

Modeling in-vitro functional assay of allosteric modulators

Biometrics and Reporting

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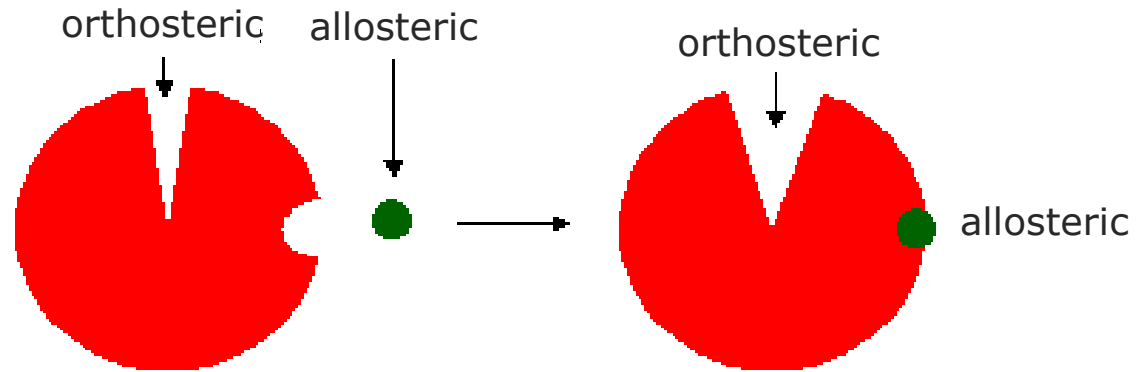


21.05.2013

What's an allosteric modulator?

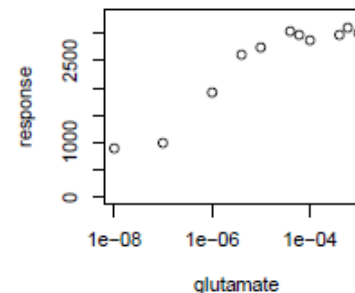
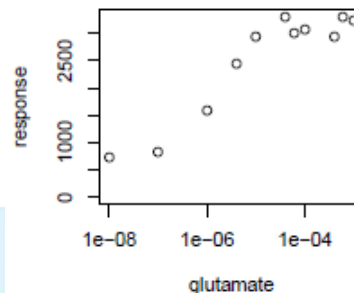
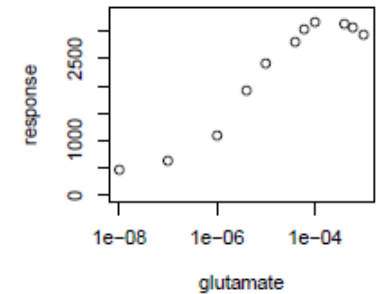
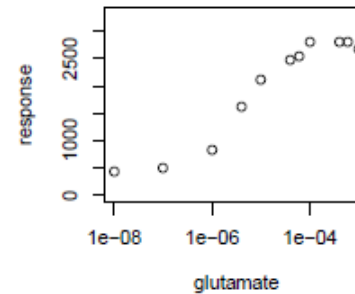
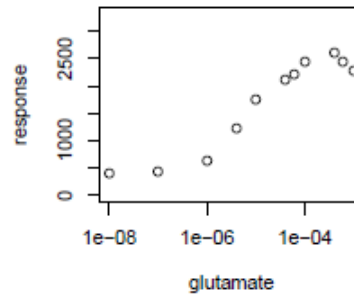
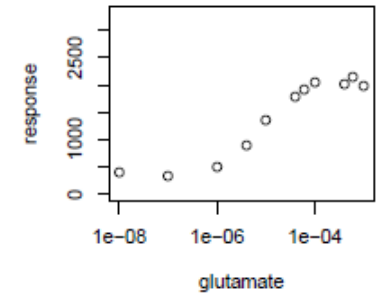
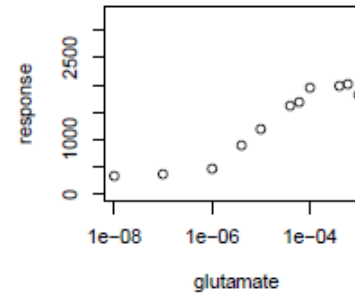
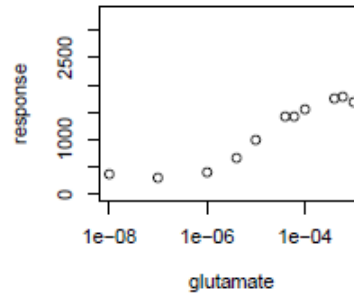
- does not bind to orthosteric site as full or partial agonists do
- has no or minimal agonist activity of their own
- increases the effects of natural neurotransmitters
- preserves the temporal and spatial integrity of the neurotransmission
- will only work in the presence of the (natural) agonist
- potentiates effect of a specific neurotransmitter ($EC_{50} \downarrow$, $E_{max} \uparrow$)
- Naively speaking: modifies the effects of a specific neurotransmitter by binding to a site in proximity of the neurotransmitter

