

Introduction:

Daily micturition frequency is a key endpoint for assessing overactive bladder disease activity.

Micturitions are count data and are commonly modeled assuming the Poisson distribution.

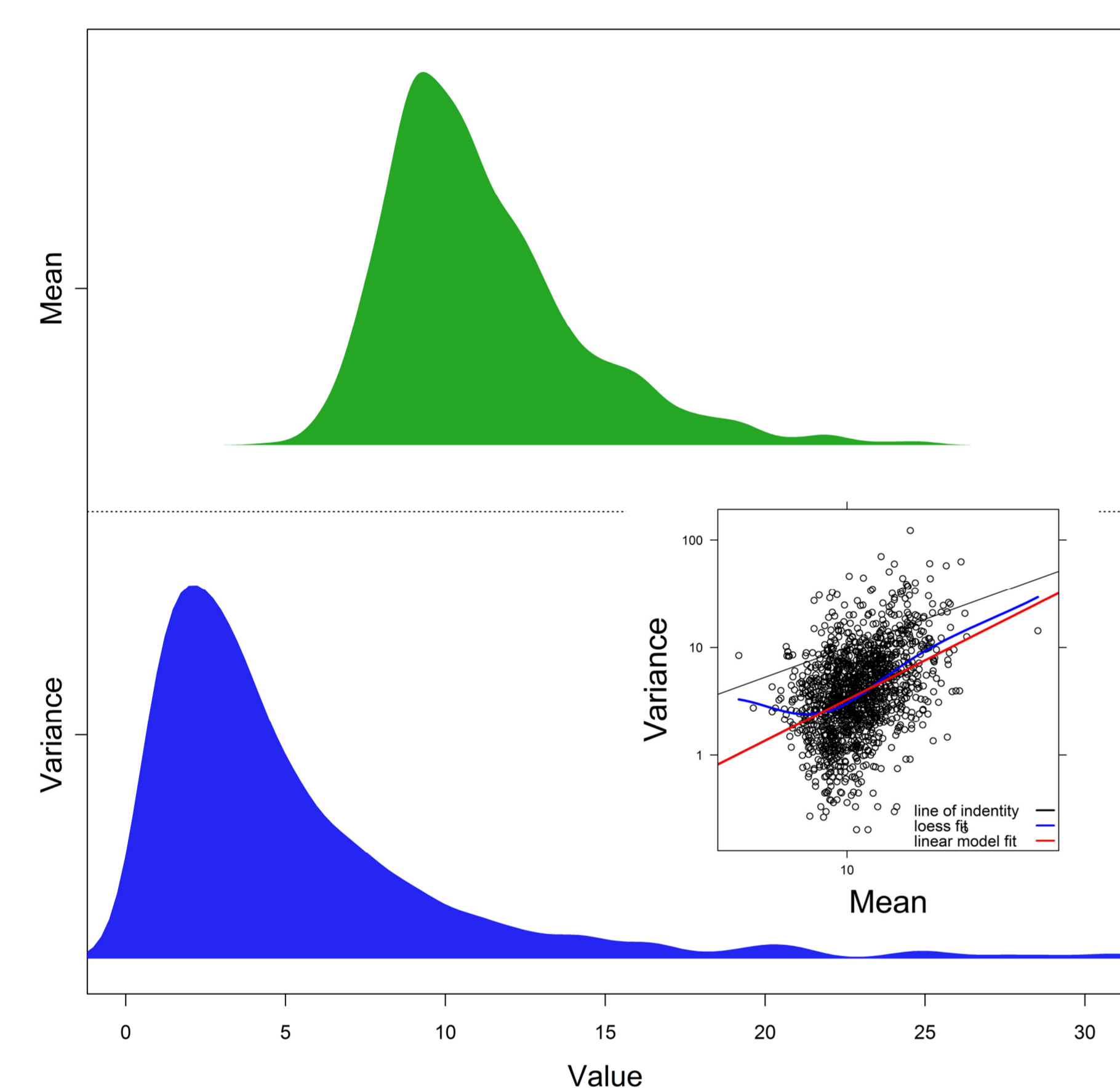
Although Poisson process assumes equi-dispersion, which means that mean and variance are the same, observed within-individual variance is consistently lower than within-individual mean micturition frequency (Figure 1).

