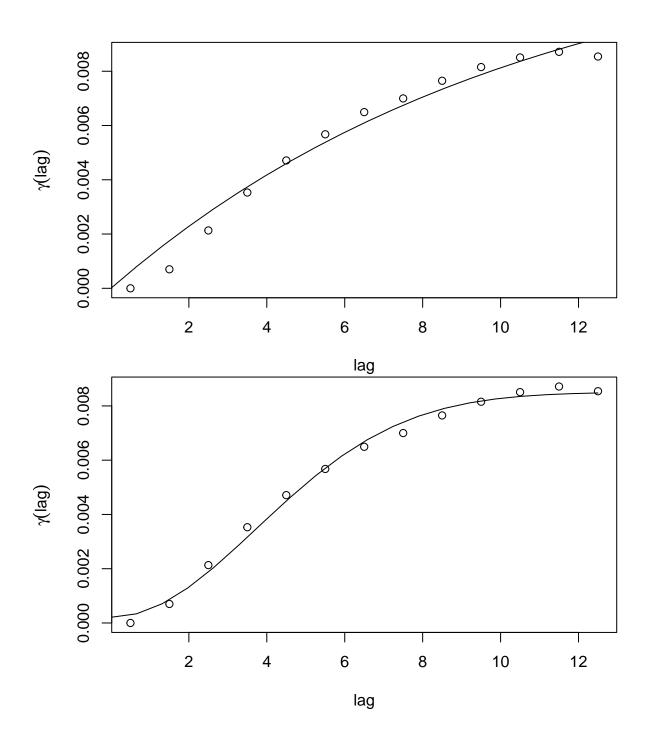


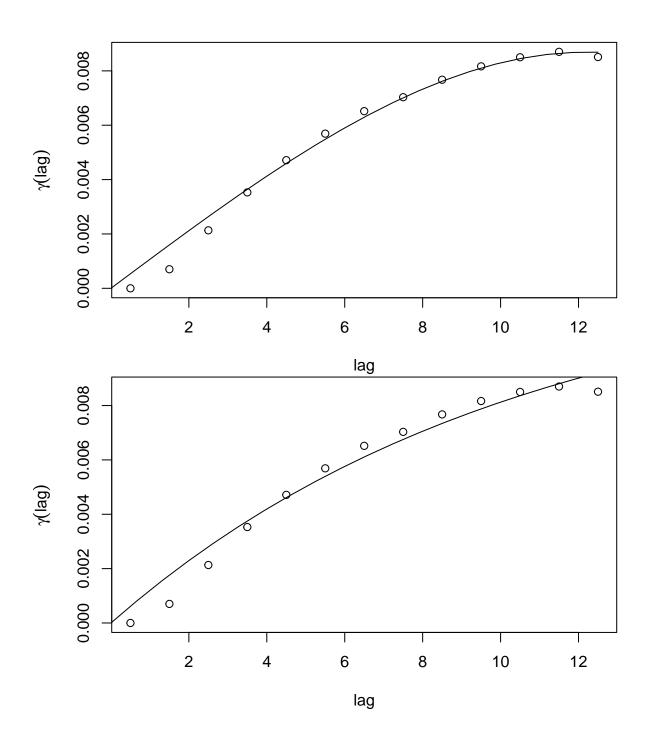


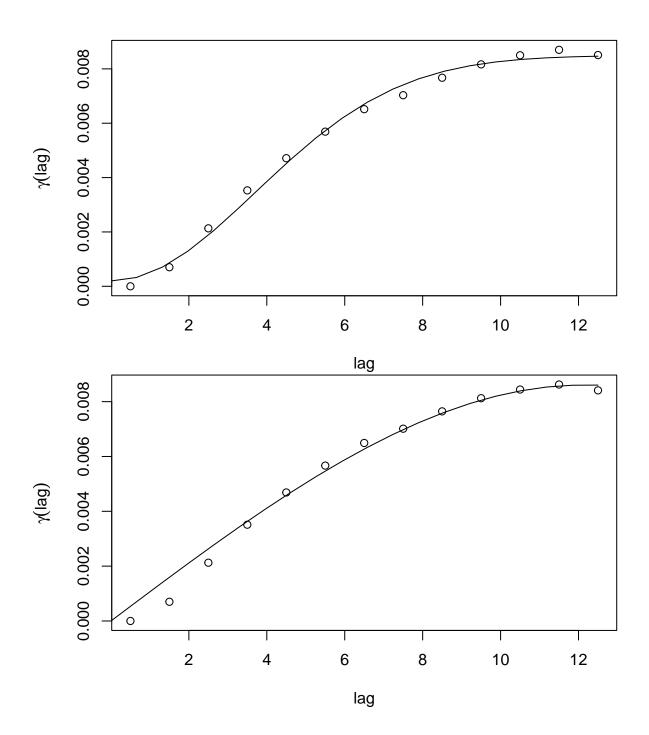


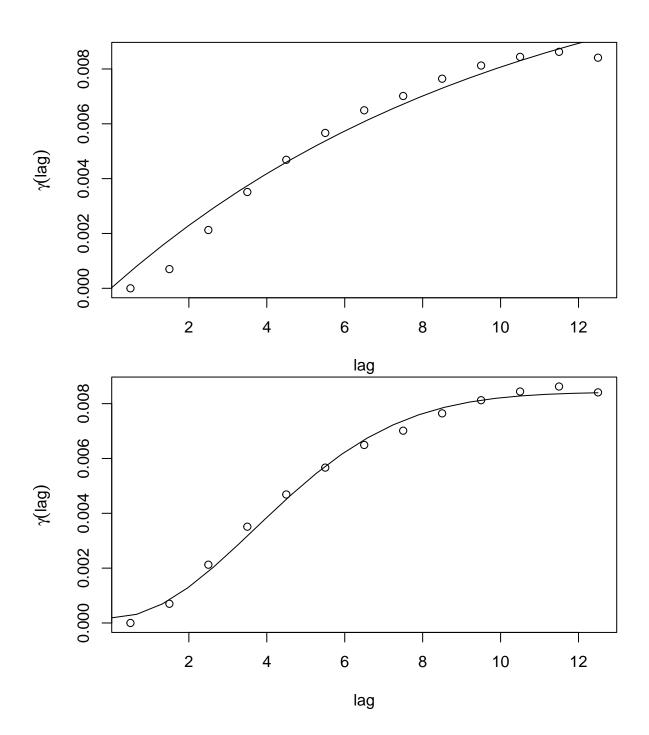


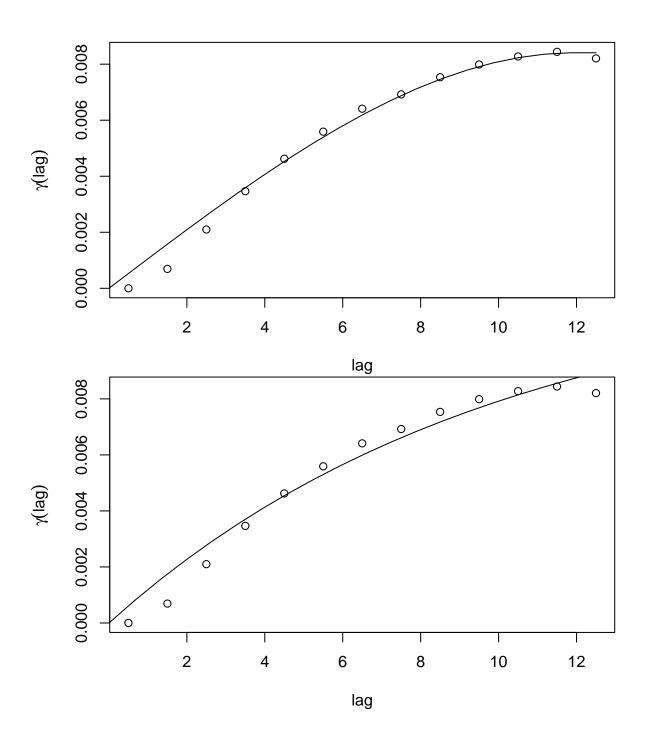


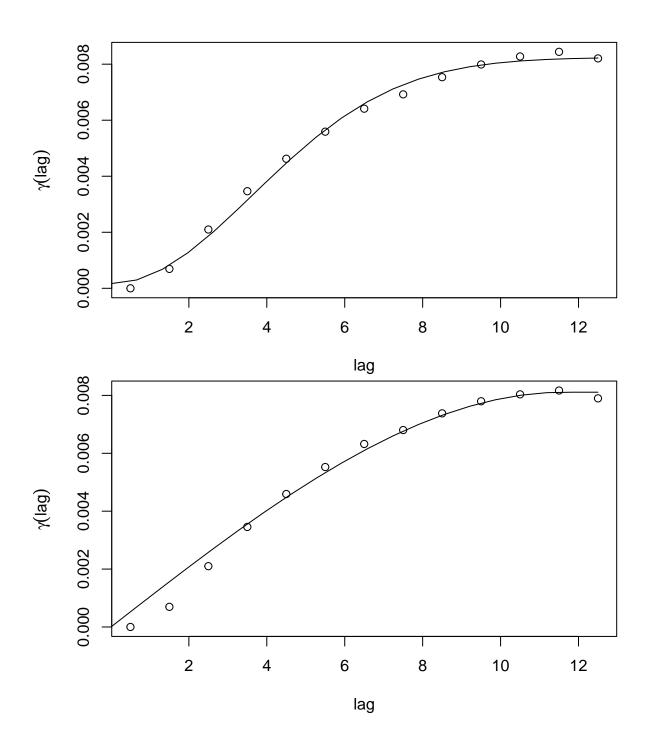


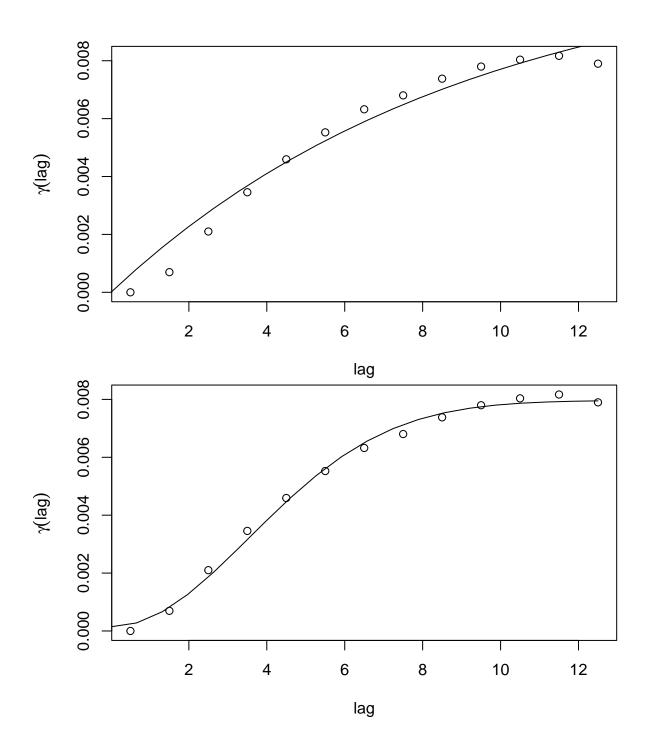


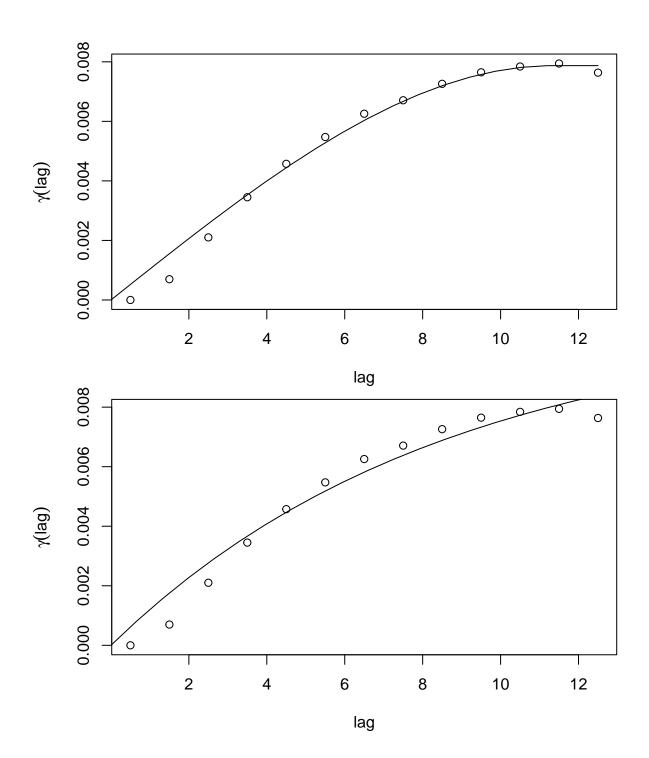


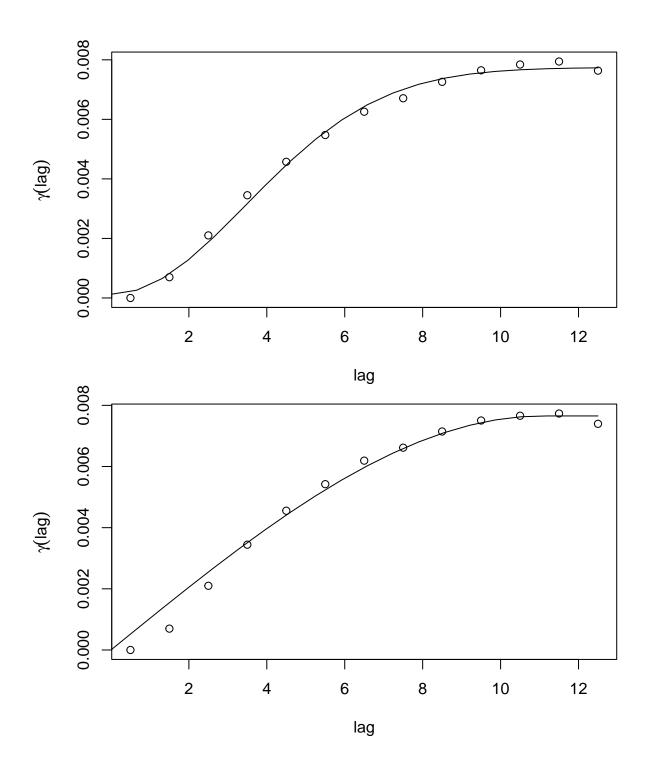


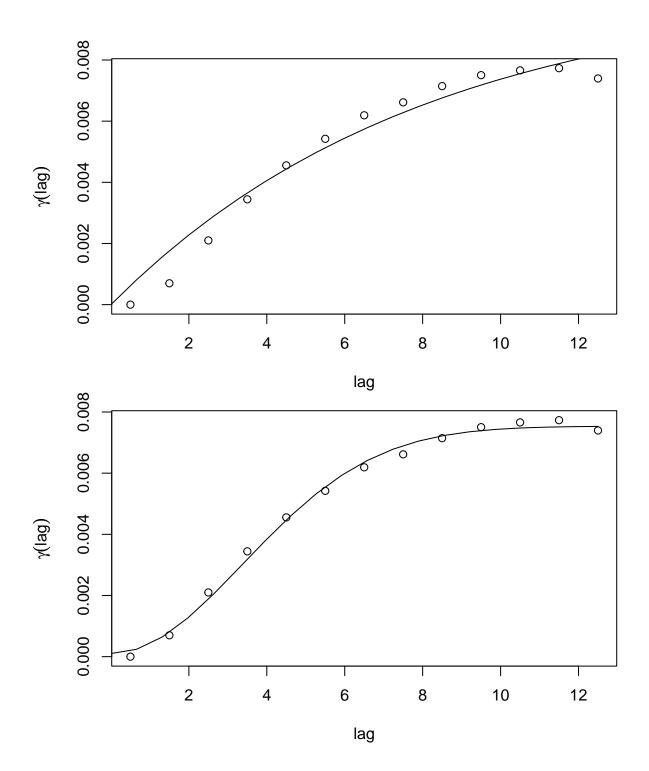


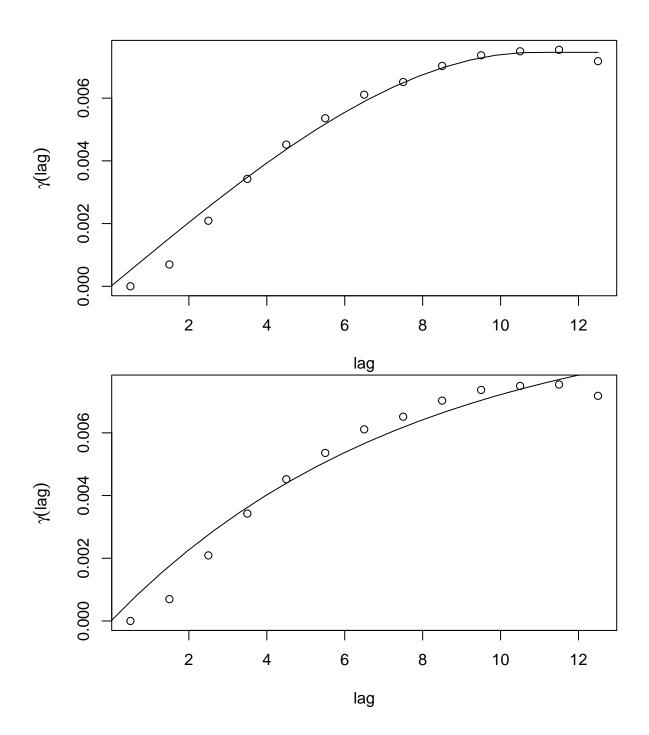


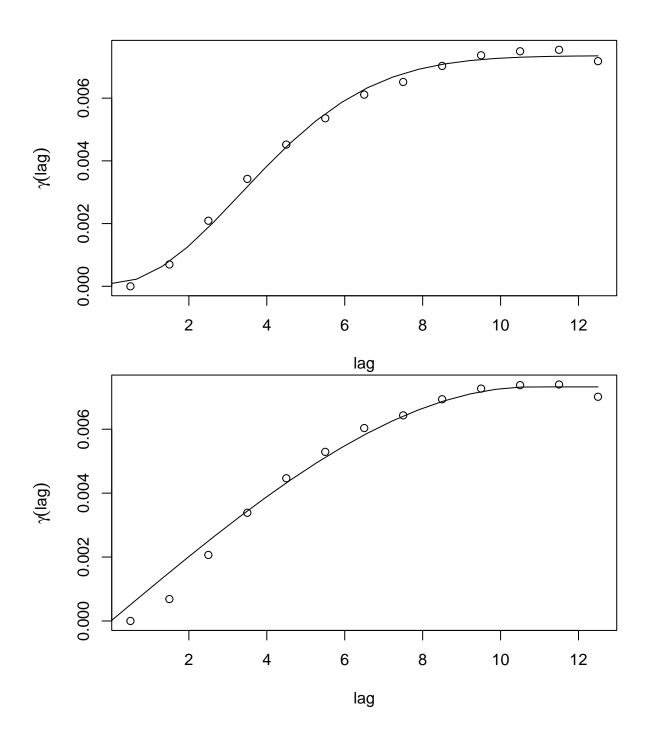


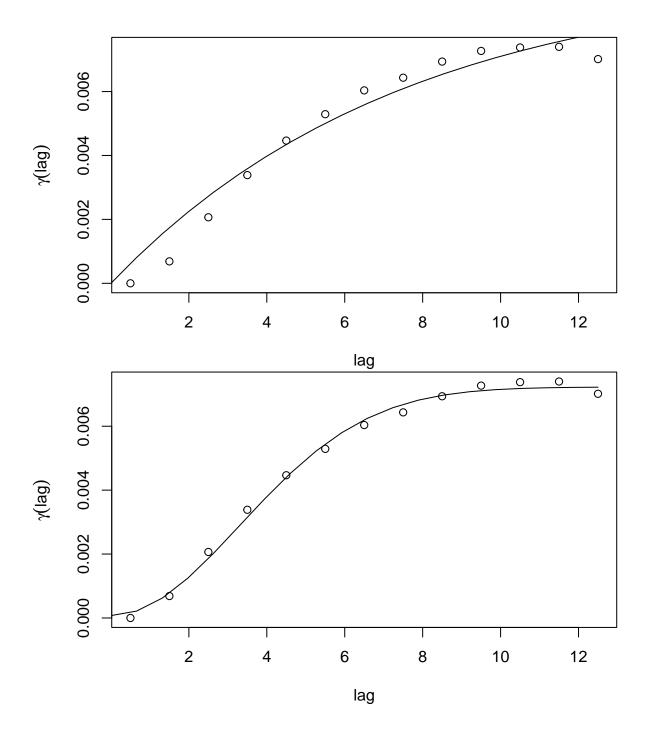


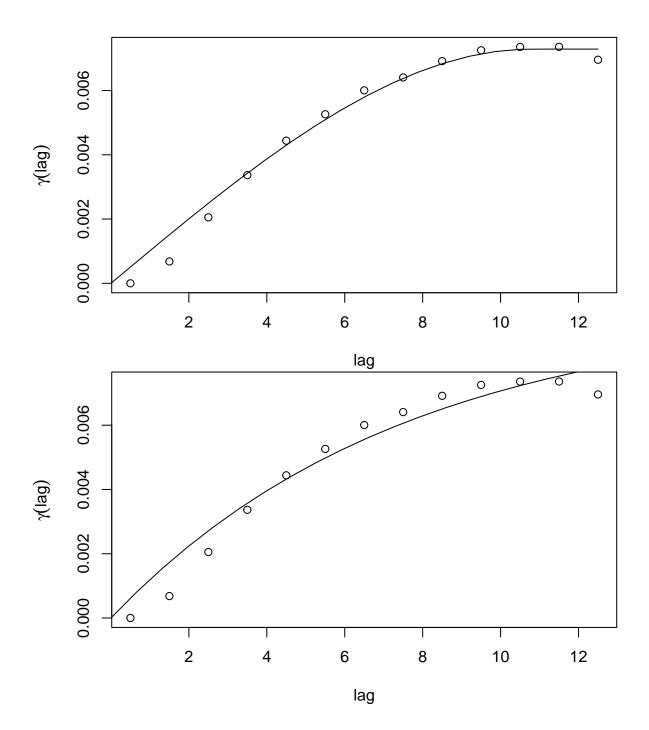


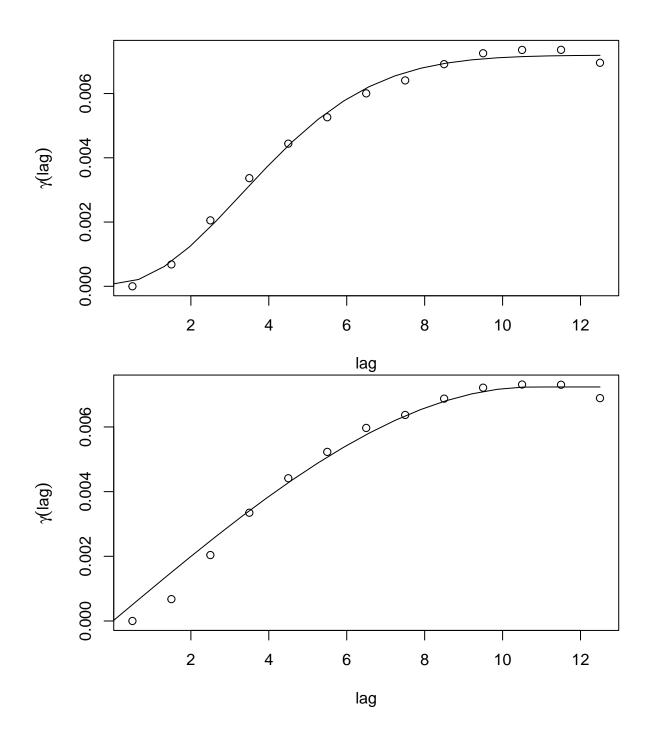


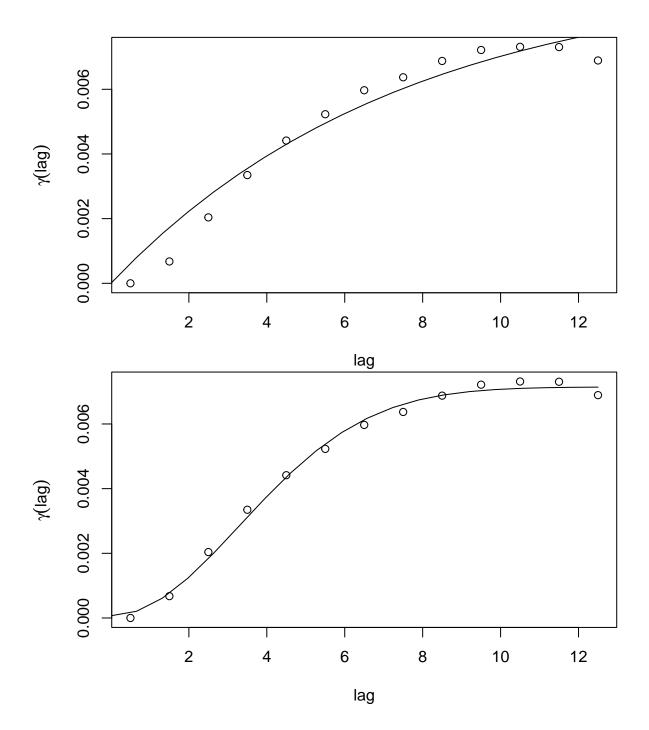