What's an allosteric modulator?



Case Study: functional assay

Study the in-vitro co-administration of an orthosteric neurotransmitter and an allosteric modulator.



Reference compound

Methodology

Based on receptor theory, one can prove that the data ought to follow following equation:

$$y = E_0 + \frac{E_m (\tau_A [A] (K_B + \alpha\beta [B]) + \tau_B [B] K_A)^\eta}{([A] K_B + K_A K_B + [B] K_A + \alpha [A] [B])^\eta + (\tau_A [A] (K_B + \alpha\beta [B]) + \tau_B [B] K_A)^\eta}$$

Where:

- E_0 : background noise in the experiment
- $[A]$, $[B]$: orthosteric and allosteric compound concentrations, resp
- K_A , K_B : equilibrium dissociation constants
- α : cooperatively factor describing the allosteric effect on the other's binding affinity
- β : positive scaling factor
- τ_A , τ_B : ability to promote direct receptor activation (direct agonism)

Leach et al (2007), Melancon et al (2013), Gregory et al (2013)

Methodology

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Note the similarities with the traditional 4PL model:

$$y = E_0 + \frac{E_m C^\eta}{EC_{50}^\eta + C^\eta}$$

Methodology

Based on receptor theory, one can prove that the data ought to follow following equation:

$$y = E_0 + \frac{E_m (\tau_A [A] (K_B + \alpha\beta [B]) + \tau_B [B] K_A)^n}{([\text{A}] K_B + K_A K_B + [\text{B}] K_A + \alpha [\text{A}] [\text{B}])^n + (\tau_A [A] (K_B + \alpha\beta [B]) + \tau_B [B] K_A)^n}$$

Note the similarities with the traditional 4PL model:

$$y = E_0 + \frac{E_m C^n}{EC_{50}^n + C^n}$$

Benefit: if estimable, then directly interpretable parameters

Practical problems:

- more parameters to estimate
- starting values?
- is the data rich enough to estimate this?

Likelihood approach

How to obtain reasonable starting values?

- try and error: doomed to fail
- based on physiological knowledge if available
- likelihood profiling

Likelihood profiling:

```
proc nlmixed data=allosteric maxiter=0;
parms E0=... emax=... logtauortho =... to ... by ... logKortho=... to ... by
... logKallos=... to ... by ... logtauAllos=... to ... by ... logalpha=...
to ... by ... logbeta=... to ... by ... sig=... to ... by ... ;
...
ods output parameters=parm; run;
```

Explore where the optimum is located.

Time consuming!

Likelihood approach

Challenges:

- set hill function to 1
- too little information for e_{max} , so fixed (instable due to outliers?)
- despite likelihood profiling no convergence could be attained

“The operational model can be fitted to experimentally derived data to provide estimates of some, or all, of its parameters”

Melancon et al, 2013

Conclusion: often too little information is available from the in-vitro experiment for the different parameters and the frequentist approach fails.

Abandon the idea and throw away the data?