Being encouraged by a recent study addressing under dispersion in Likert pain rating scales [3], we wanted to evaluate if the generalized Poisson (GP) that flexibly describes under and over dispersion describes micturition counts better than the standard Poisson (PS) distribution.

Objectives:

To evaluate if the generalized Poisson describes micturition counts better than the Poisson distribution.

Figure 1. Distribution of individual means and variances



Equations and examples of dispersion

$$P(Y = n) = \frac{e^{-\lambda} \cdot \lambda^n}{n!} \quad \text{Poisson}$$

$$P(Y = n) = \frac{e^{-\lambda-n\delta} \cdot \lambda(\lambda + n\delta)^{n-1}}{n!} \quad \text{Generalized Poisson}$$

Parameters in green are estimated

Thus, a Poisson distribution is a generalized Poisson distribution with dispersion factor $\delta = 0$. Mean count and variance are given by:

$$\bar{x} = \frac{\lambda}{1 - \delta} \quad \text{var}(x) = \frac{\lambda}{(1 - \delta)^3}$$

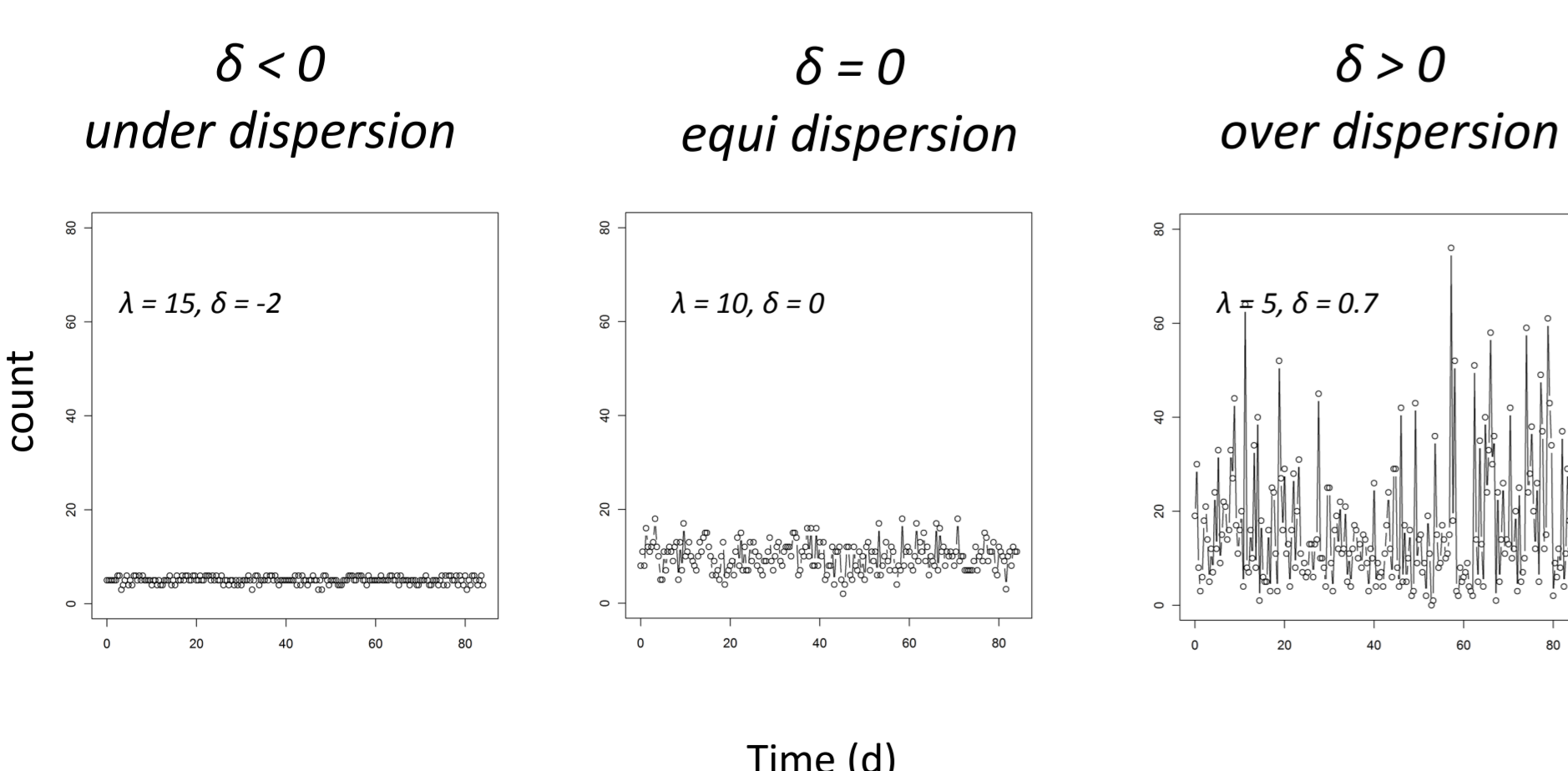


Figure 2. VPC of Poisson model

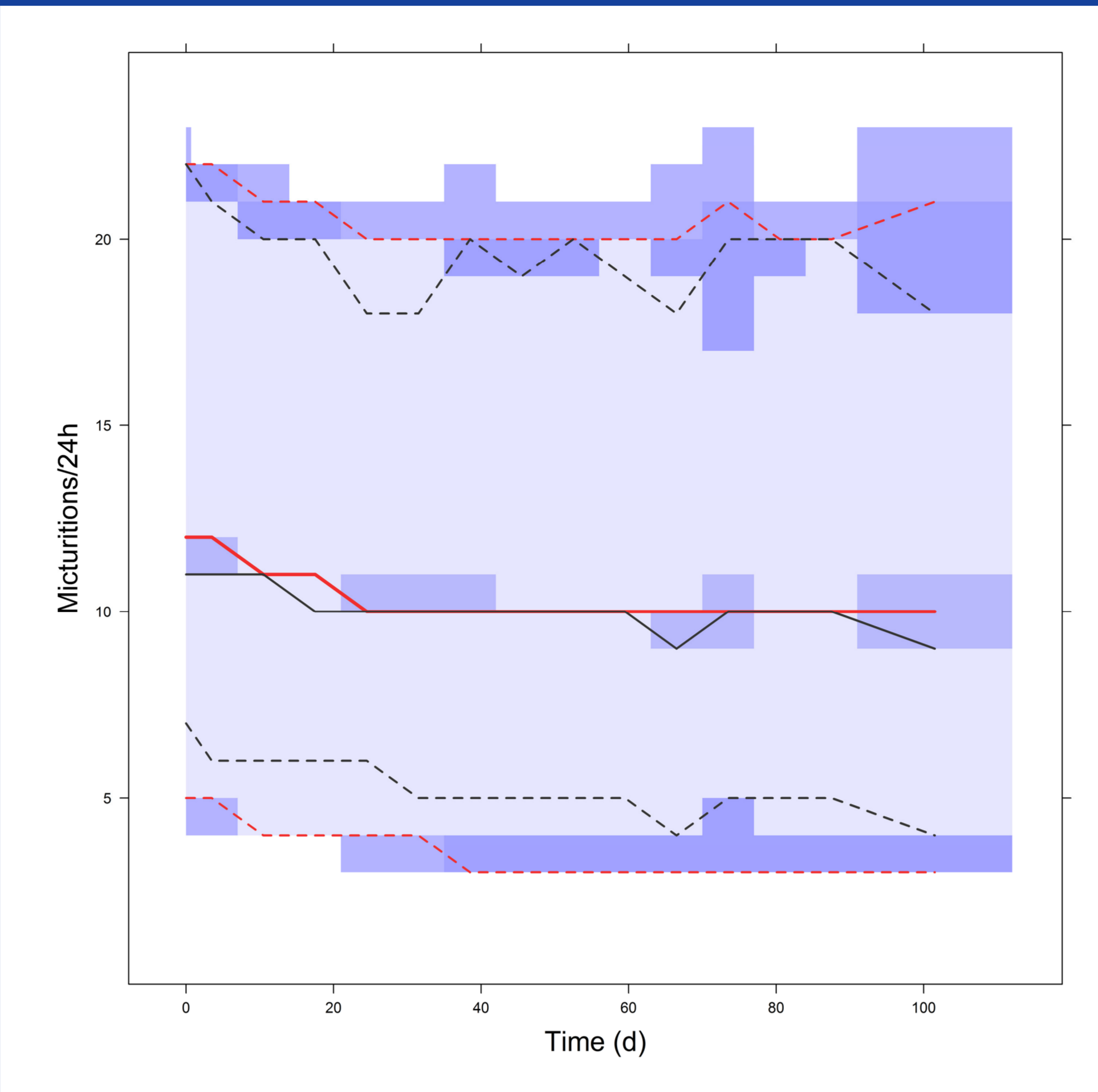
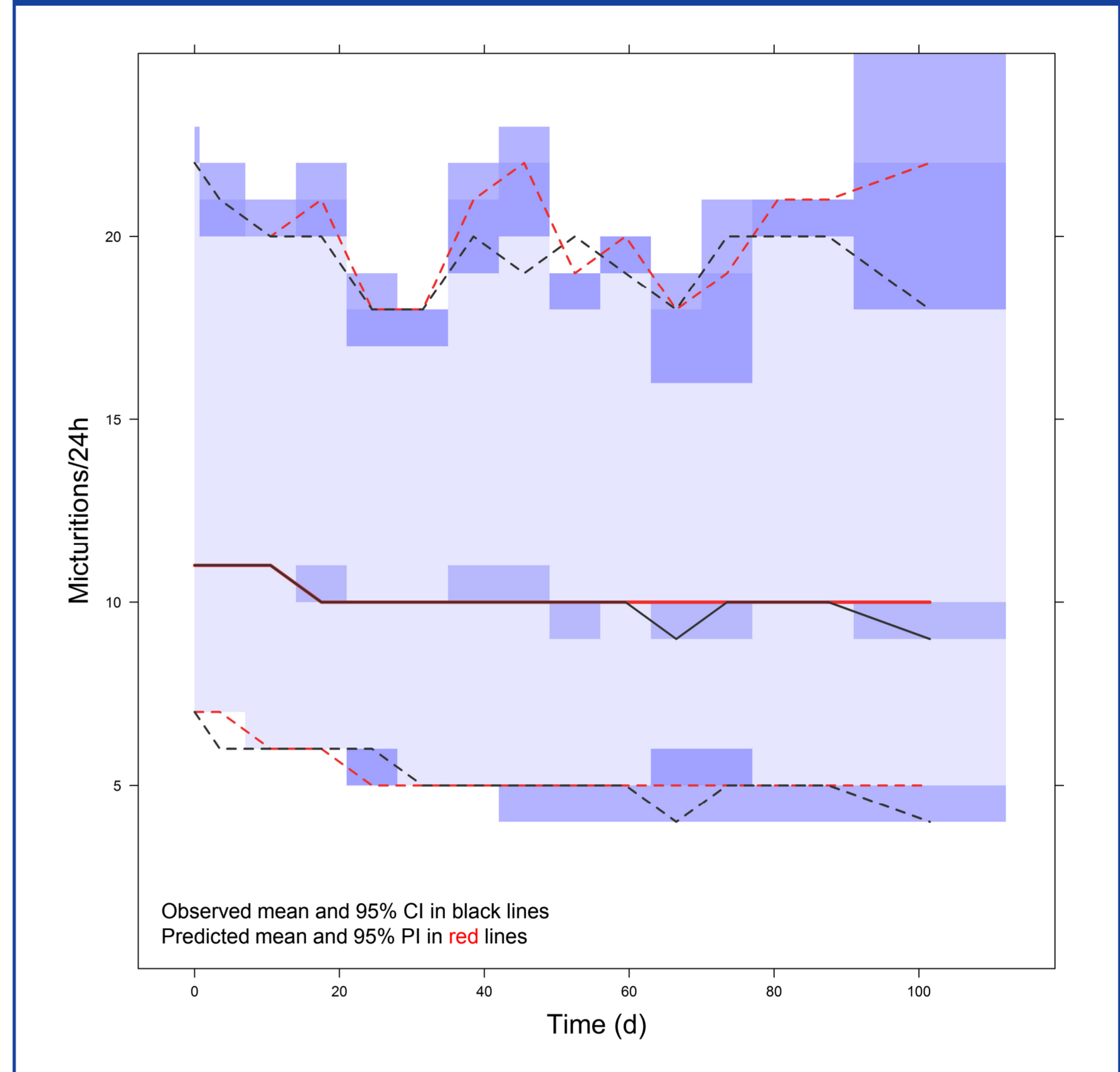