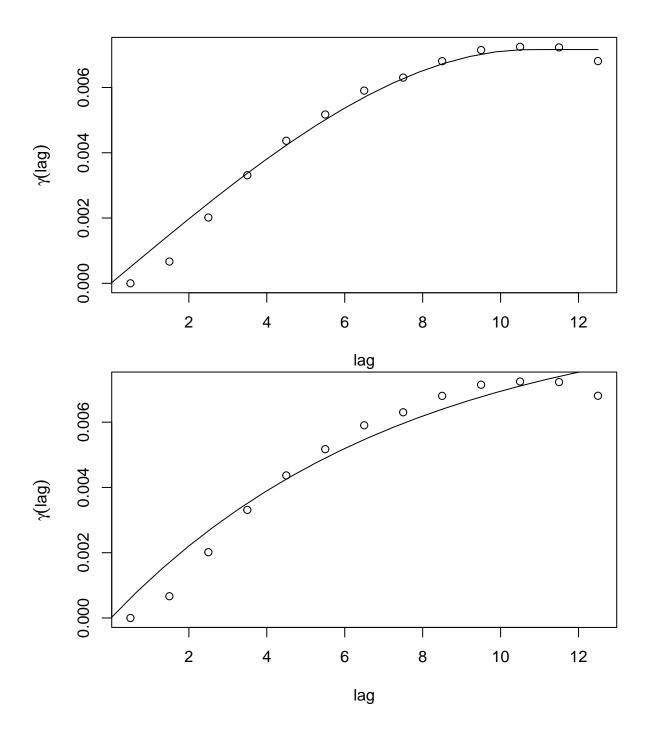


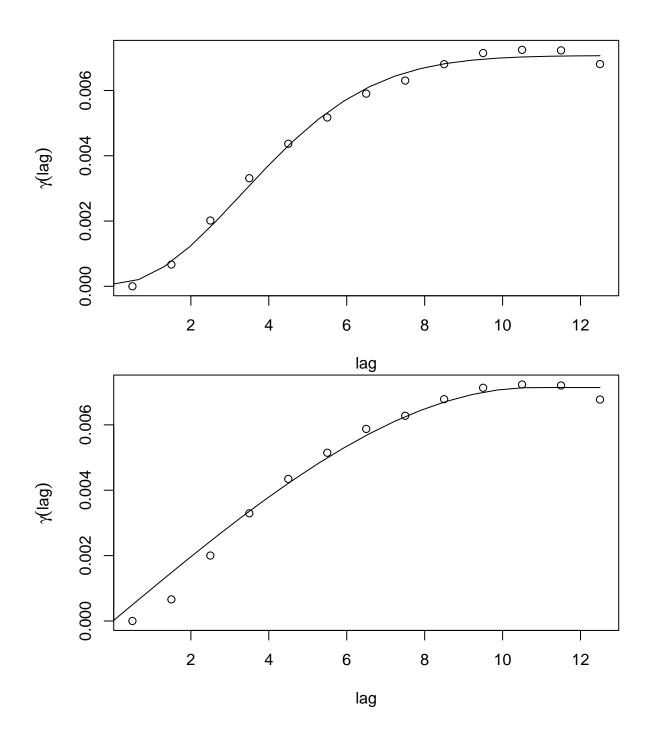


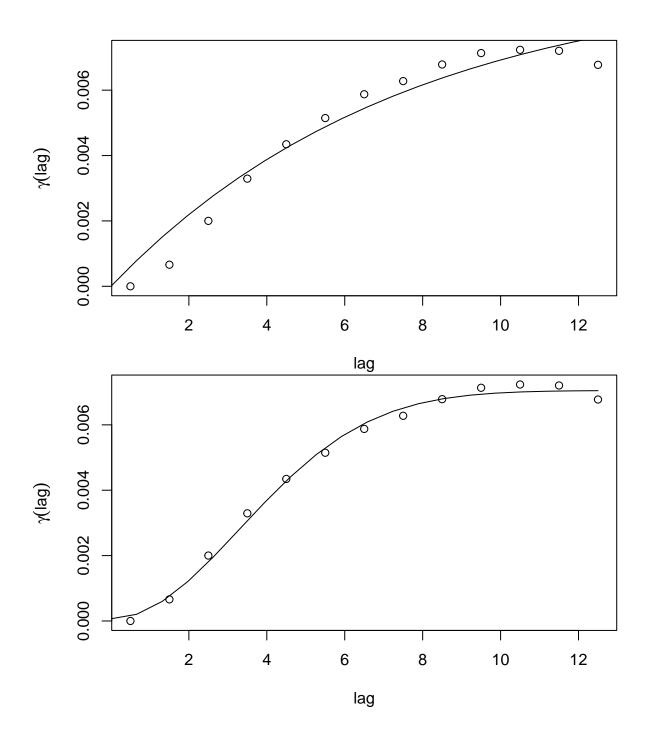


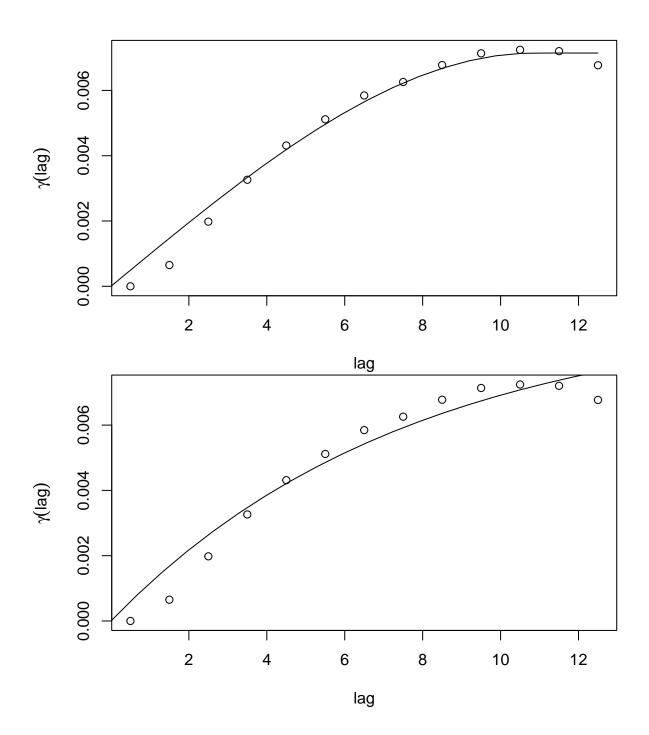


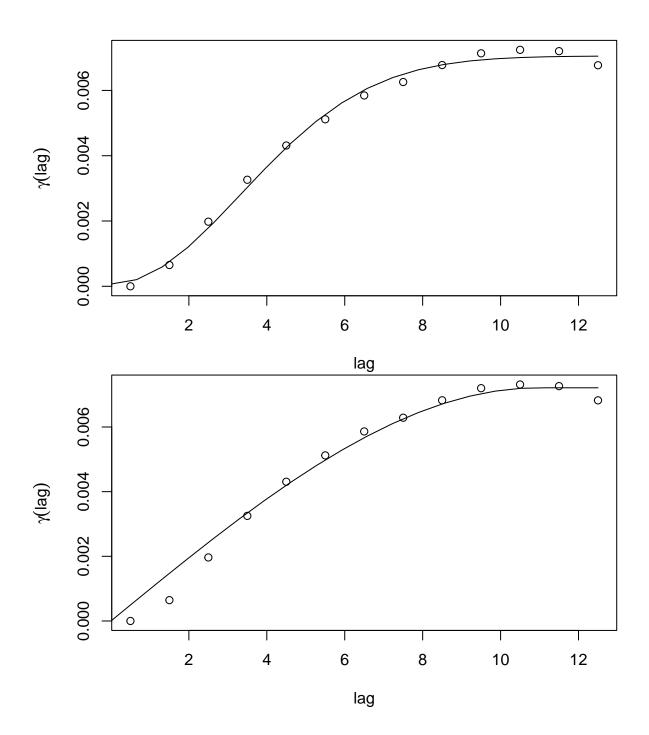


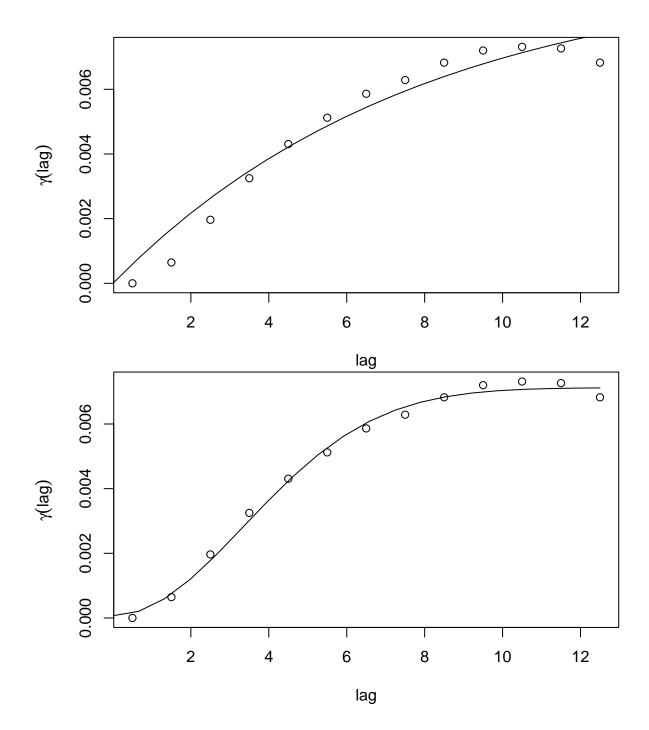


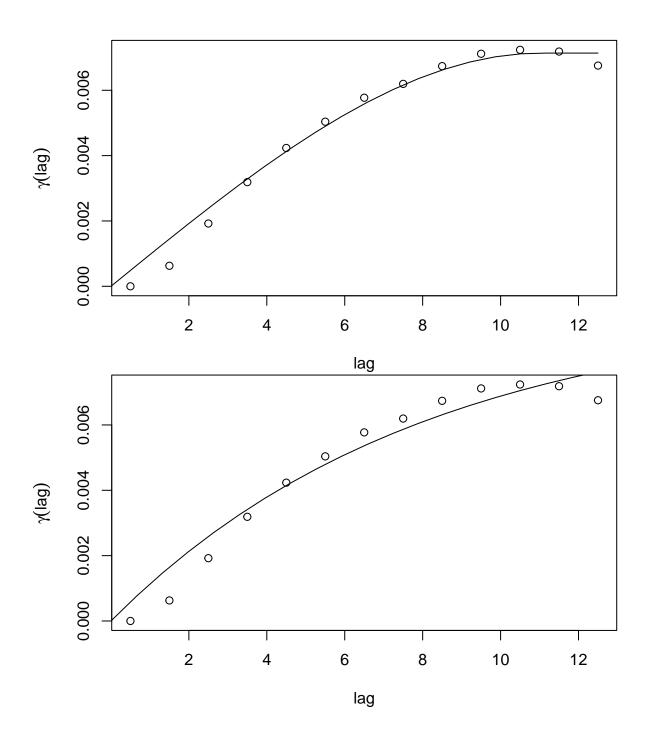


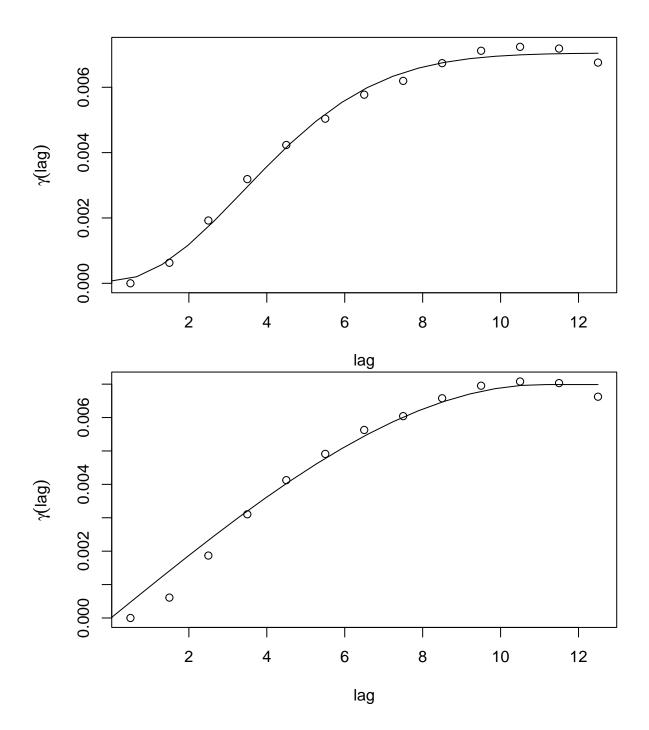


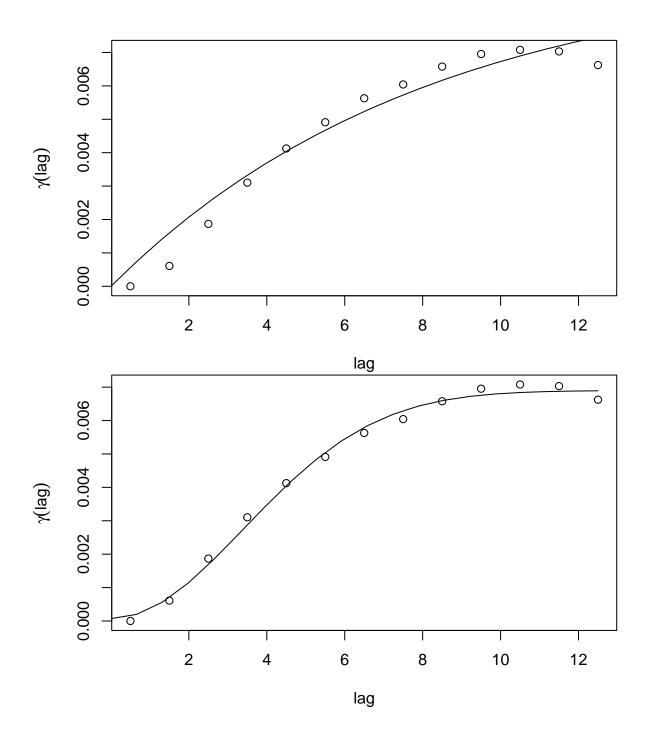


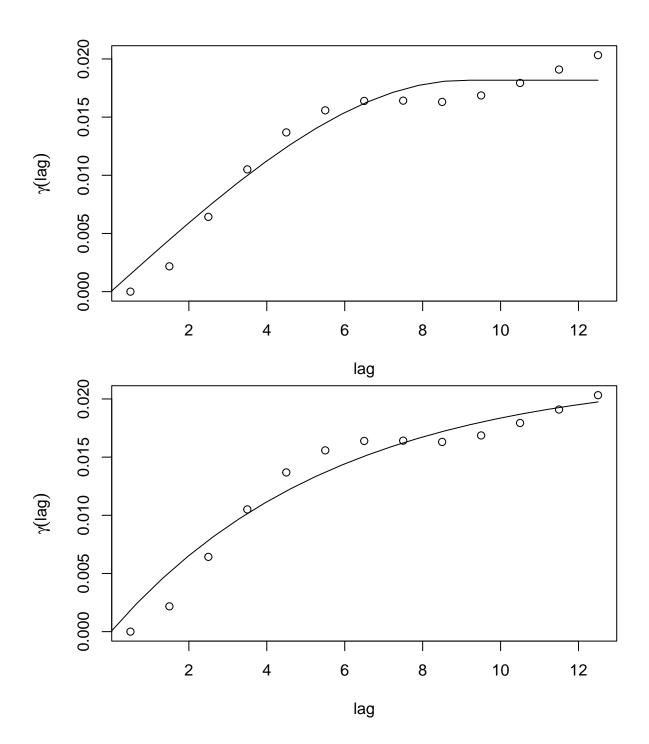


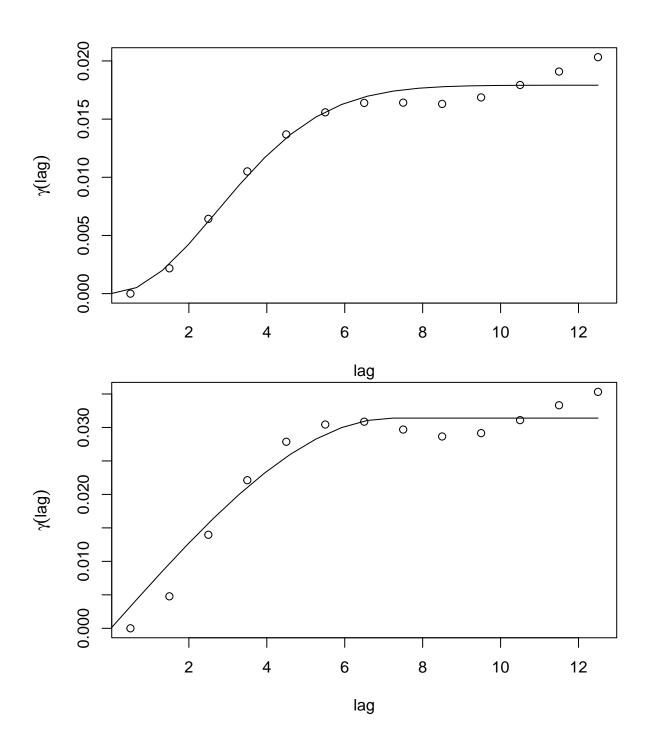


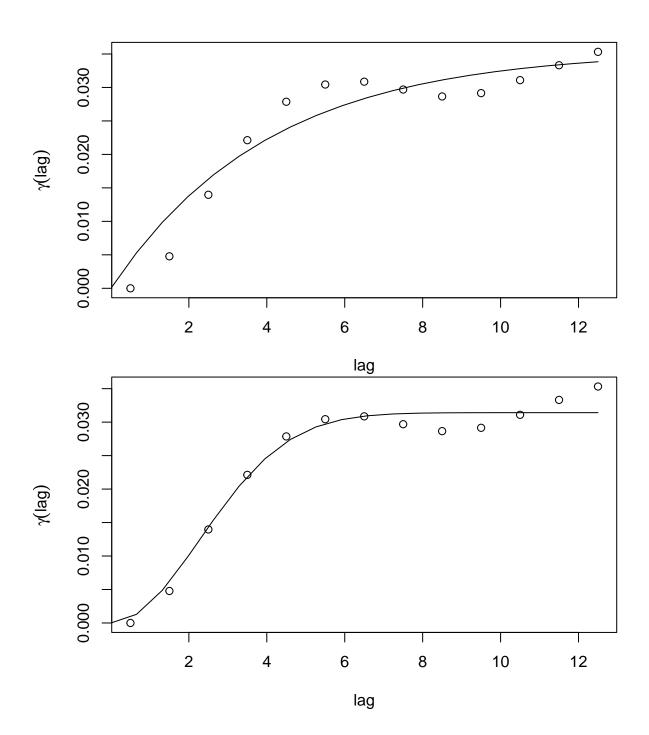


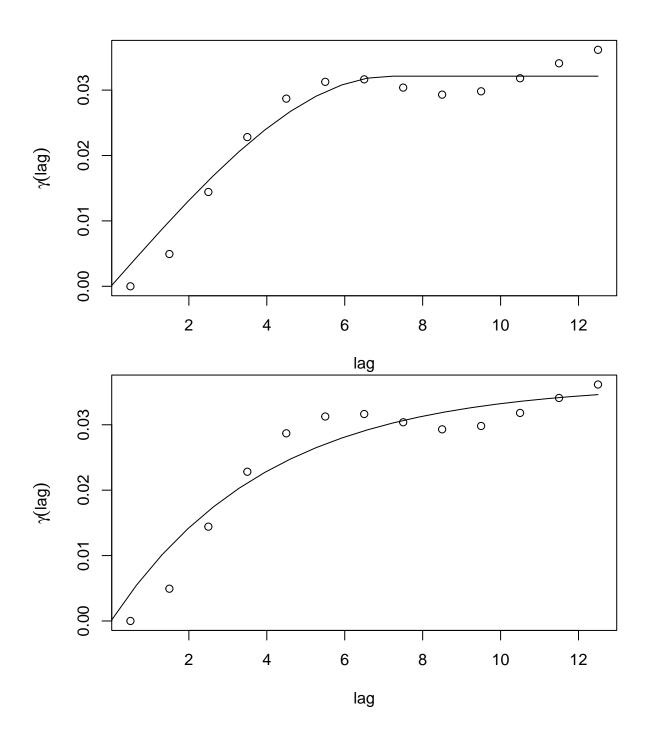


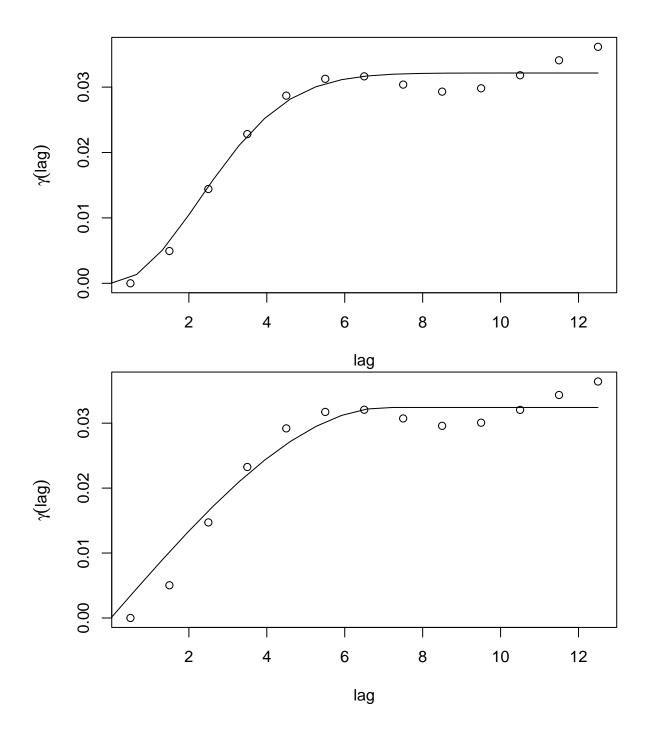


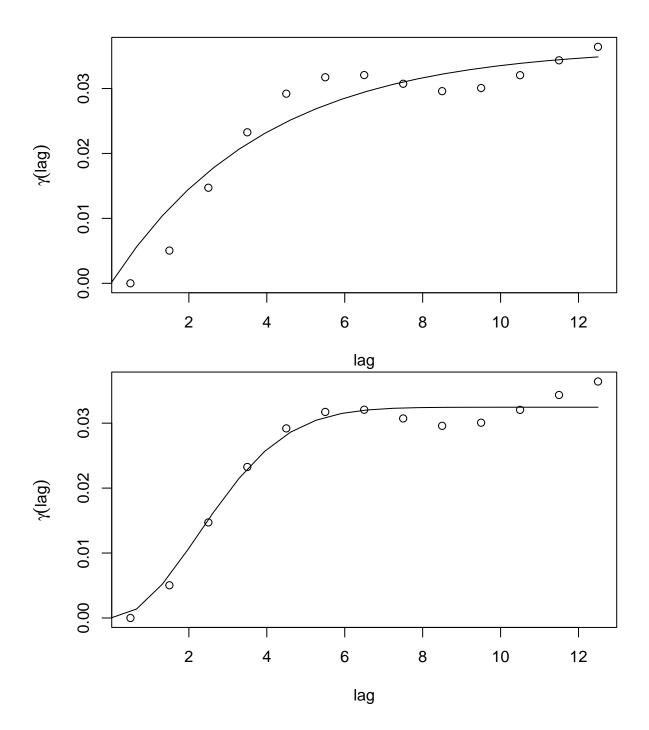


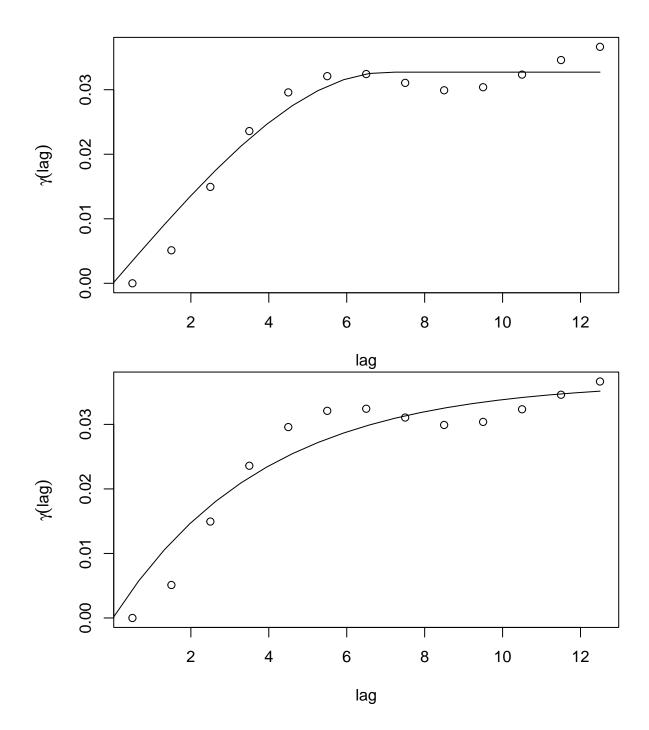