Bayes approach: use of priors

“too little information for e_{max} , so fixed”

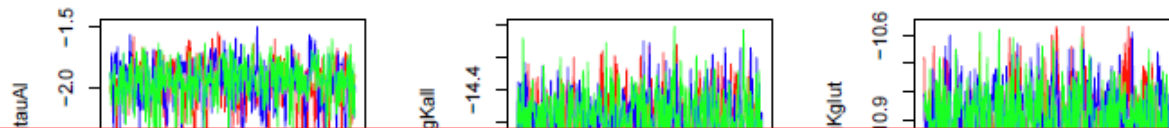
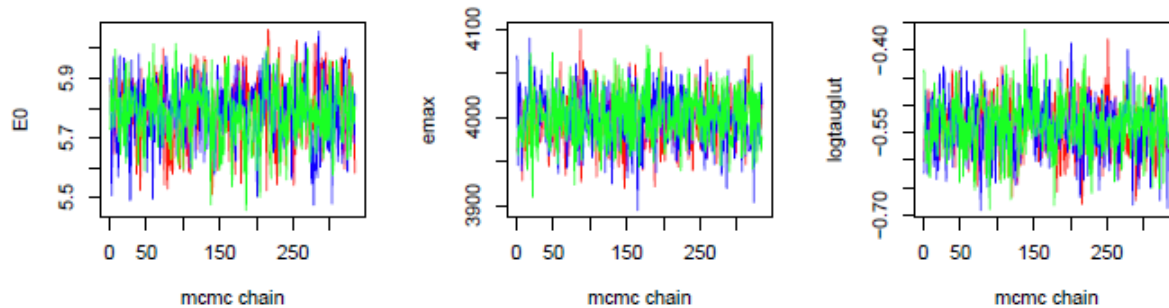
Alternative solution: give some reasonable feedback to the model, while allowing some flexibility with a weakly informative prior (“the e_{max} is within a specific range”).

Incorporate information from literature, likelihood profiling, etc regarding the orthosteric and allosteric compound (eg K_A) using (weakly) informative priors.

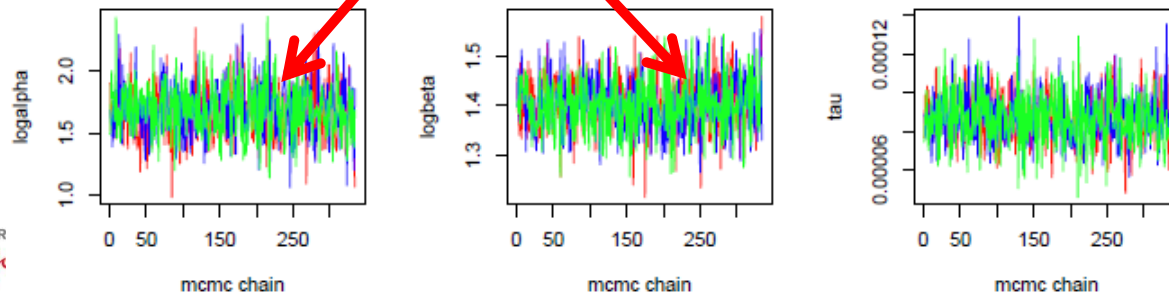
Next steps:

- check bayesian convergence, mcmc chains
- check model fit
- sensitivity check to the priors