Figure 3. VPC of Generalized Poisson model



Methods:

Data

Placebo micturition count data from 1480 patients participating in 7 studies were used.

Model

Micturition counts (mict) were modeled as follows:

$$mict_{i,t} = mict_{base} \cdot (1 - Eff \cdot (1 - e^{-\theta_4 \cdot t}))$$

$$mict_{base} = \theta_1 \cdot e^{\eta_1}$$

$$Eff = \frac{e^{\theta}}{1 + e^{\theta}} + \eta_3, \quad 0.01 \leq Eff \leq 0.99$$

$$mict_{i,t} \sim genPoisson(\lambda, \delta)$$

$$\delta = \theta_2 \cdot e^{\eta_2}$$

Where $mict_{i,t}$ is the micturition count in the i^{th} individual at time t . Lognormal between subject variability (BSV) was assumed on λ (PS) or λ_1 (GP) and additive BSV was assumed on Eff (PS, GP).

FOCE LAPLACE did not prove stable for the generalized Poisson model, thus resampling tools were used: SAEM followed by MCMC BAYES. The SAEM was merely aiming to get some priors, hence 50 burn-in and 50 sampling iterations were requested:

```
$EST METH=SAEM LAPLACE -2LL NBURN=50 NITER=50 PRINT=1
```

The MCM Bayesian option was used for the final regression:

```
$EST METH=BAYES CTYPE=3 NITER=2000 NBURN=2000 PRINT=50 FILE=run1.bay
```

In the special case of $\delta=0$, the GP model collapses to a PS. Since the likelihood functions for PS and GP are exactly the same the OFV of these models can be compared.

Models were compared by :

- Objective Function Value (OFV)
- ability to capture mean trends and observed variability using Visual Predictive Check (VPC) using the vpc tool from Perl-speaks-NONMEM
- precision of parameter estimates.

Results:

- The GP model was significantly better than the PS model as compared by the lower mean OFV (90382 vs. 73020) which is a >17,362 point drop (~12 points per individual).
- The mean trend was better captured with the GP model.
- The VPC (Figure 2 and 3) showed that the PS model under predicted the 5th and over predicted the 95th confidence interval, while the GP model captured them remarkably well.
- Parameter estimates $mict_{base}$ and k (rate of effect onset) were 41 and 46% more precise for the GP model. The parameter for placebo effect size was 25% less precise.

Parameter	Poisson			Generalized Poisson		
	Estimate	SE	CV (%)	Estimate	SE	CV (%)
lambda	11.7	0.0823	0.7	19.7	0.144	0.7
delta	0.0	--	--	-0.688	0.00823	1.2
$mict_{base}$	11.7	0.0823	0.7	11.54	0.0437	0.4
Eff	0.132	0.0054	4.1	0.119	0.00607	5.1
k (d^{-1})	0.0649	0.00529	8.2	0.0608	0.00292	4.8

Note: $mict_{base}$ was derived by $\lambda/(1-\delta)$

Key References

- [1] Consul PC. (1989) Generalized Poisson Distributions. Properties and Applications. Statistics: textbooks and monographs Vol 99. Marcel Dekker, Inc.
- [2] Consul & Jain (1973). A Generalization of the Poisson distribution. Technometrics 15, 791-799
- [3] Plan & Karlsson (2009). New models for handling correlated underdispersed Likert pain scores. PAGE 2009.

Conclusions

- The GP model was found to be superior to the PS model
- GP better described variability observed in micturition count data
- GP yielded more precise estimates but not for all parameters.
- As a result, the GP model is expected to provide more accurate inferences, such as drug efficacy predictions and clinical trial simulations.