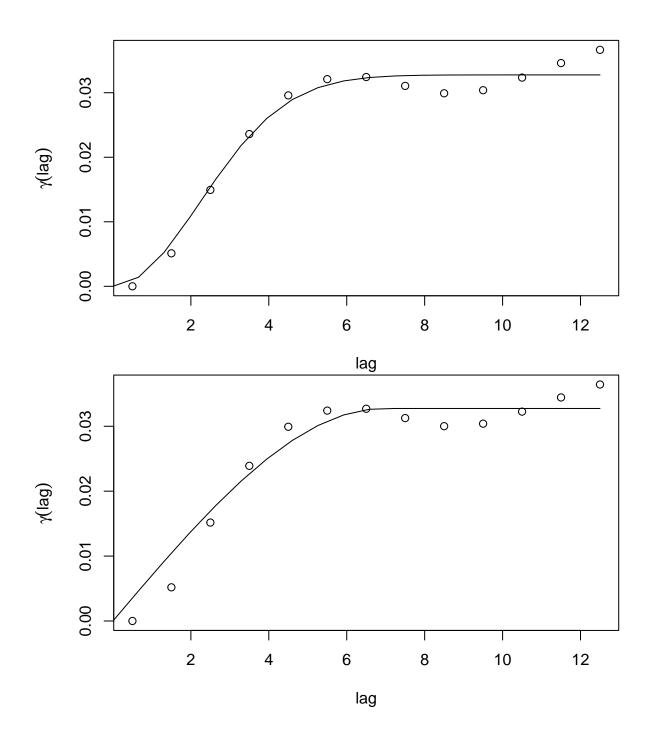


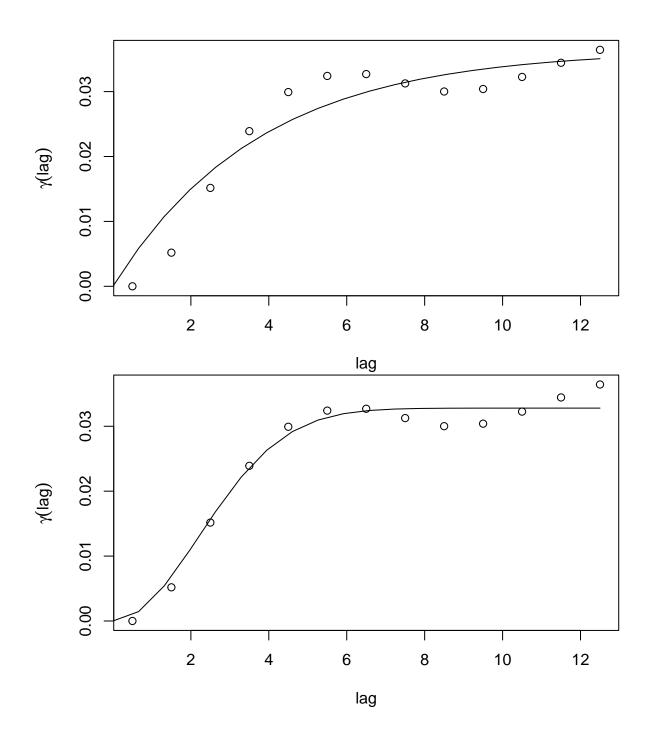


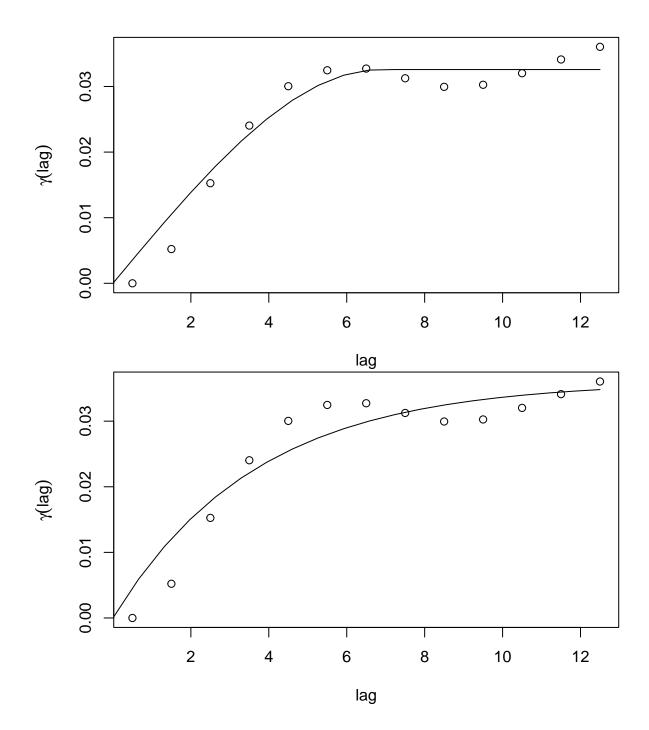


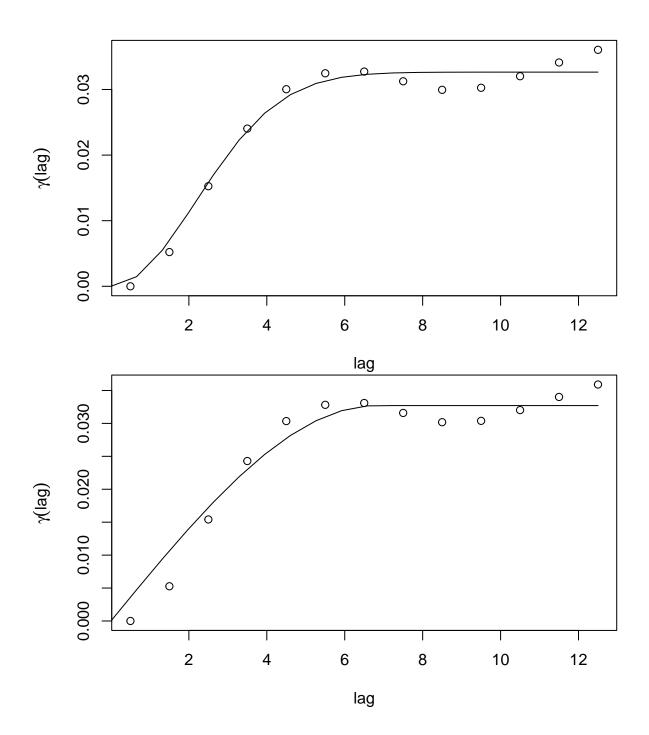


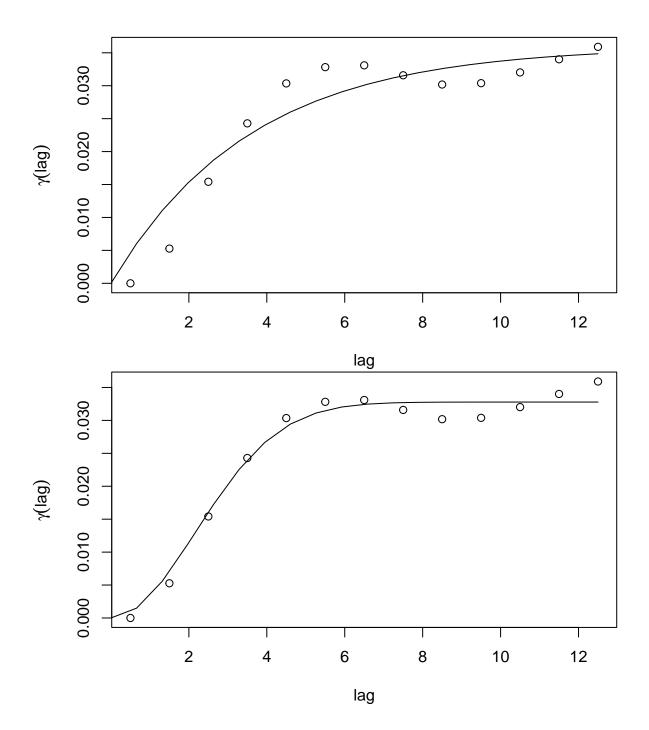


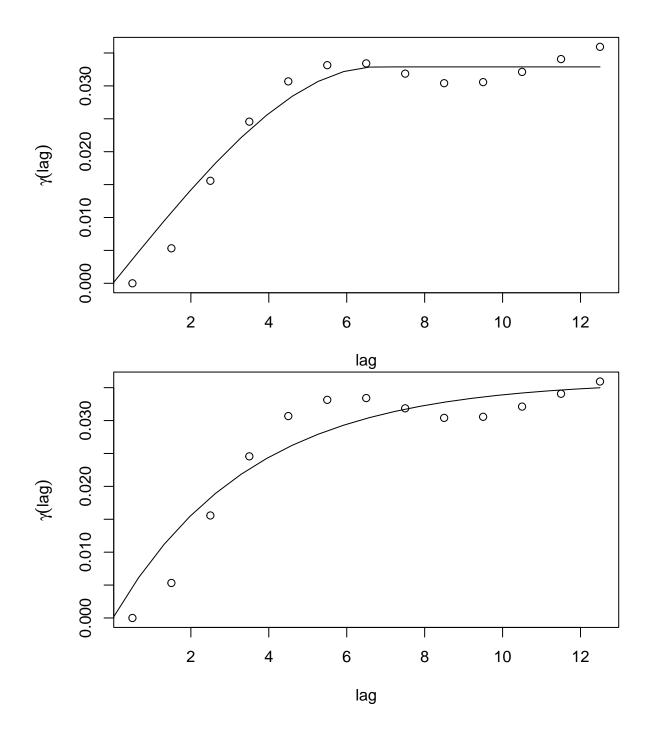


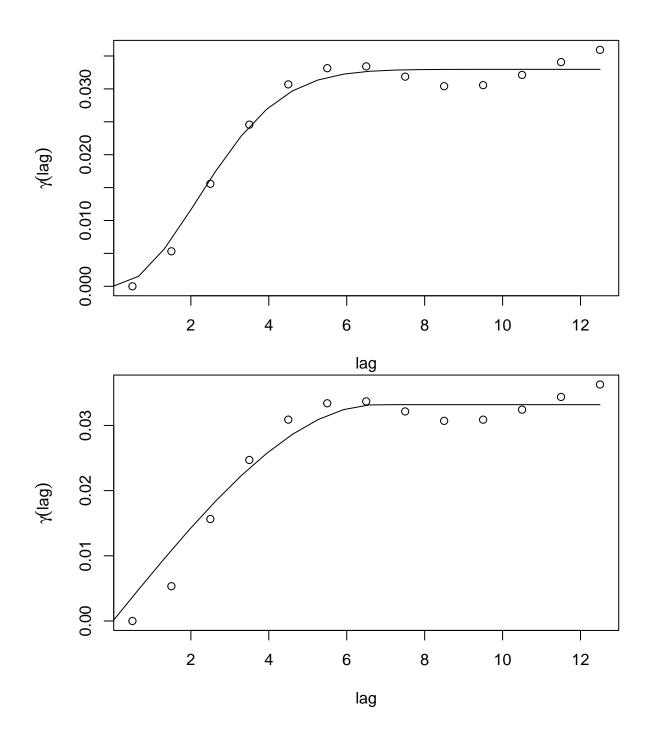


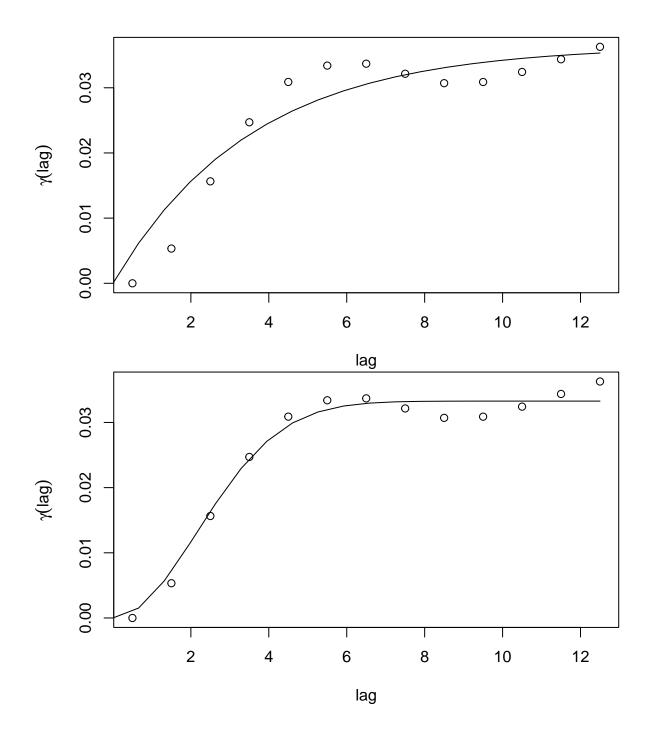


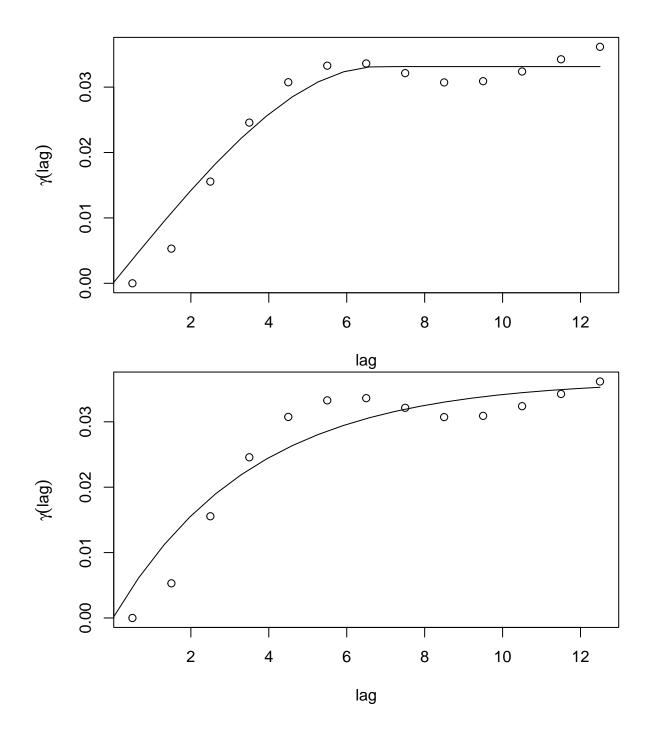


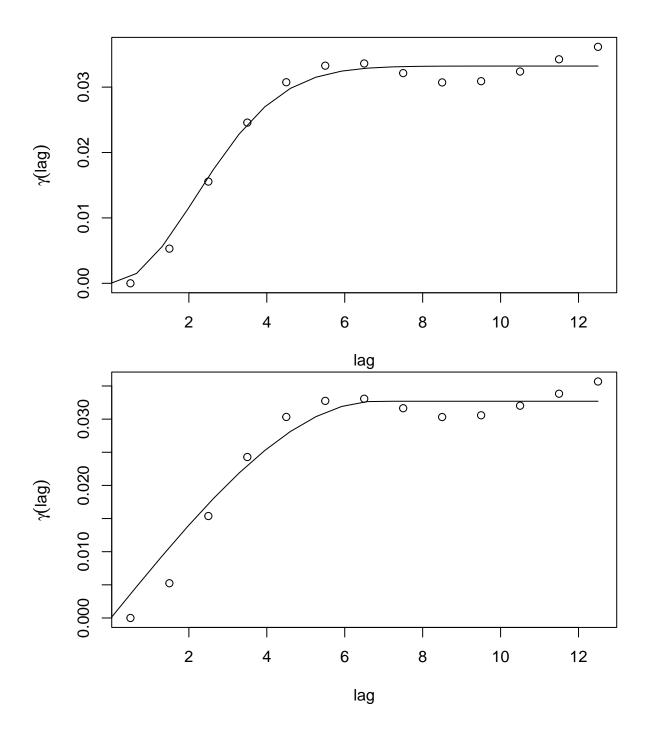


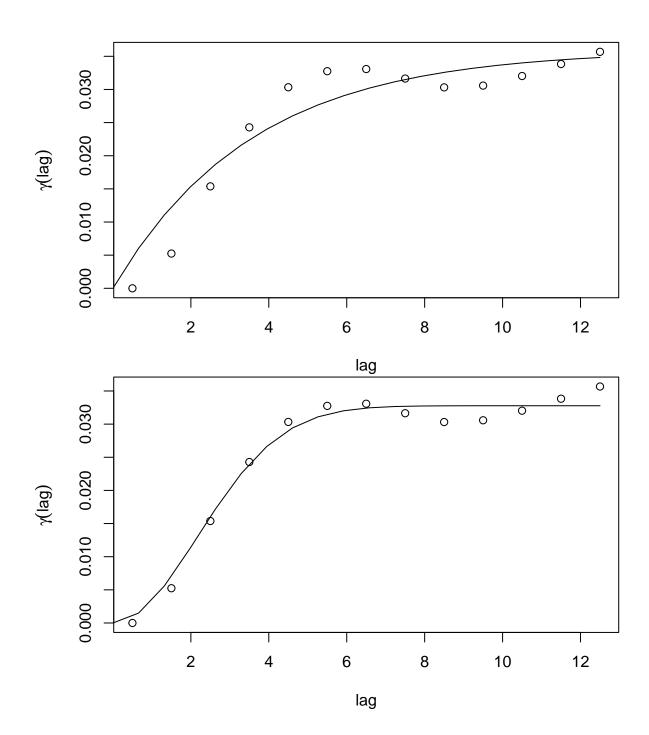


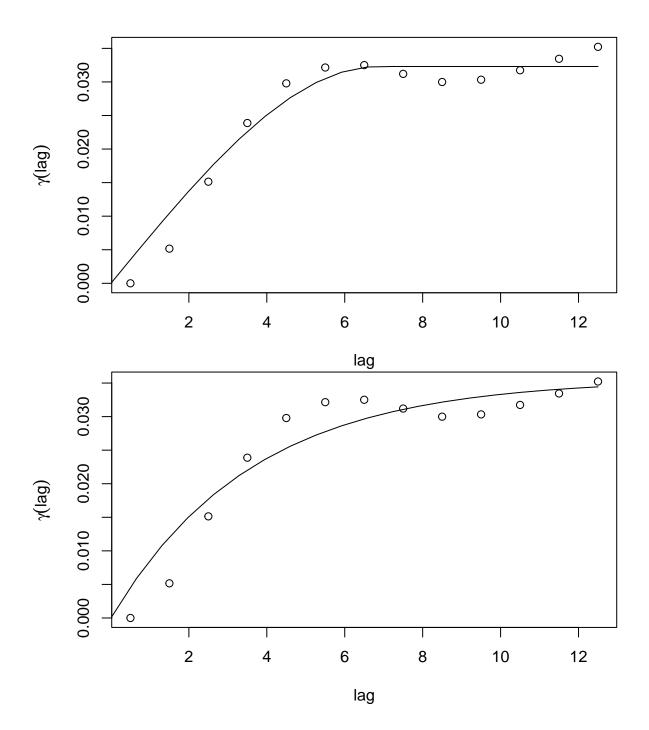


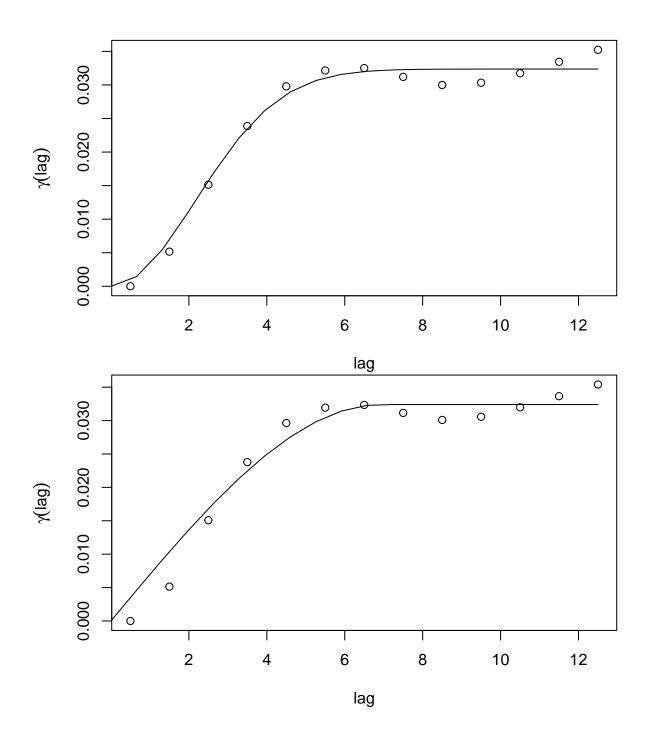


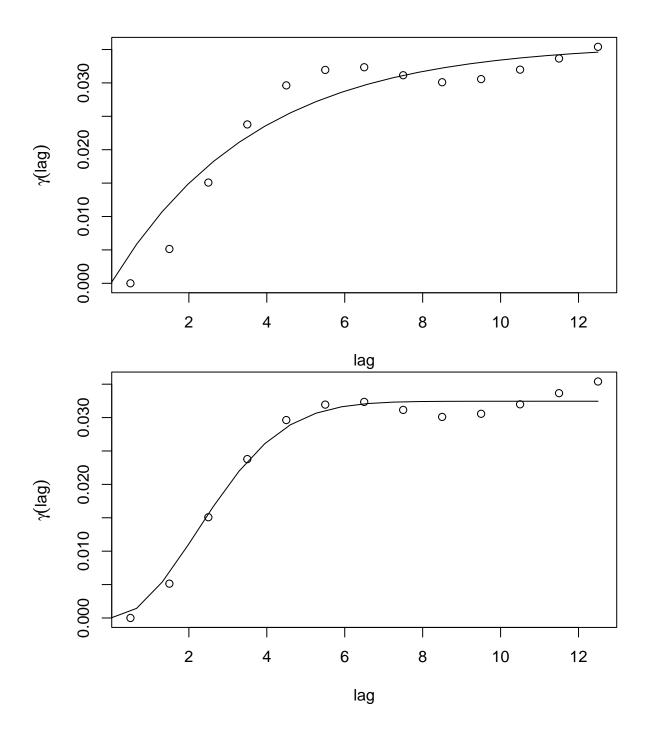


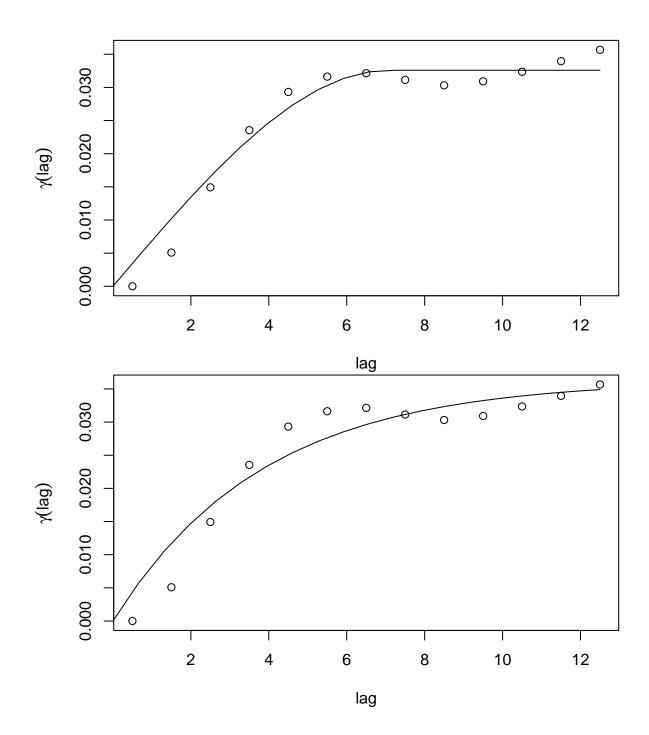


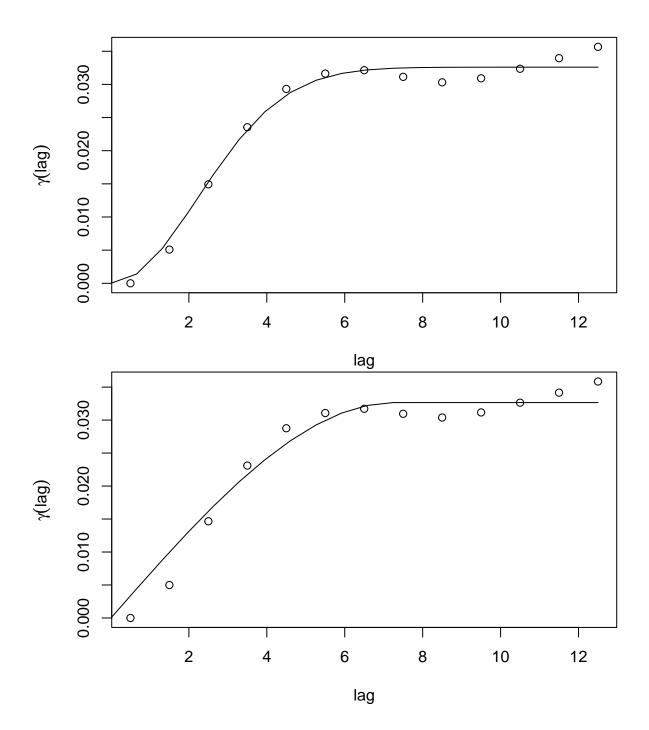