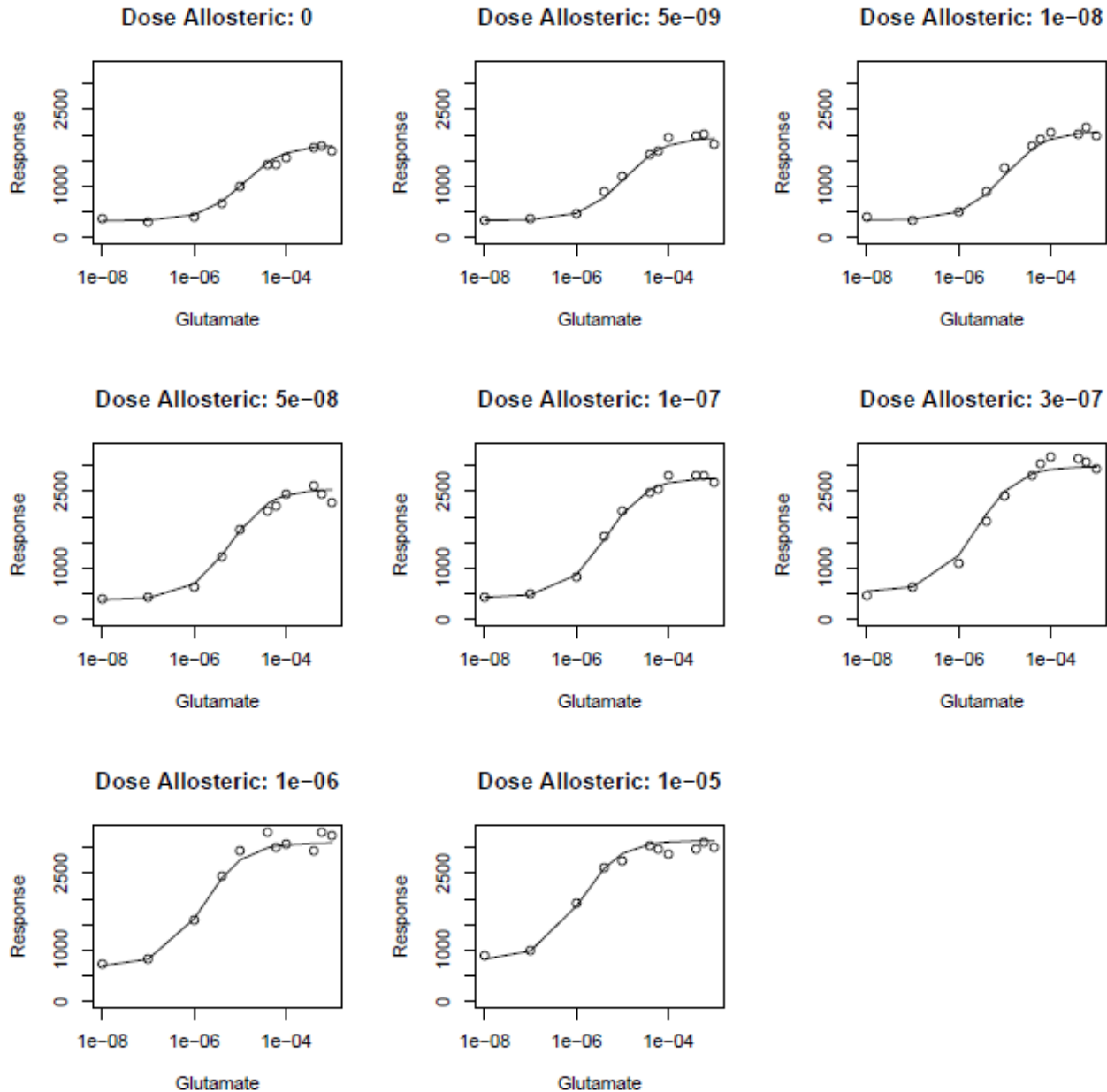
Bayes approach: mcmc chains



α and β are "more difficult" to estimate, stronger correlations in chains, take prior emax more informative to resolve

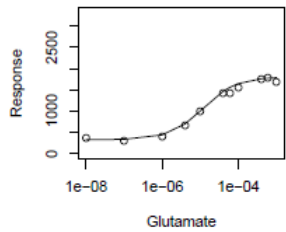


Bayes approach: model prediction

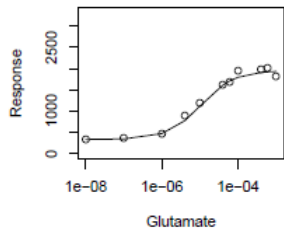


Bayes approach sensitivity analysis: priors and initial values

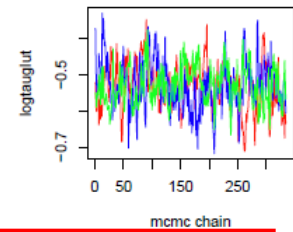
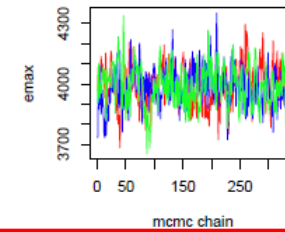
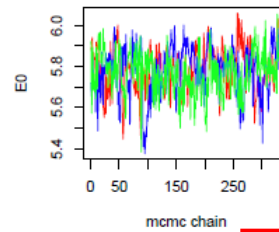
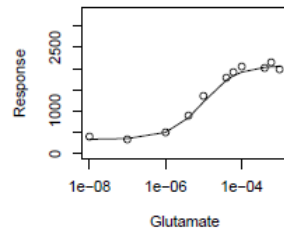
Dose Allosteric: 0