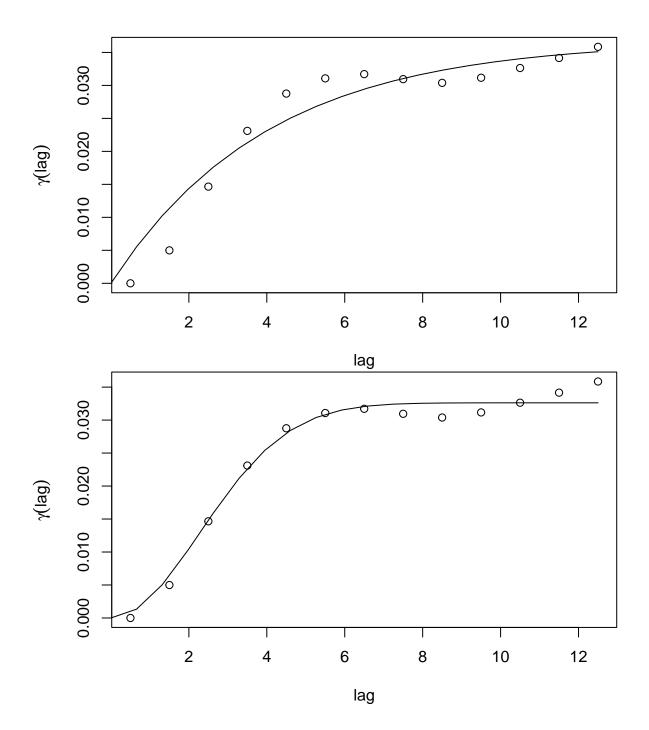


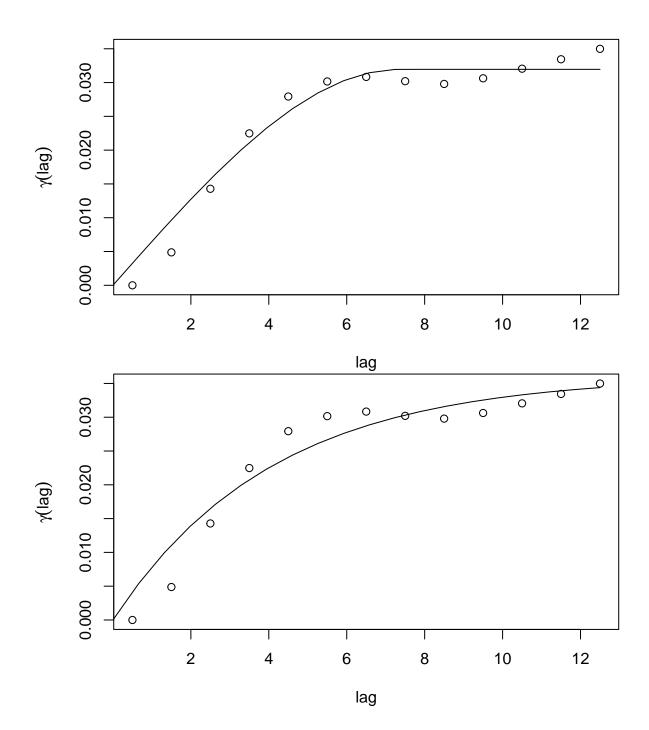


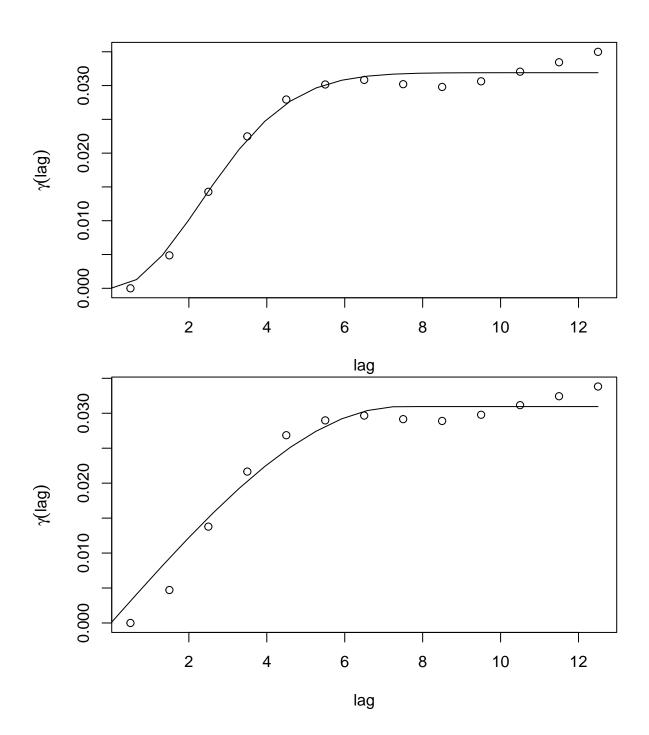


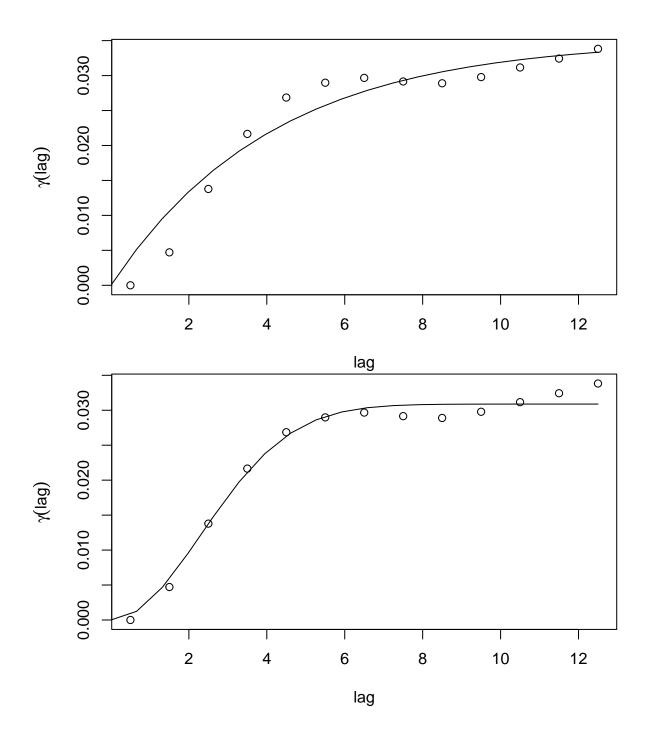


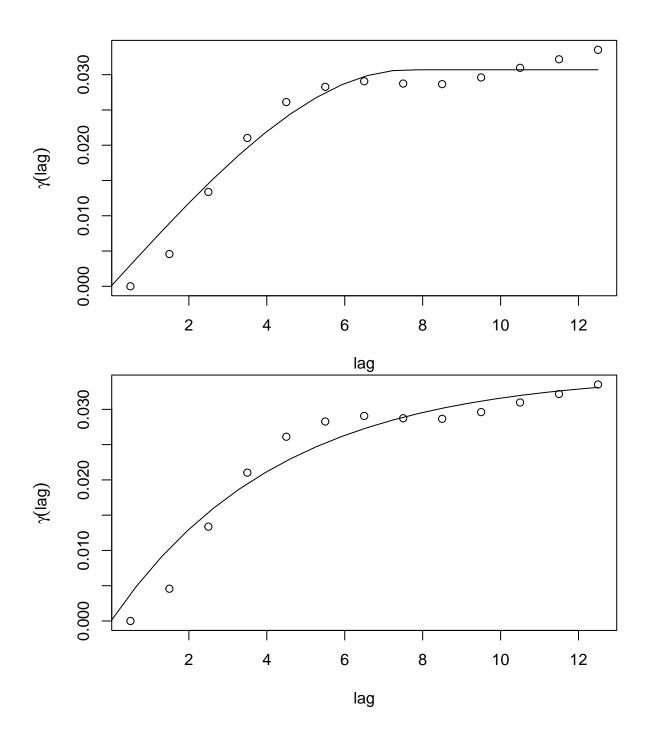


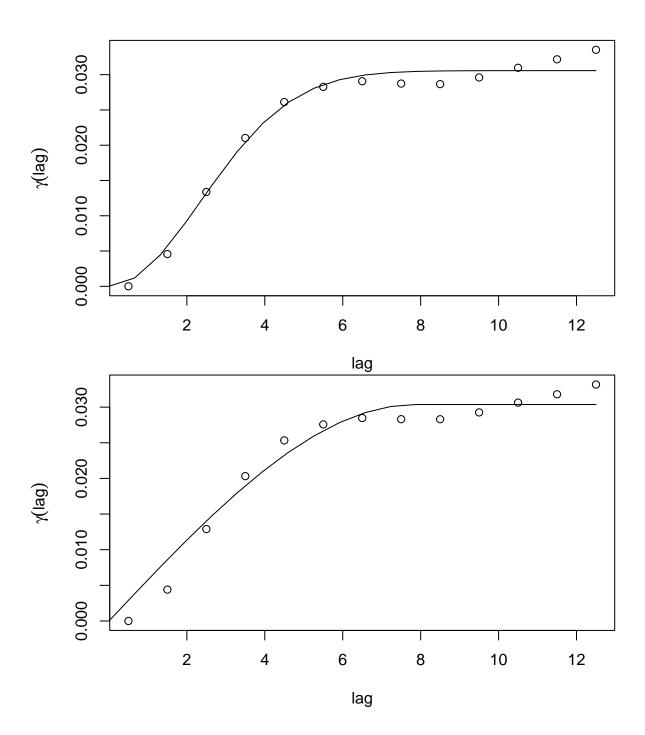


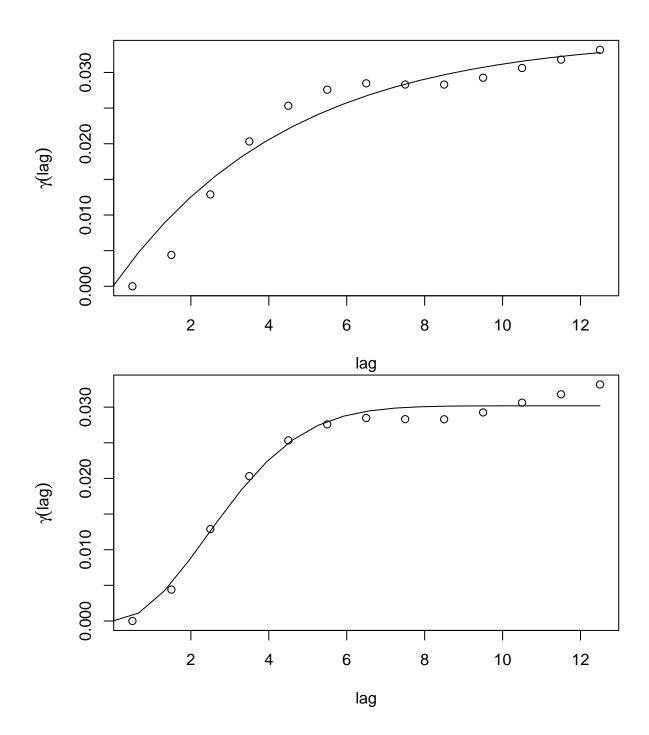


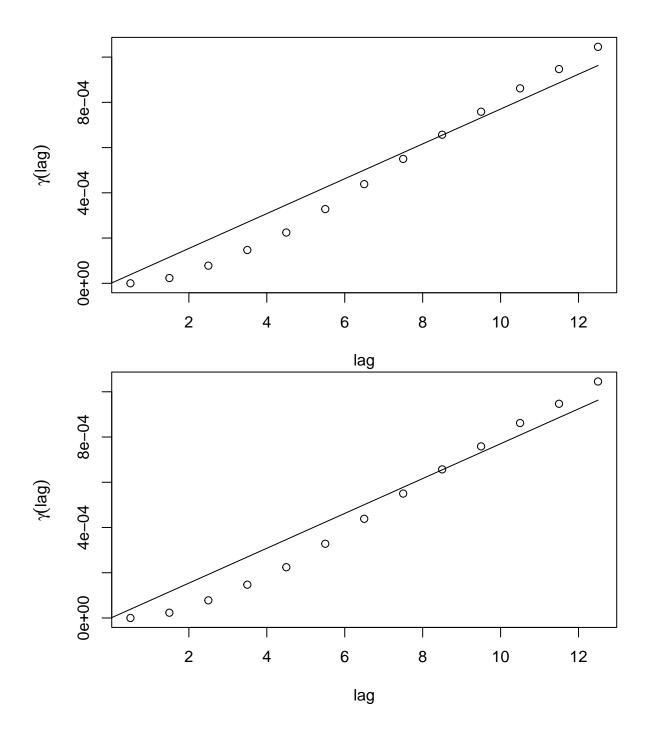


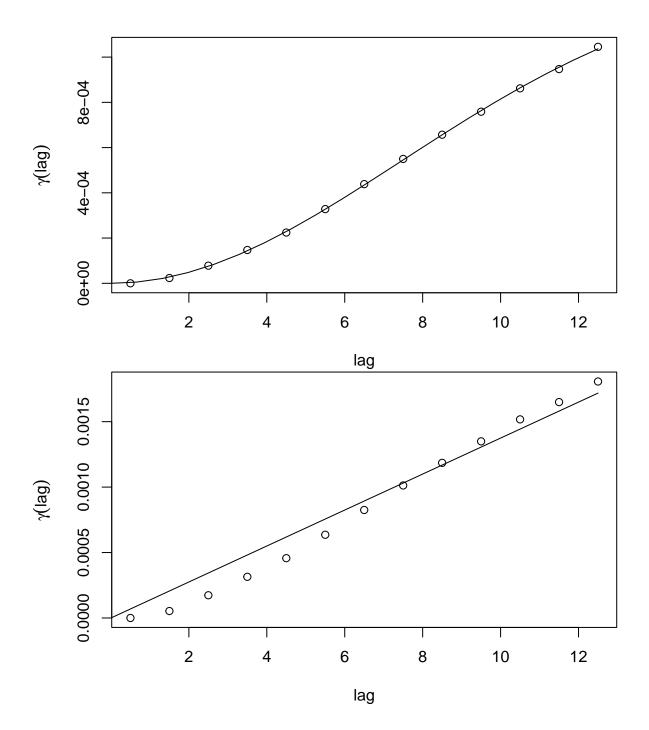


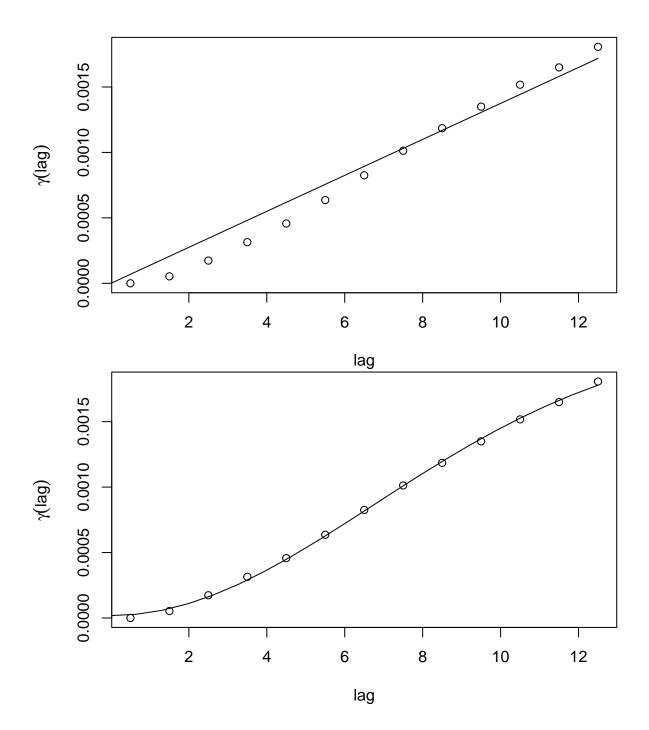


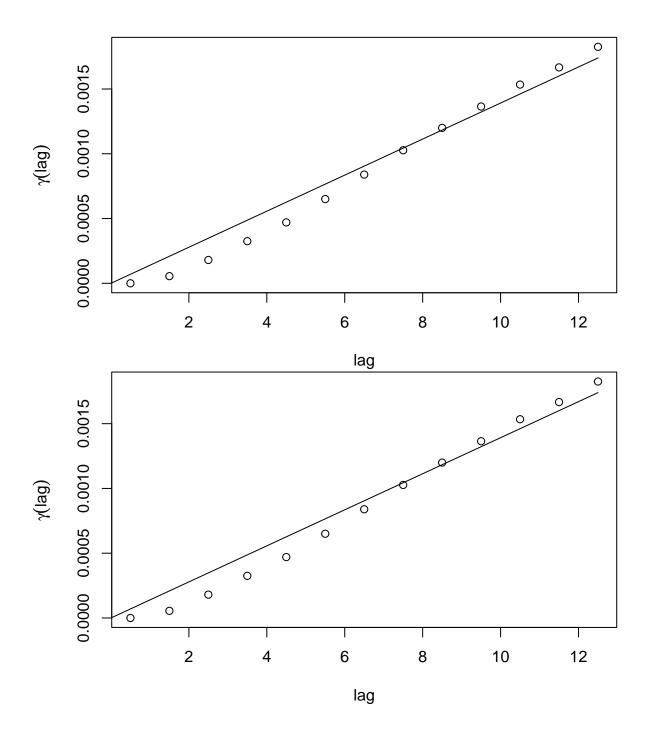


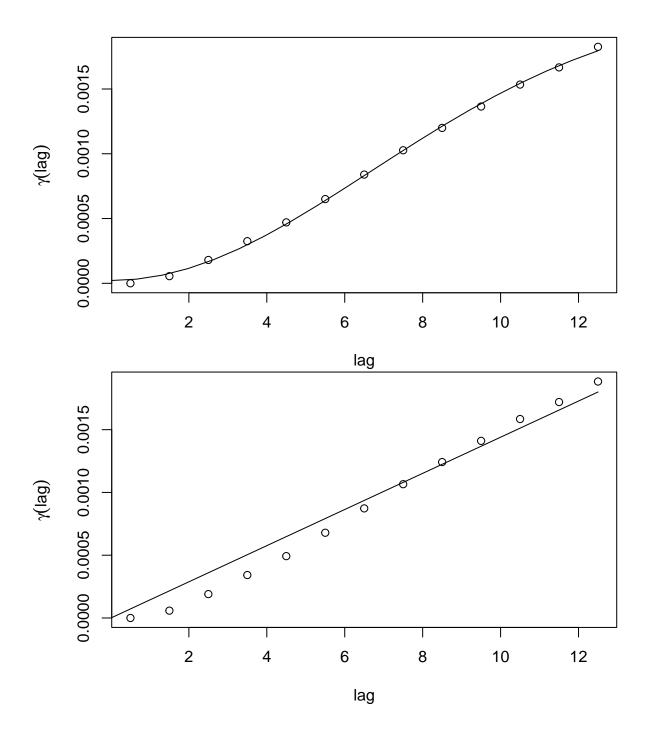


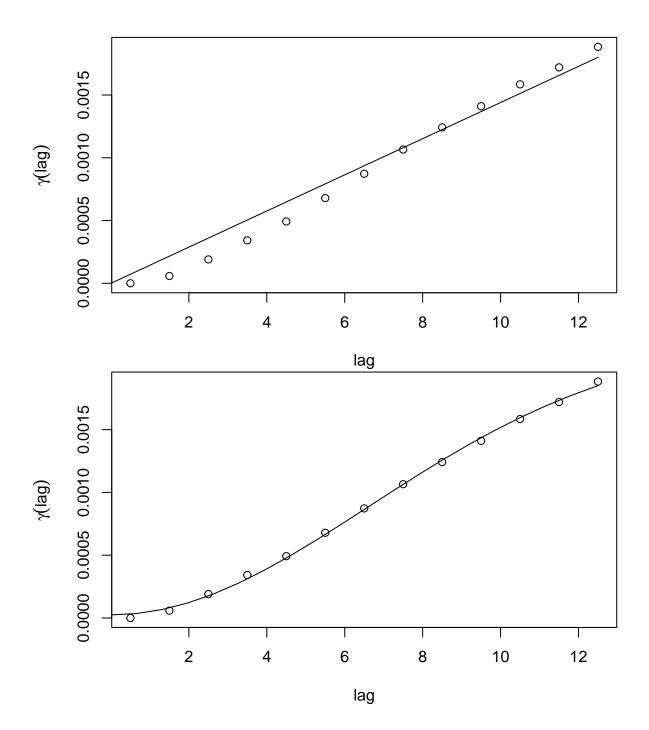


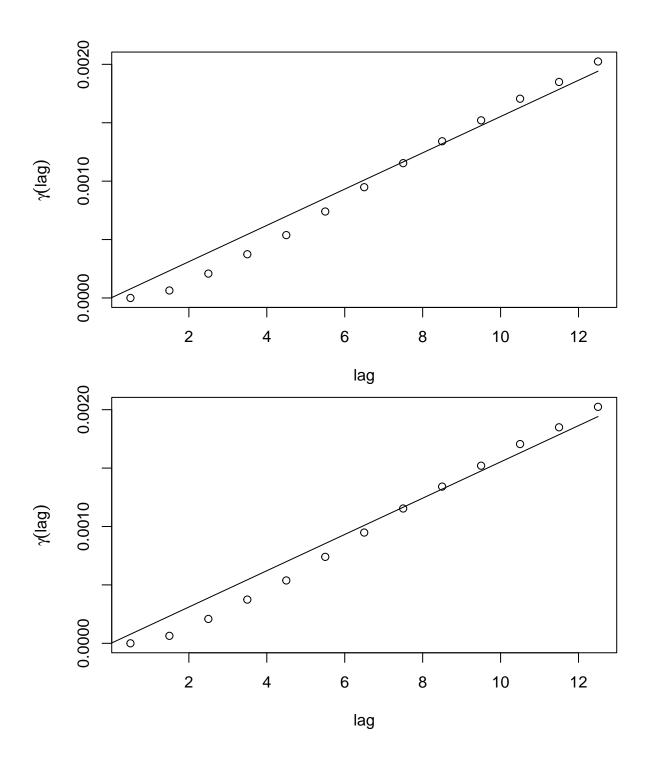


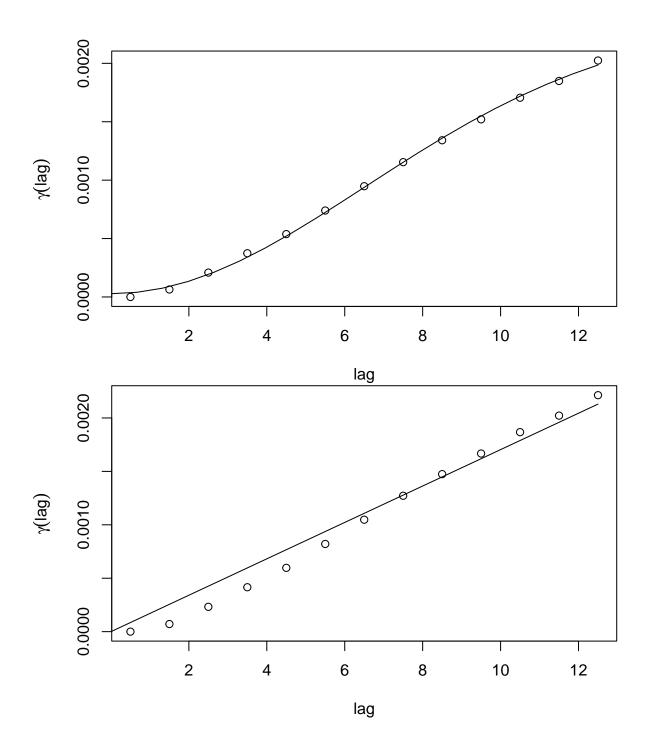


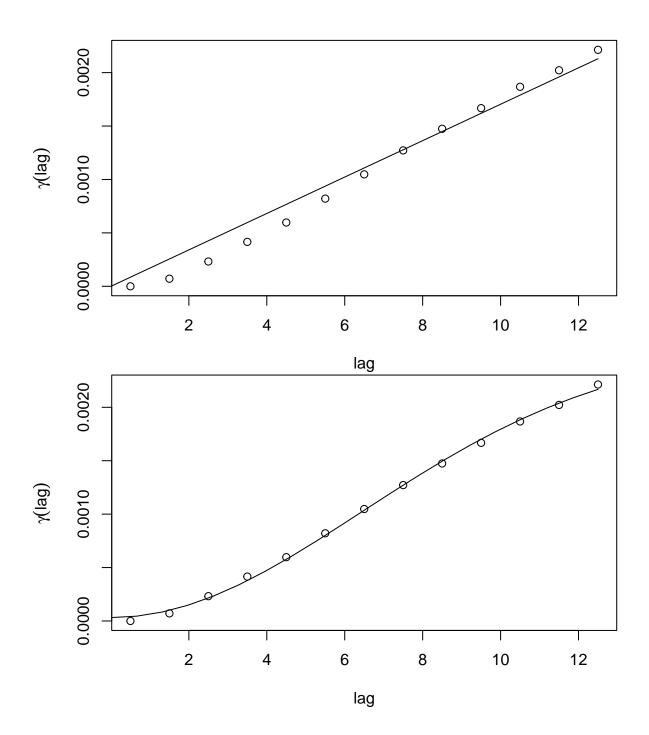


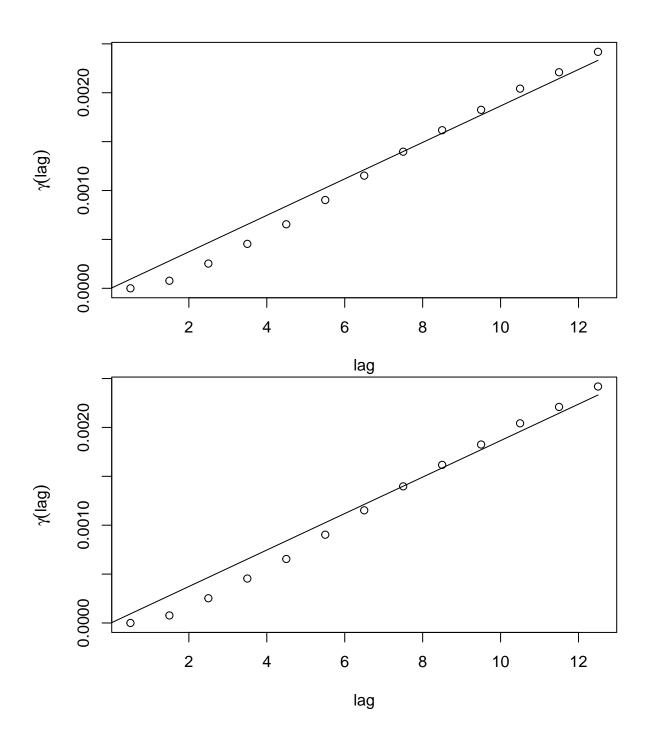


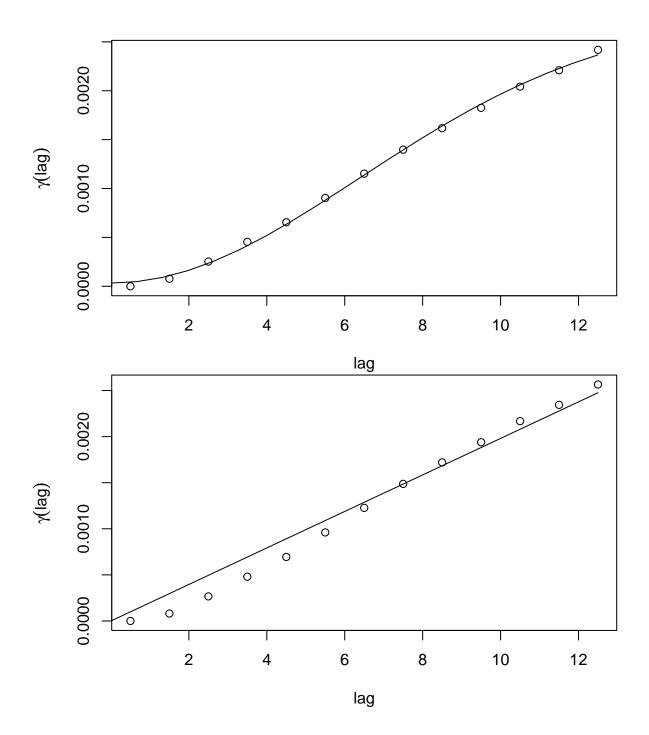


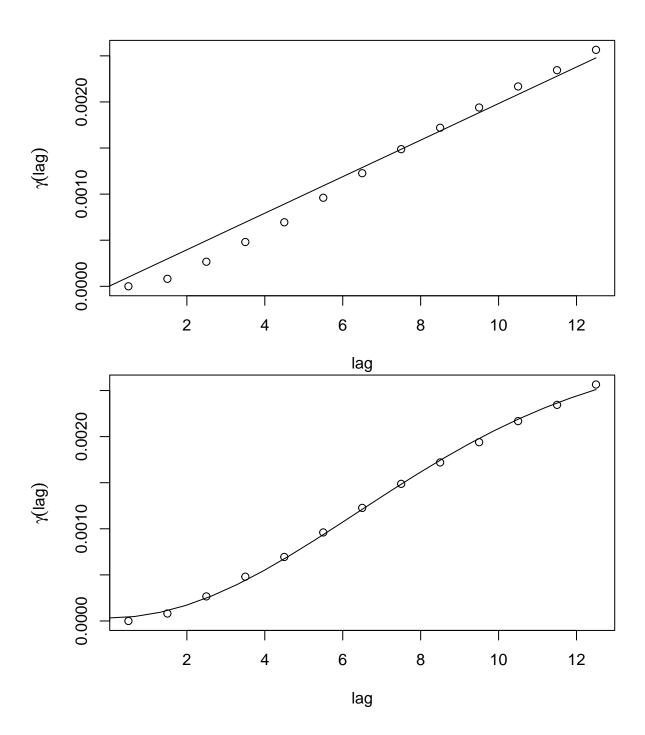


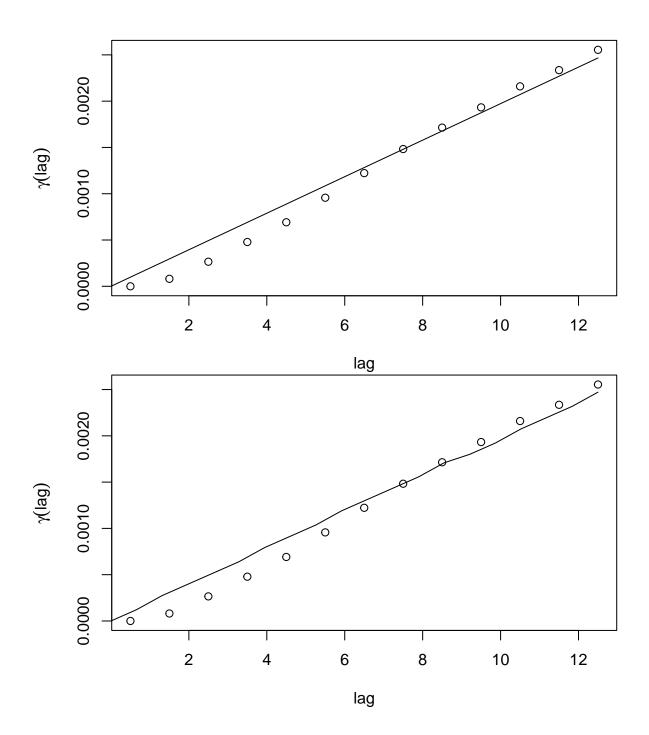


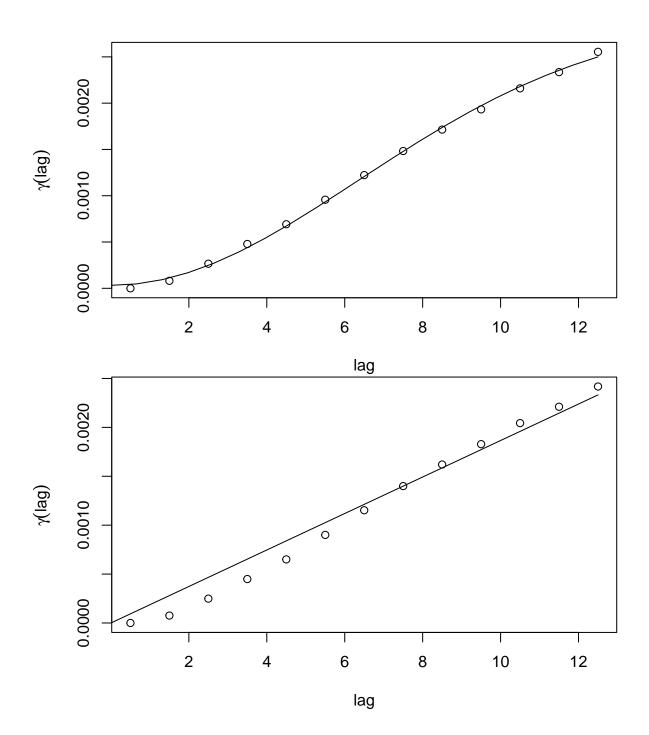


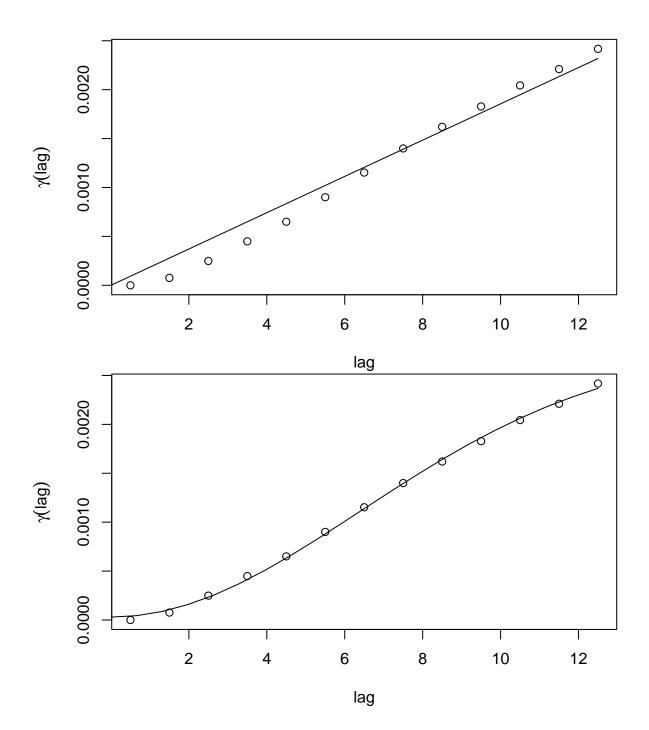


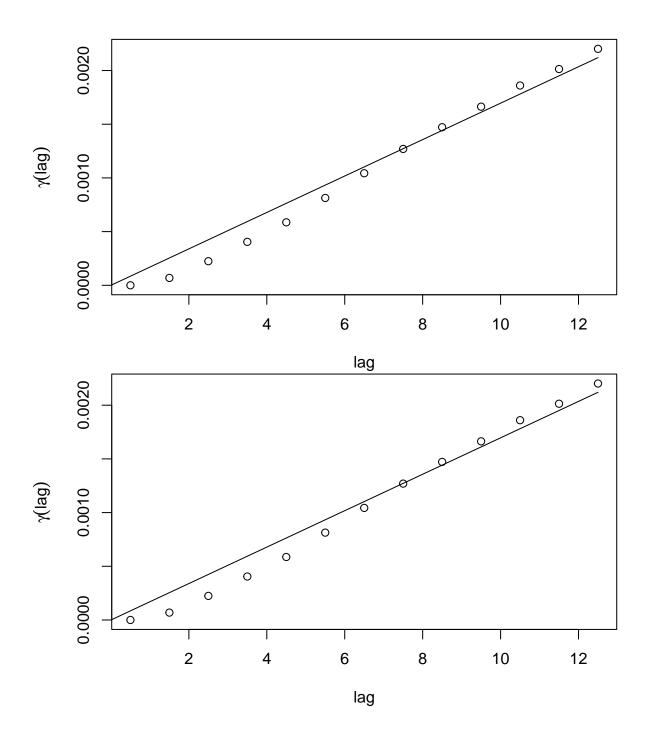


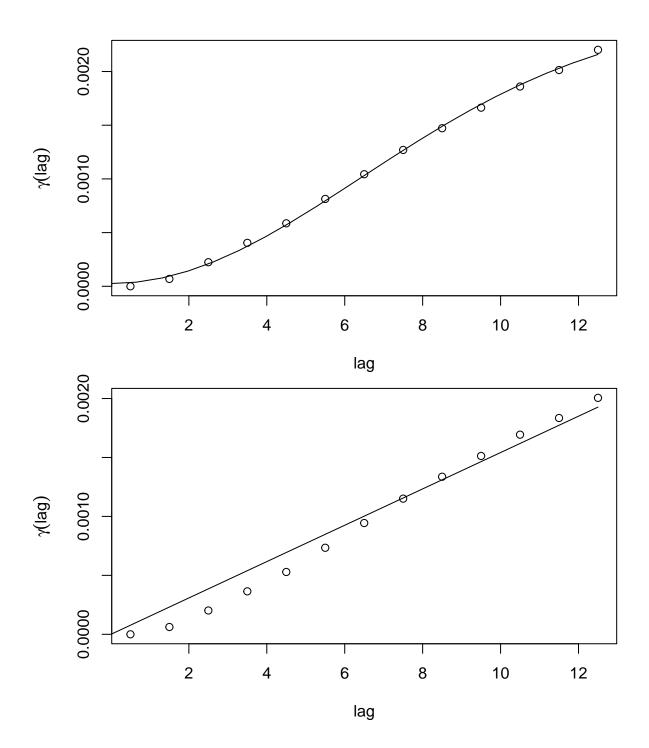


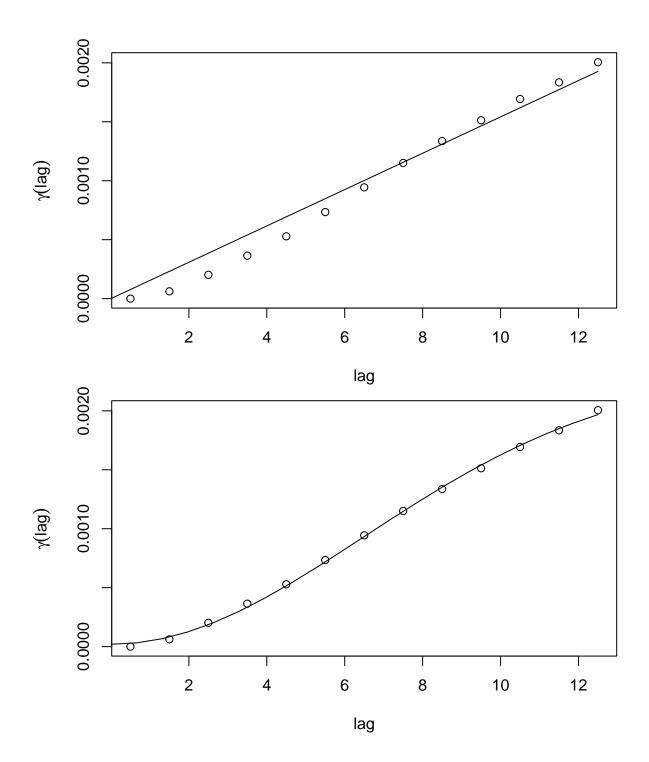


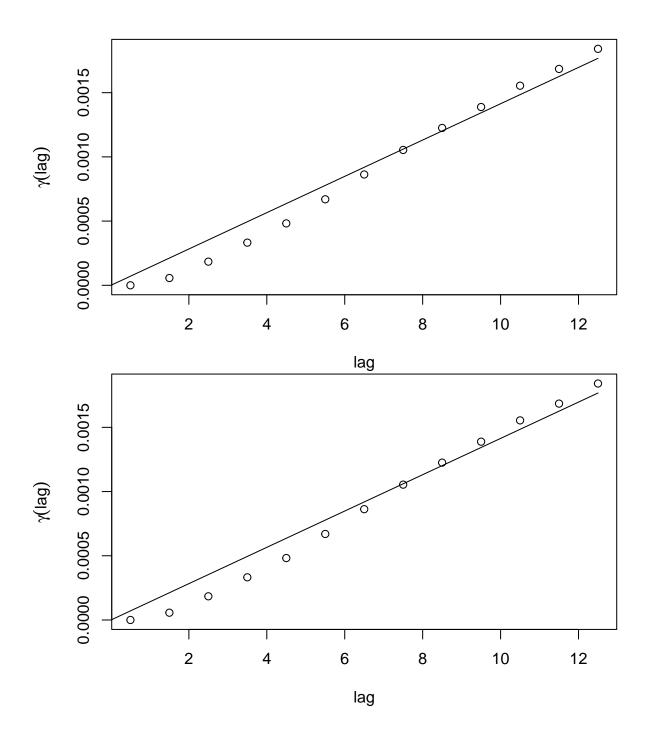


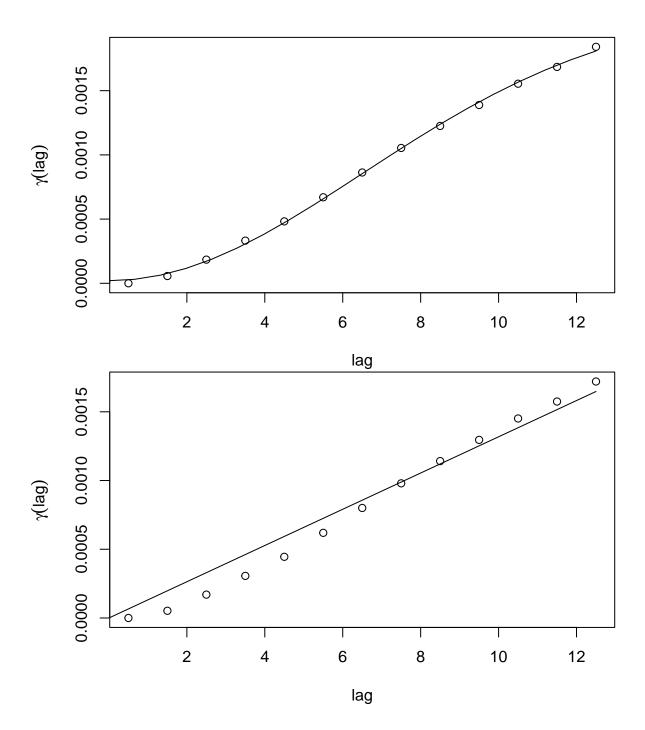


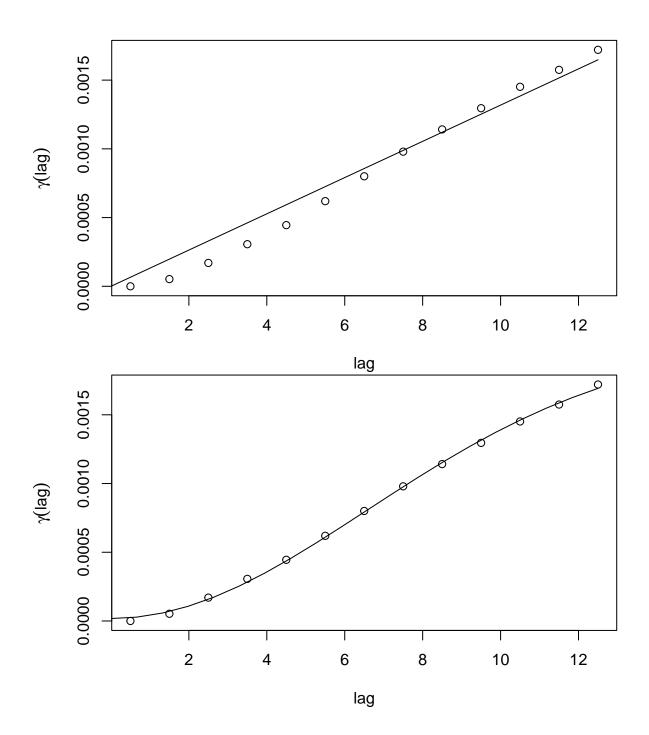


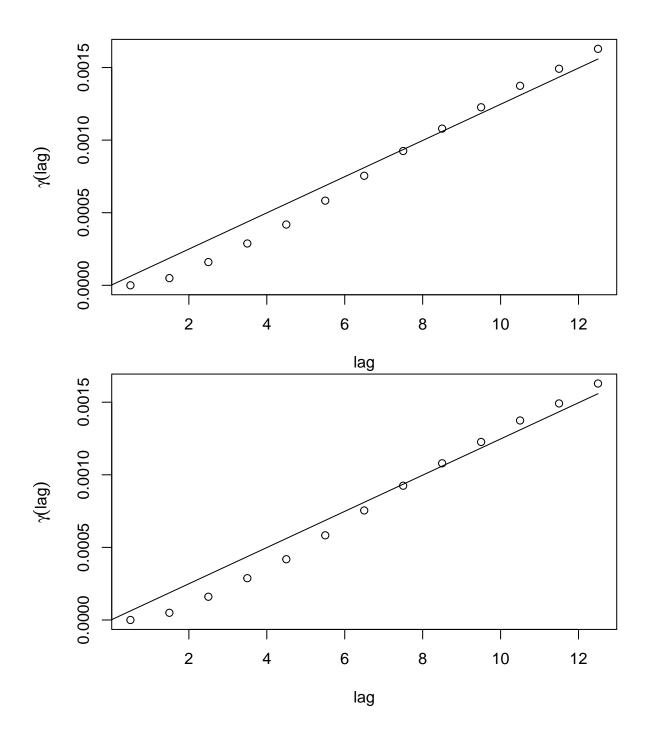


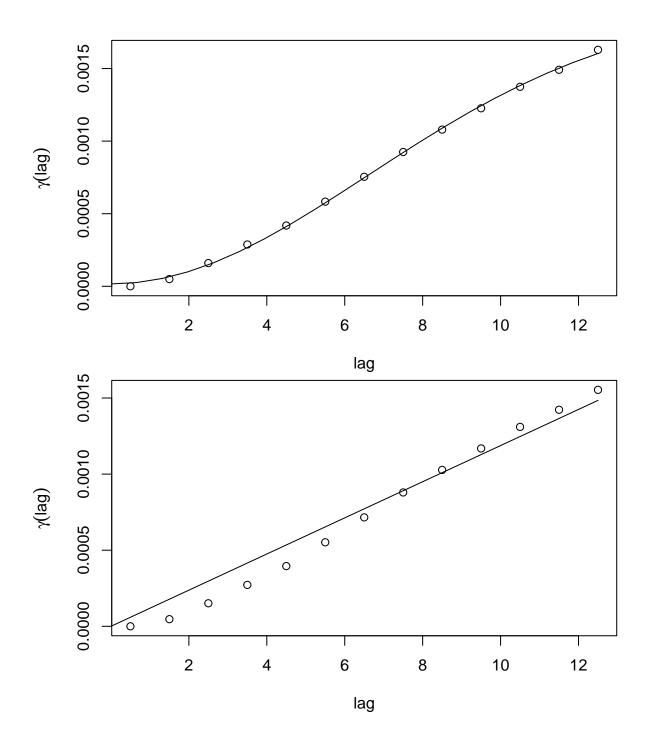


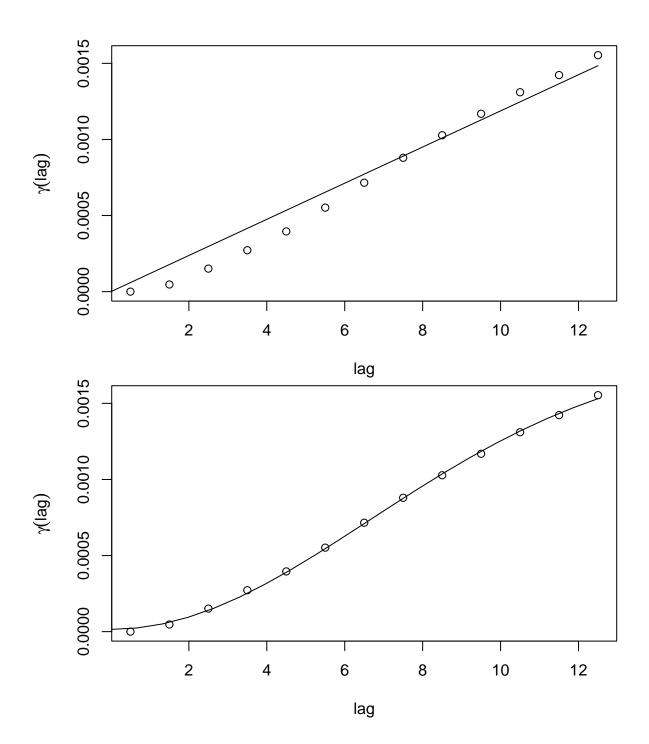


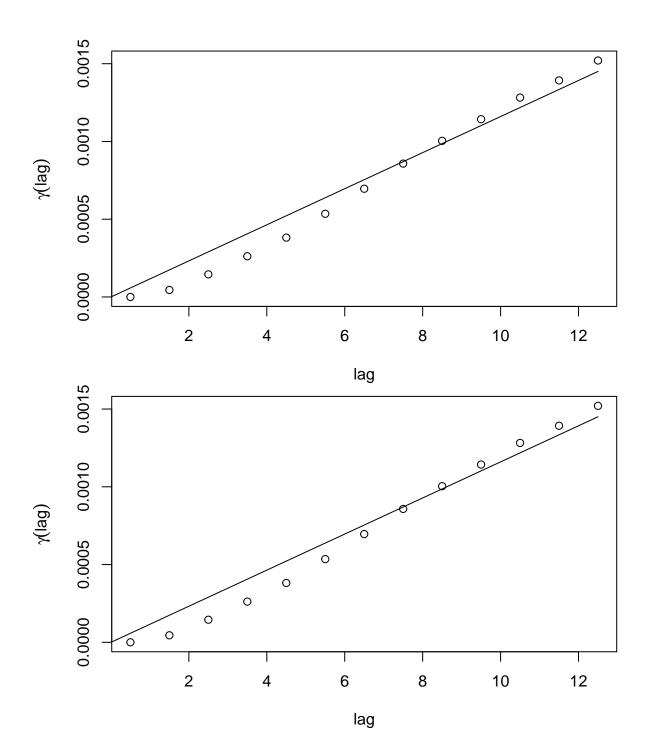