Dose Allosteric: 5e-09

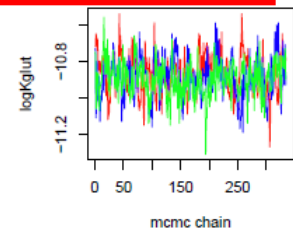
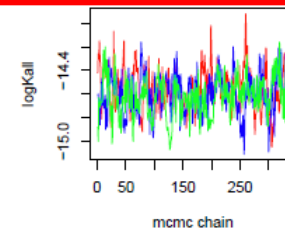
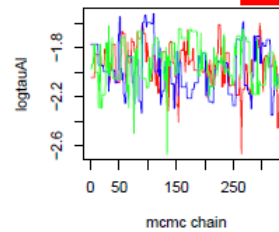
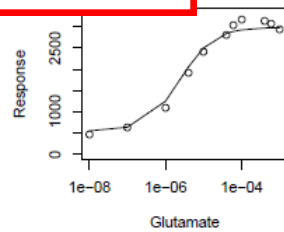
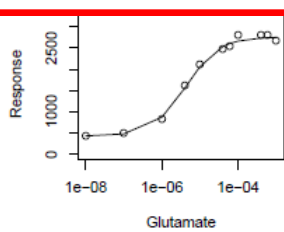
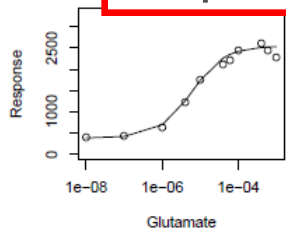


Dose Allosteric: 1e-08

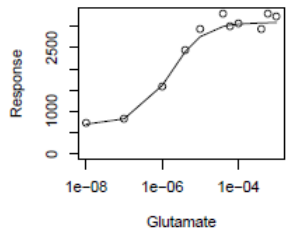


Dose Allosteric: 3e-07

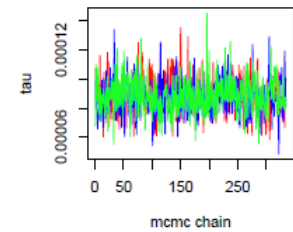
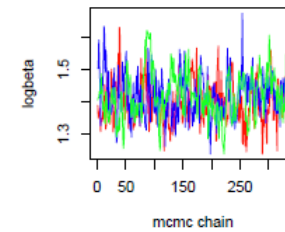
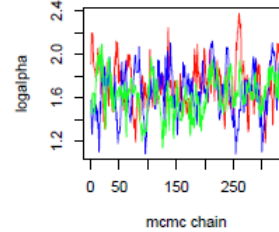
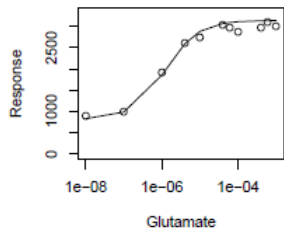
Impact chains: limited



Dose Allosteric: 1e-06



Dose Allosteric: 1e-05



Parameter estimates

	Est. (95%CI)
log(E0)	5.79 (5.57 ; 5.97)
E _{max}	3999 (3941 ; 4057)
Log(K _{all})	-14.6 (-14.9 ; -14.2)
Log(K _{glut})	-10.9 (-11.1 ; -10.7)
Log(alpha)	1.69 (1.28 ; 2.13)
Log(beta)	1.40 (1.30 ; 1.52)
Log(tau _{AI})	-1.97 (-2.40 ; -1.66)
Log(tau _{glut})	-0.539 (-0.642 ; -0.438)

Conclusion

Giving reasonable feedback to the model using (weakly) informative priors allows answering the scientific question. It is a perfect intermediate solution between allowing parameters to vary freely and fixing them to a specific value.

The mcmc chains suggested for which parameters (E_{max} , α , β) the data was less appropriate.

Although helping the model to converge, the predictions and parameter estimates are reasonable, but some caution remains.

Acknowledgements

- Filip De Ridder
- Eef Hoeben
- Jan Serroyen
- An Vermeulen

References

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