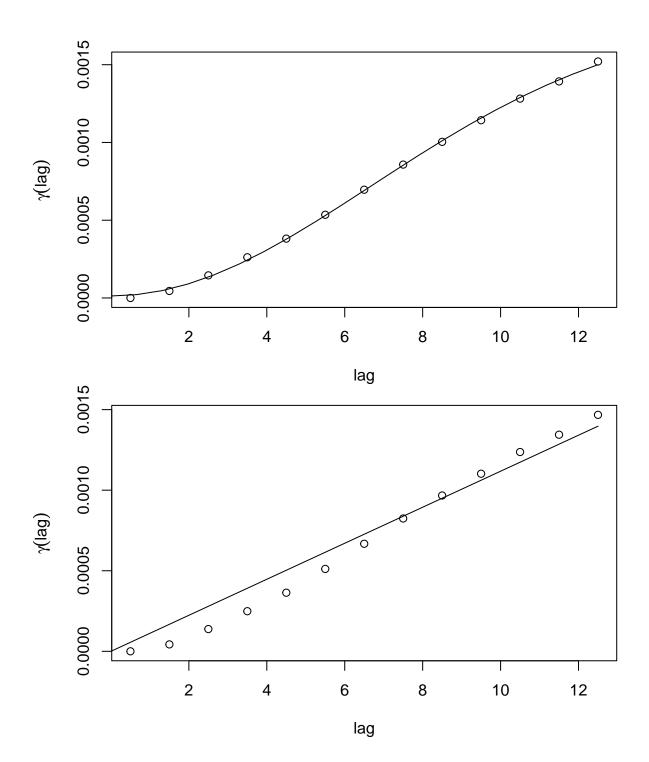


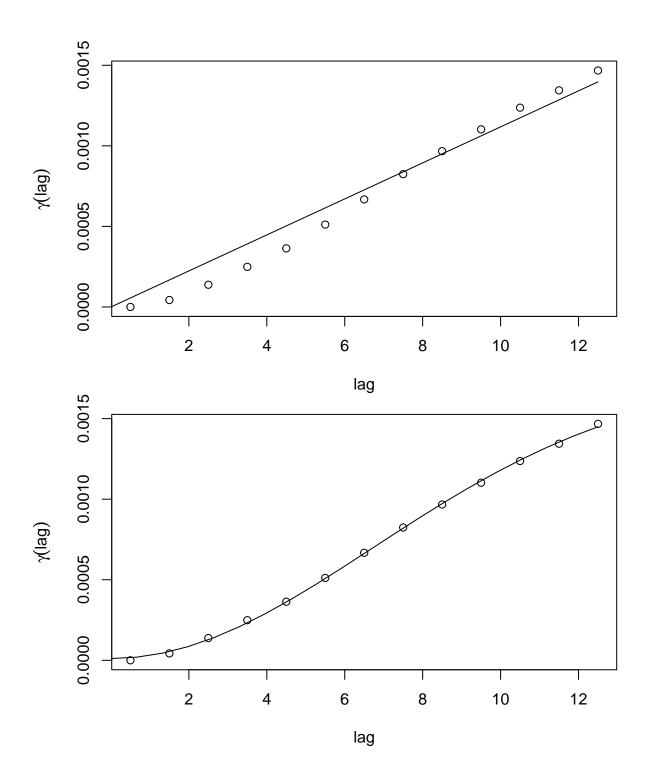


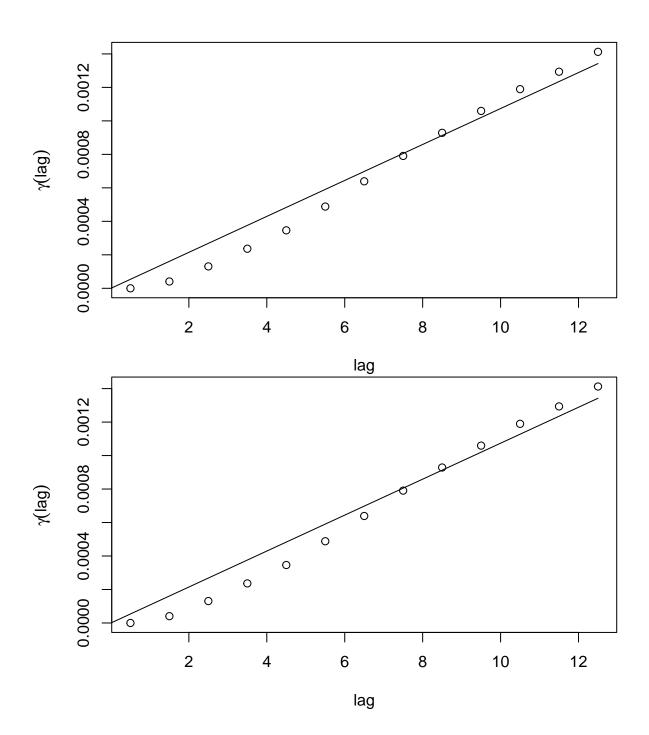


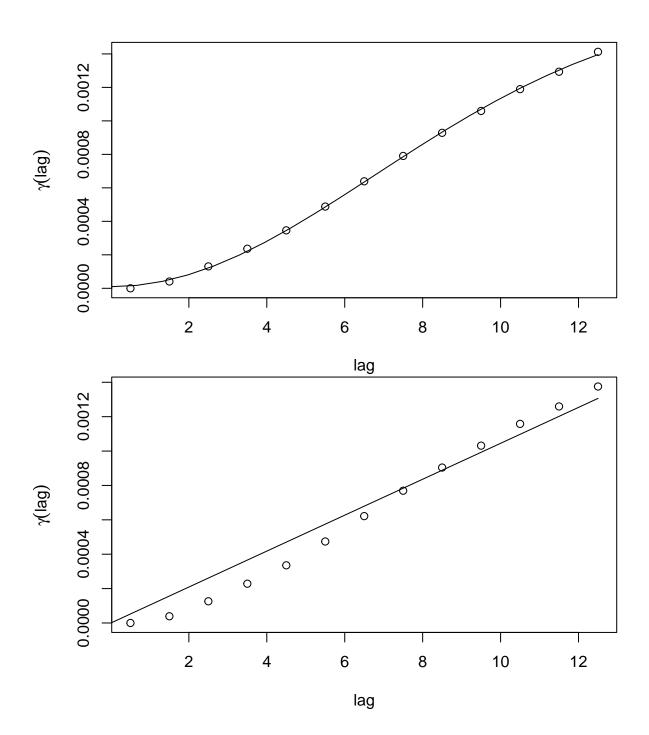


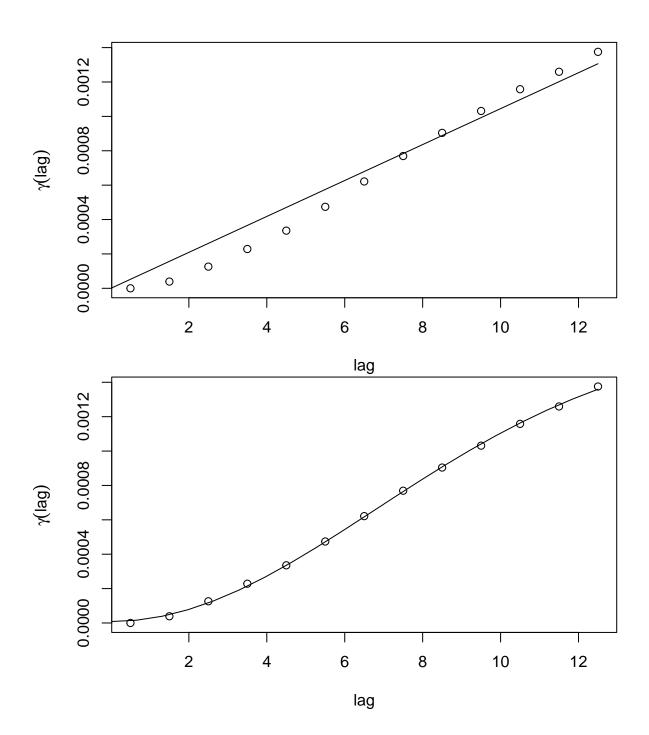


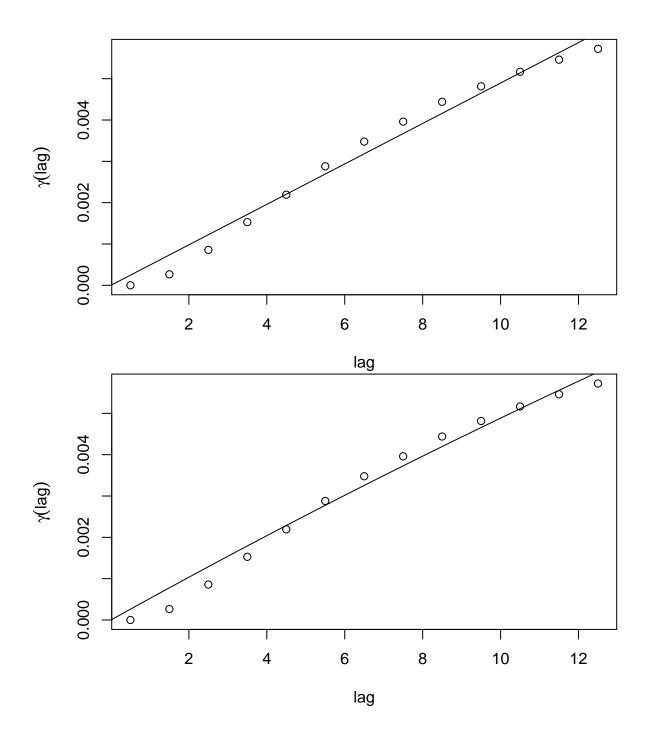


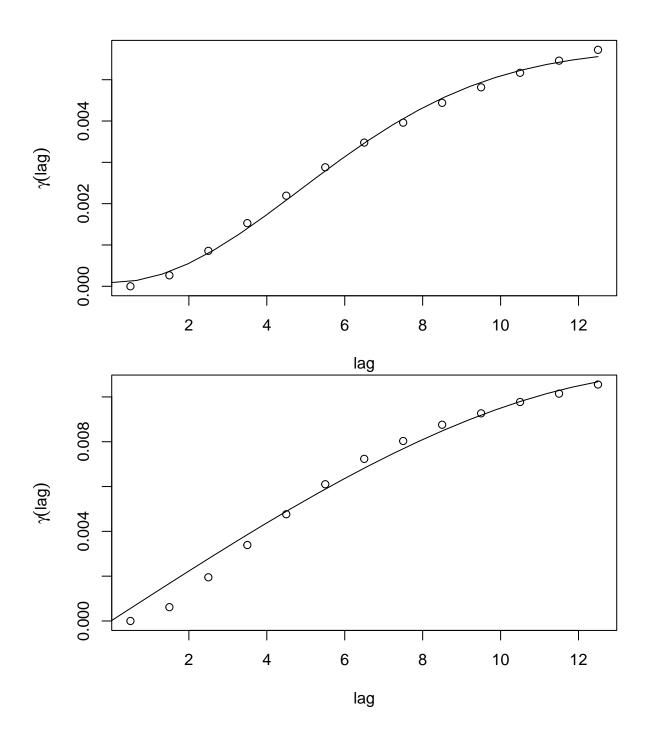


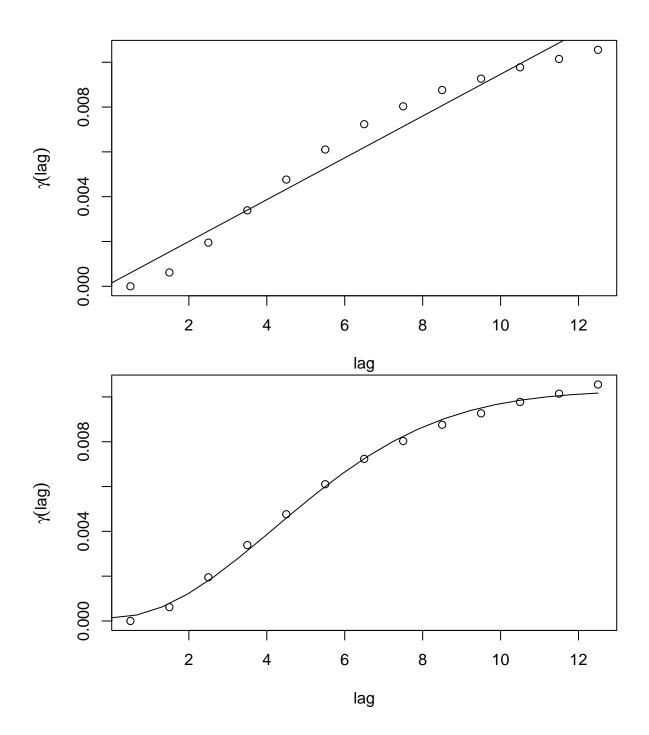


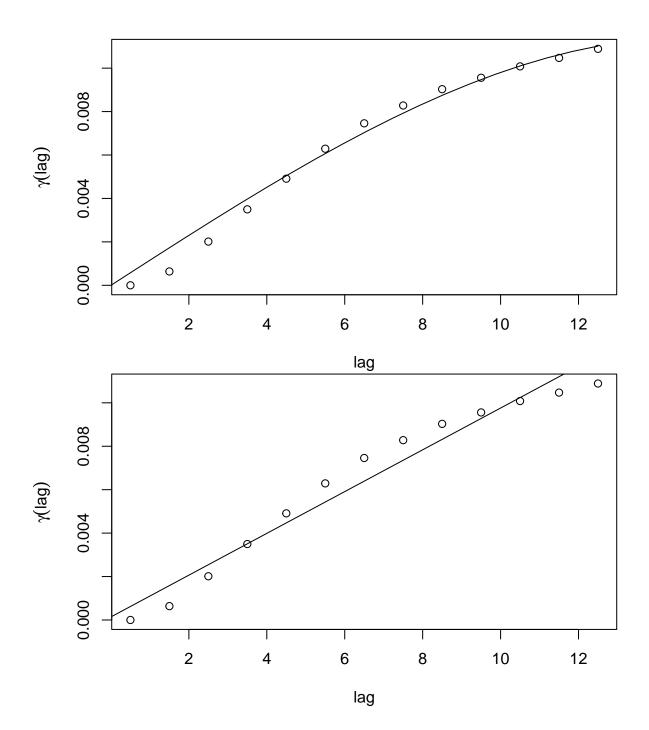


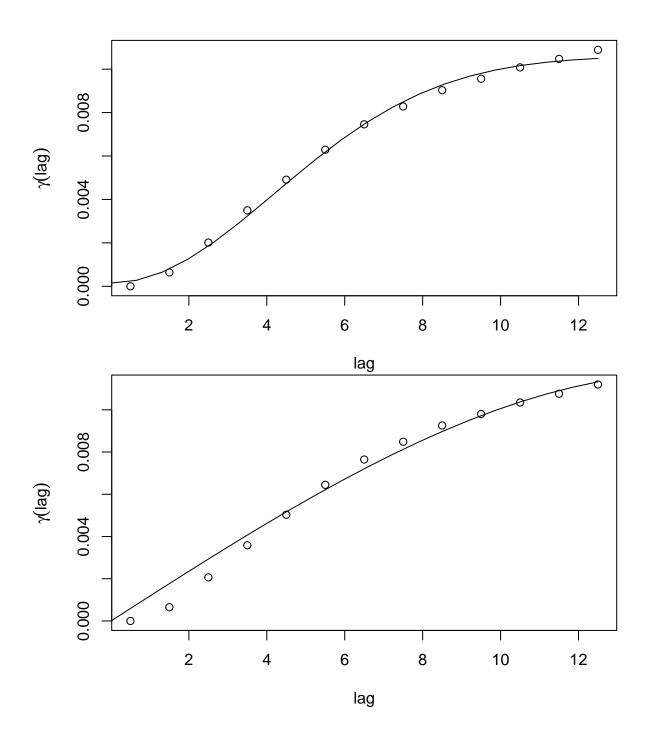


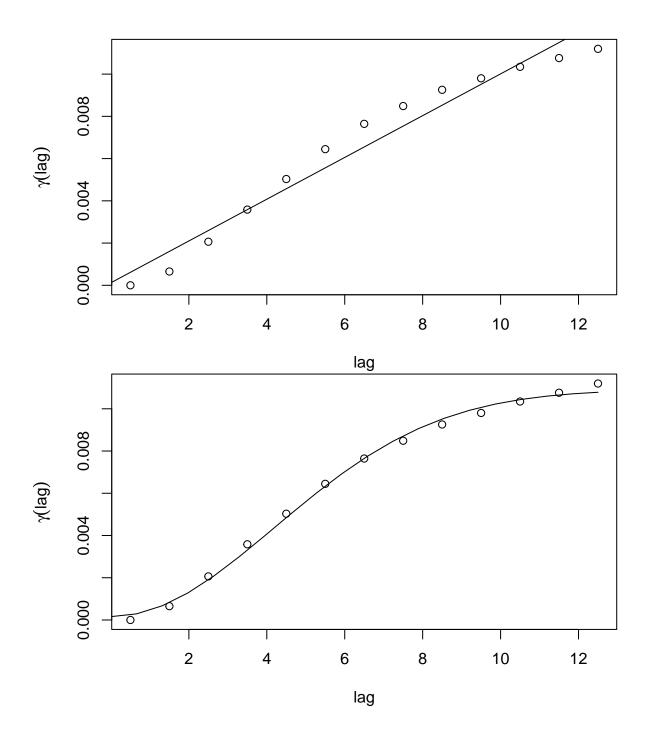


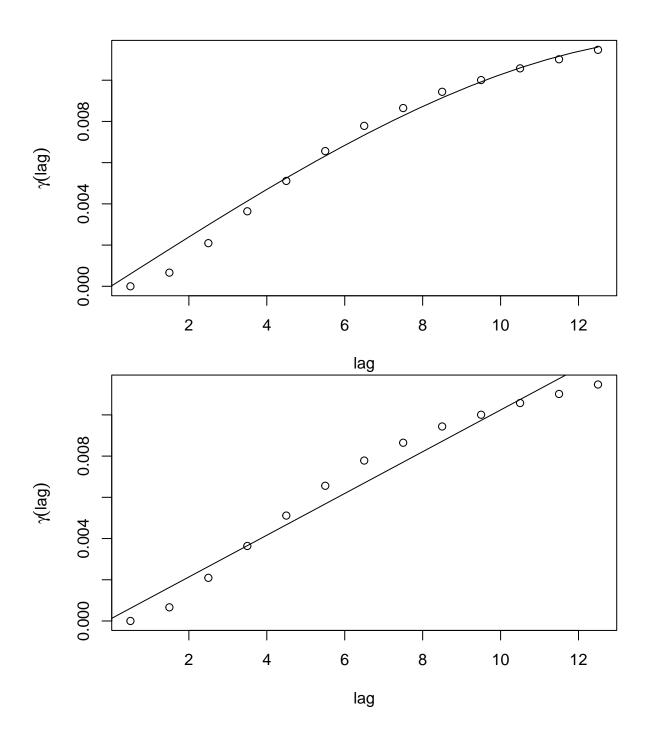


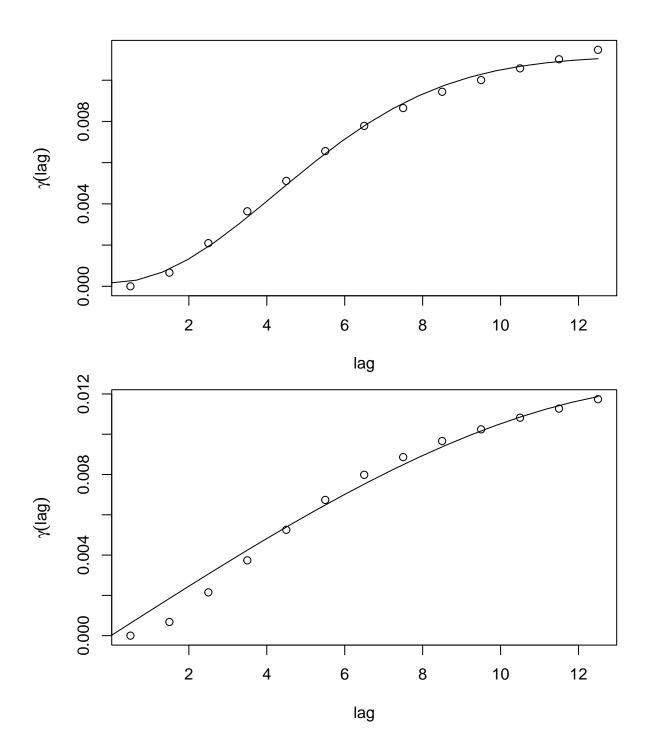


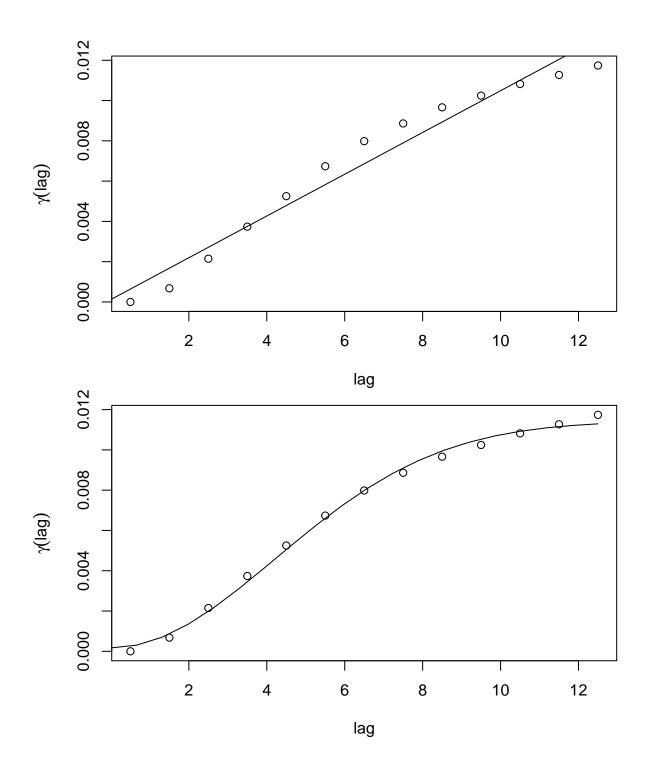


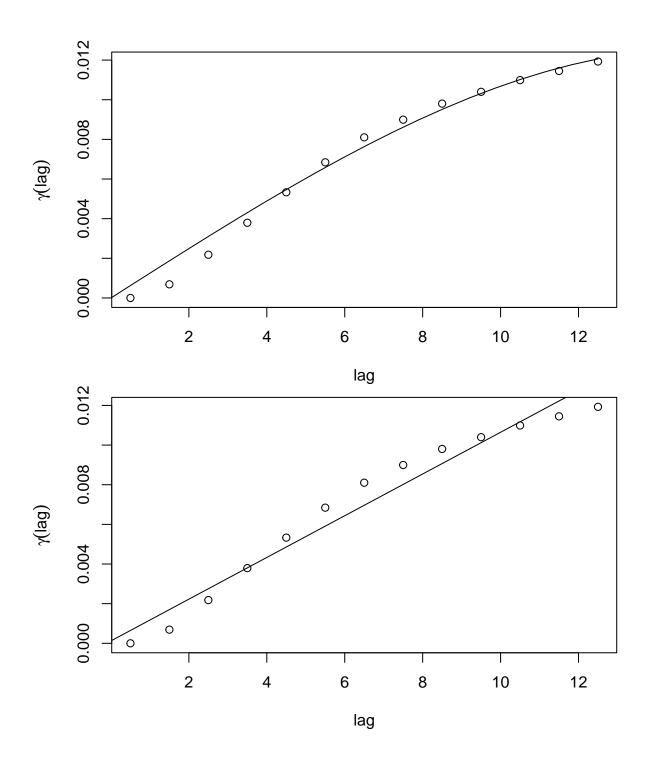


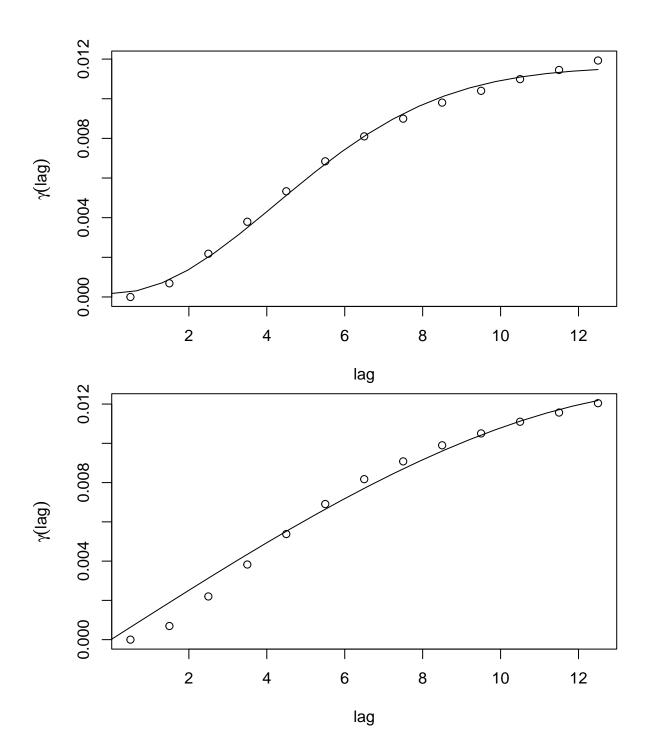


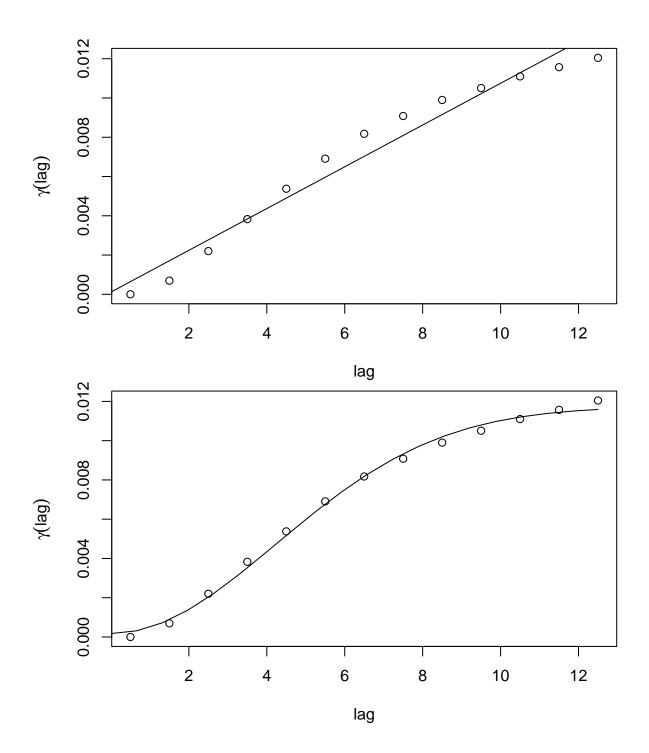


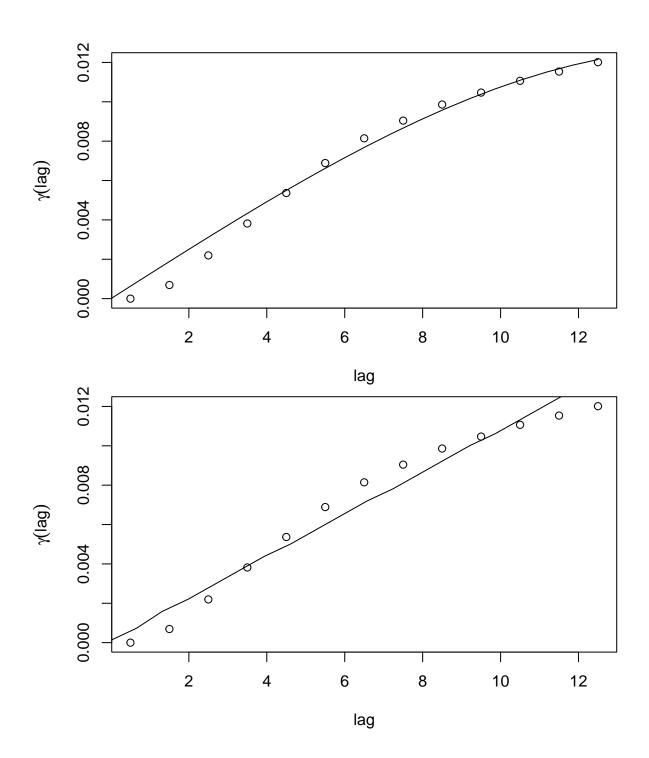


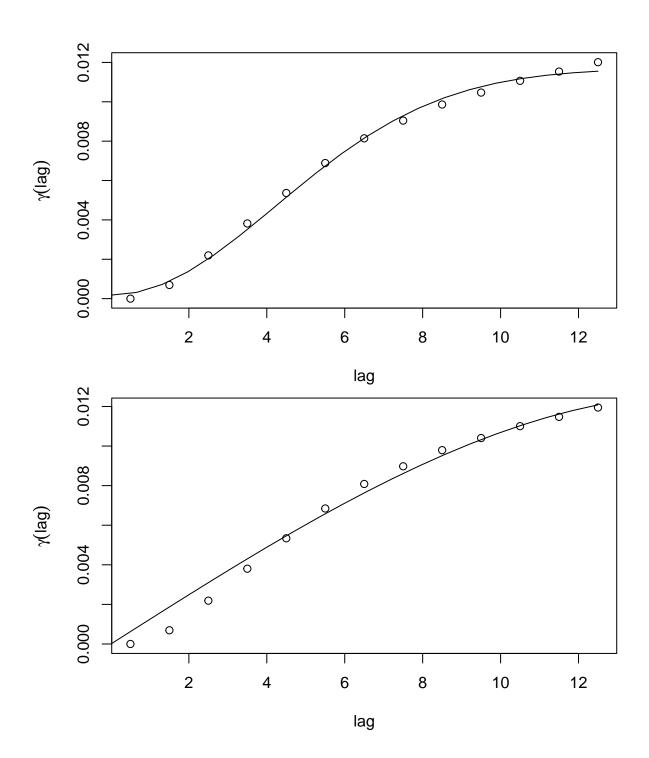


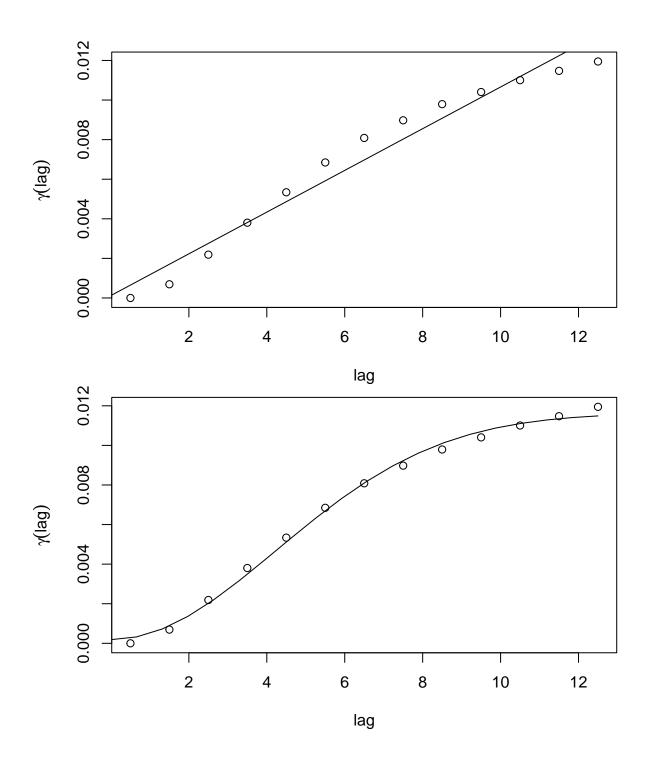


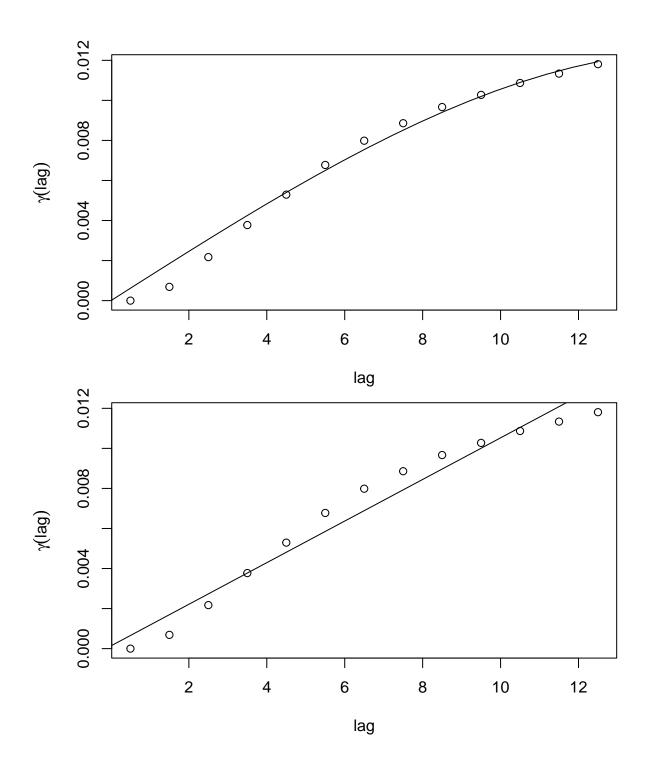


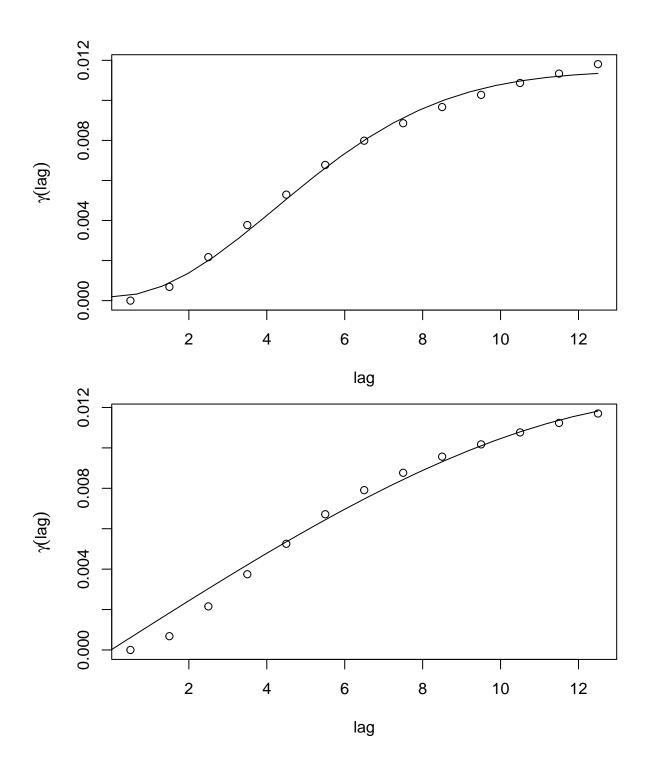


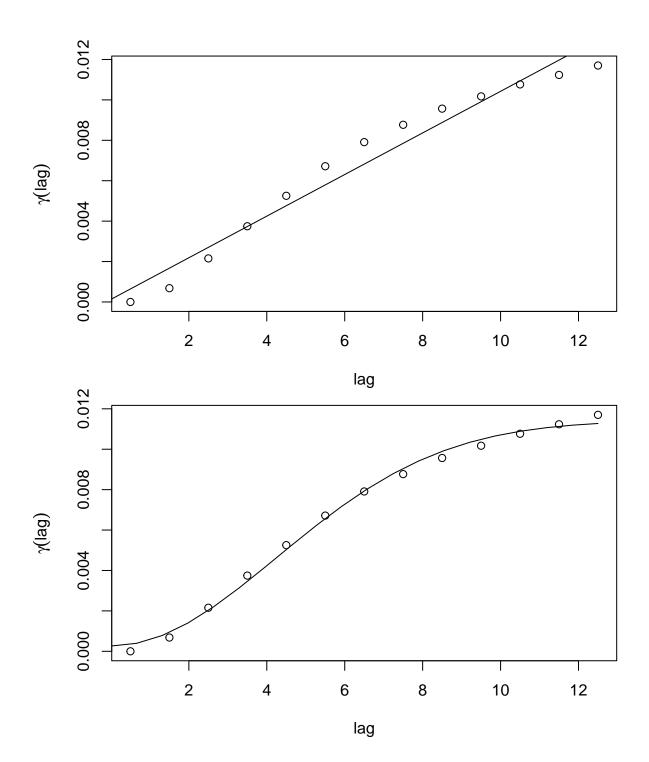


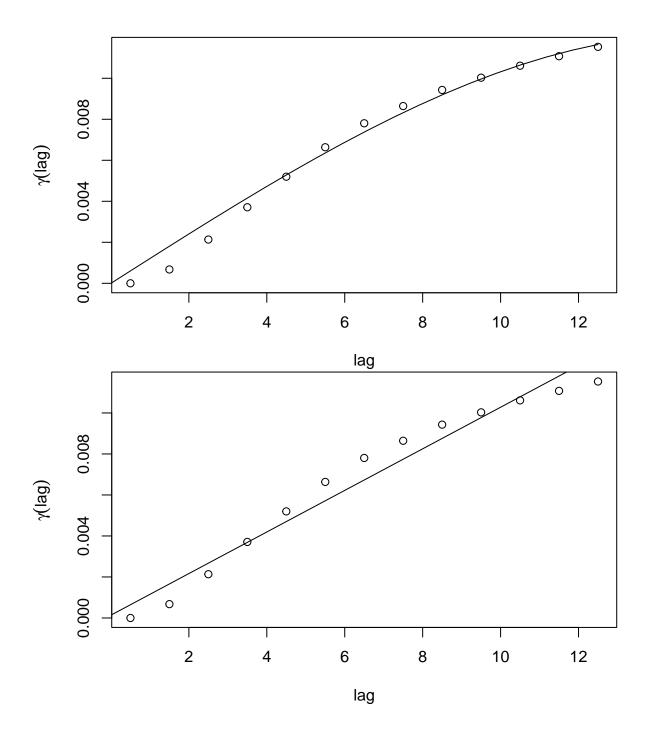


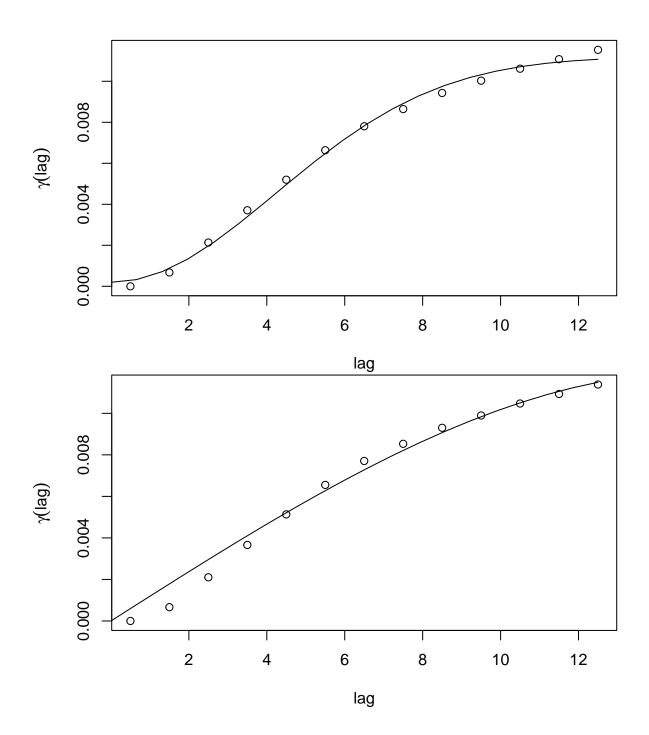


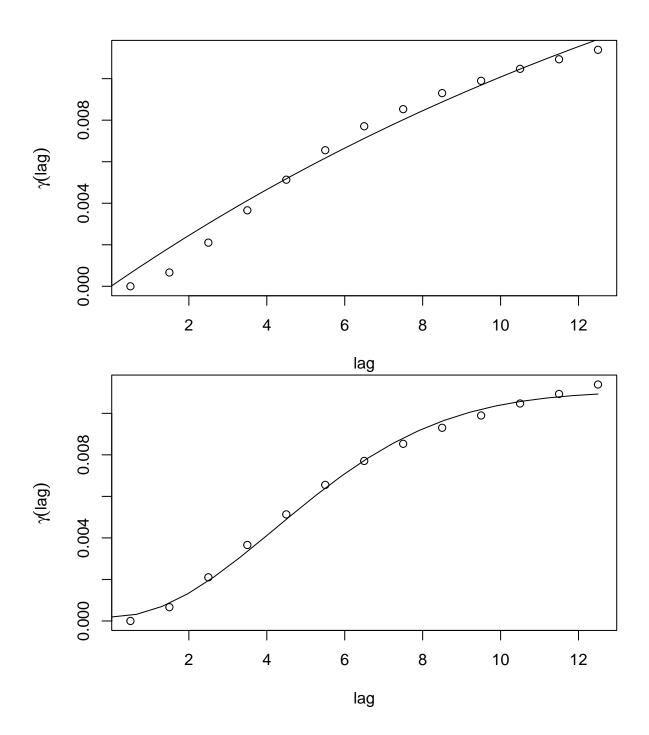


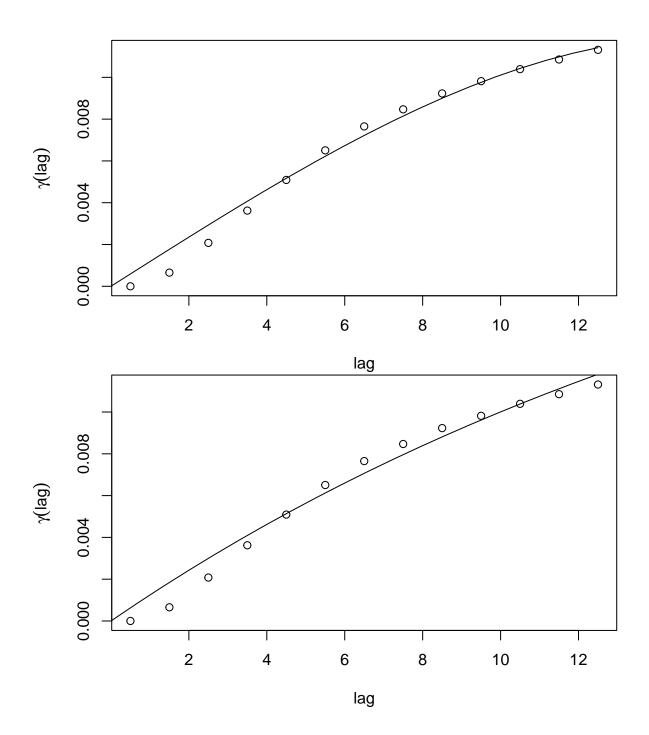


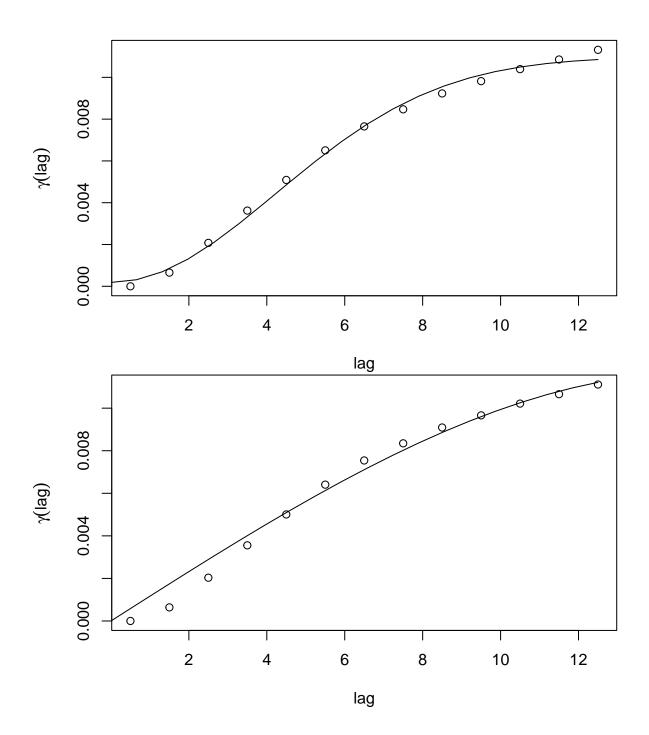


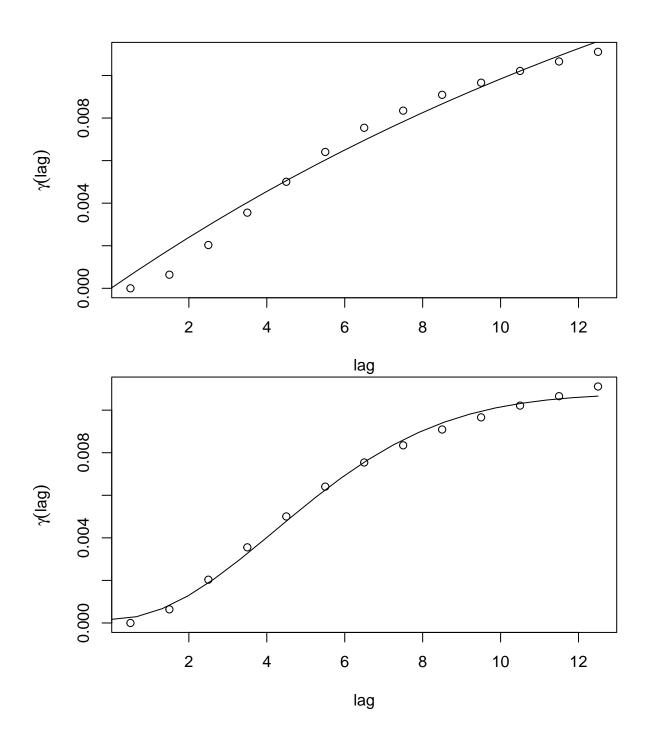


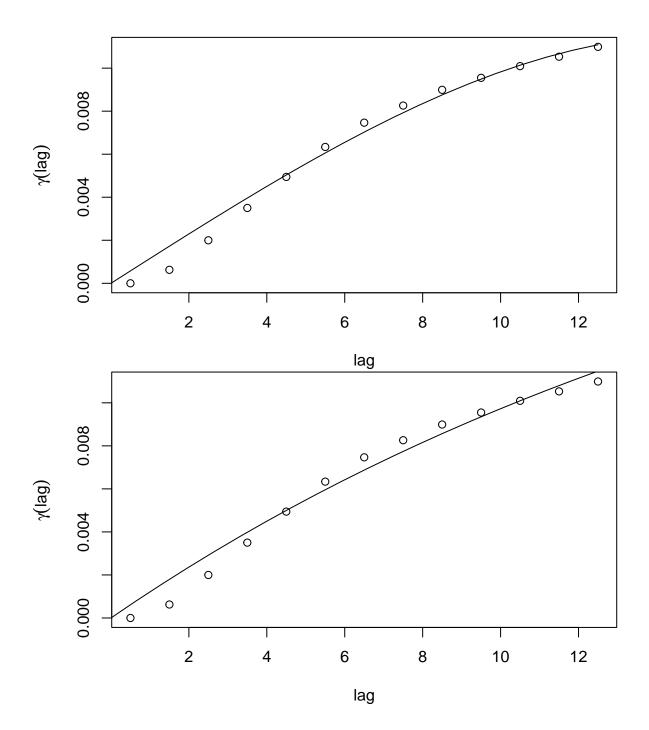


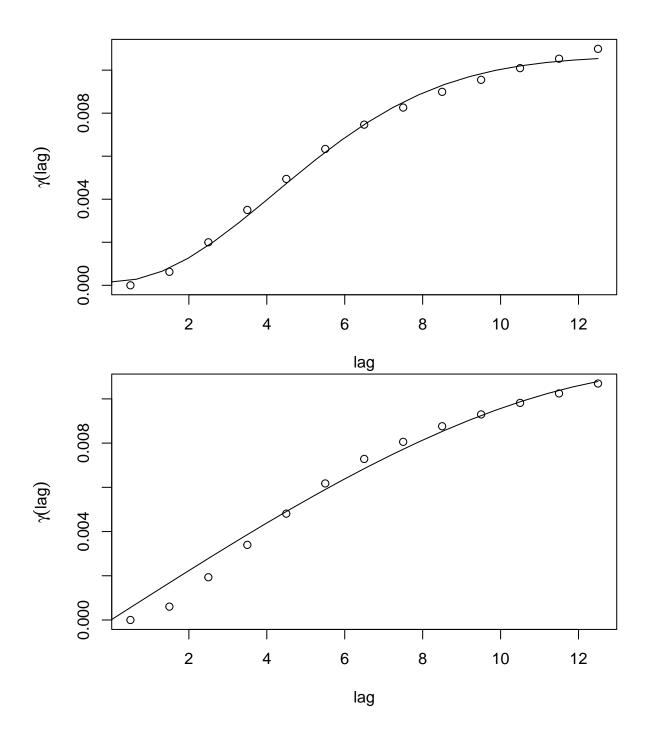


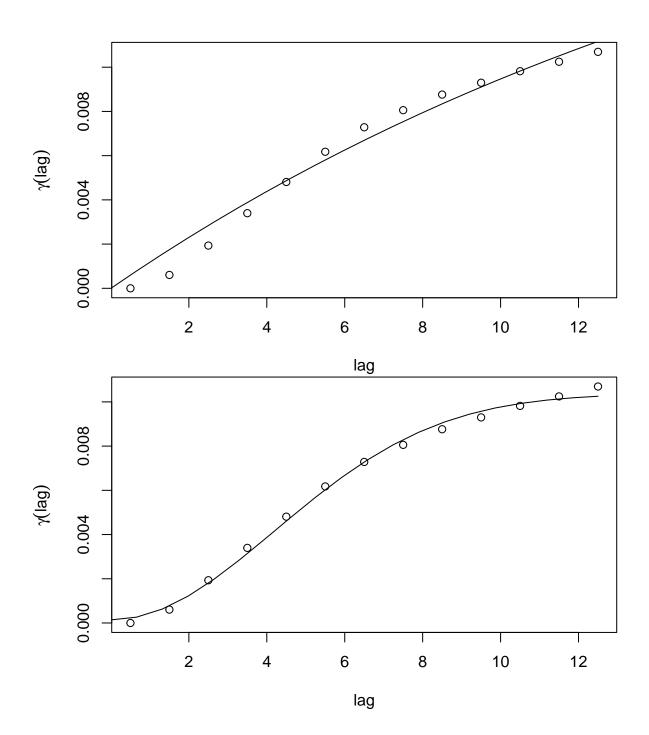


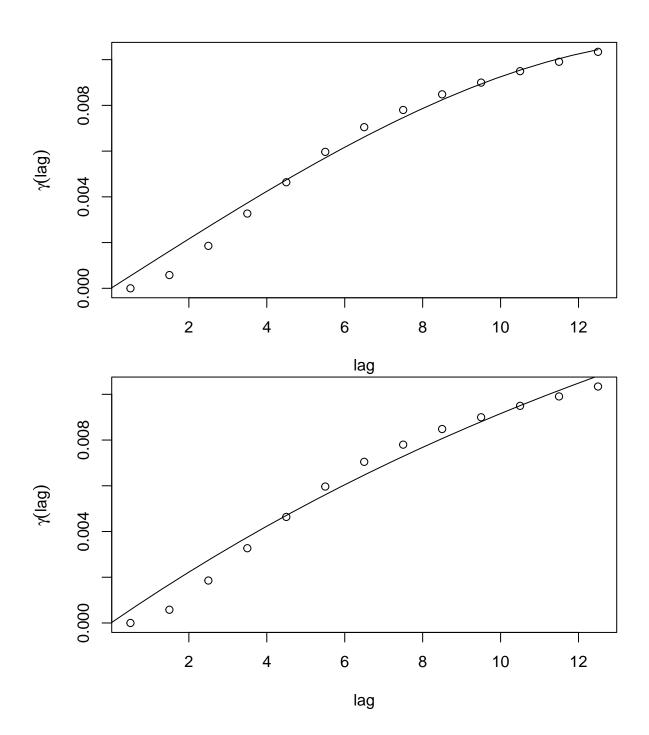


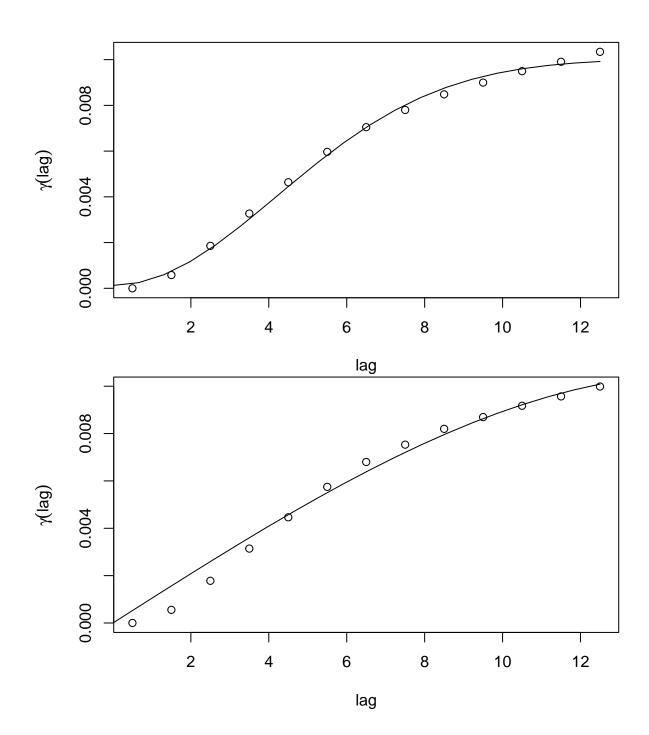


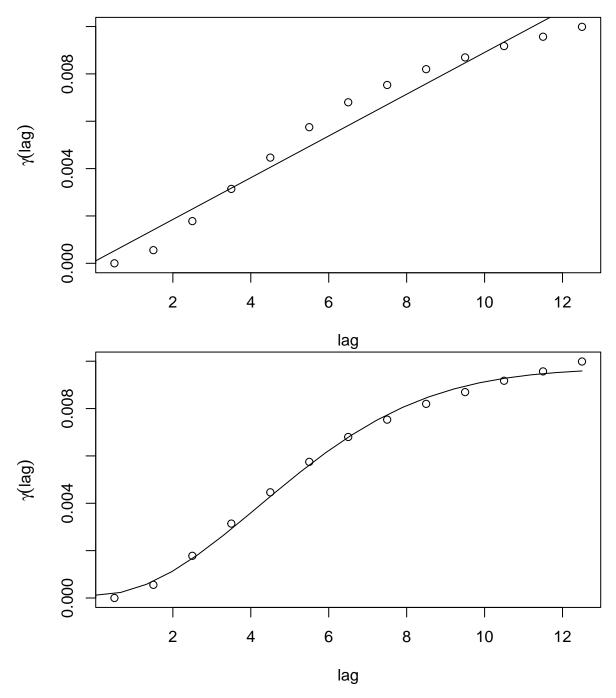












Based on visual assessment, the Gaussian model is the best fit to the relative abundance reconstructions. We can now use the parameters from this mdoel to make inference about the distance of spatial correlation and its strength.

Here are the distances after which grid cells are deemed independent (in number of grid cells, the "range" parameter):

Next, I will compute the omnidirectional variogram for each taxon at each time step. I do these at each time

step because I expect that the spatial correlation may change as a function of time.