# How many bits of information did my study provide? Kullback-Leibler information gain, standard errors and shrinkage

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# **Background and Objectives**

Information theory concerns the quantification, storage, and communication of information. The key measure in information theory is entropy which can be expressed in the units of **bits**. The Kullback-Leibler divergence  $D_{\kappa}$  is a widely used measure in information theory. Many other quantities can be interpreted in terms of  $D_{\kappa}$  and it is often referred to as "information gain". Our objective was to explore measures of information gain from model estimation.

We evaluated the following properties for a information measure:

1) Units of *bits*, the natural unit of information 2) Increase proportionally with number of individuals 3) Increase with the number of observations per individual 4) Unchanged by non-informative individuals 5) Increase with decreasing observation error 6) Increase with decreasing parameter relative standard-error

# Methods and Results

### Methods

Simulations were performed using a onecompartment PK model with absorption:

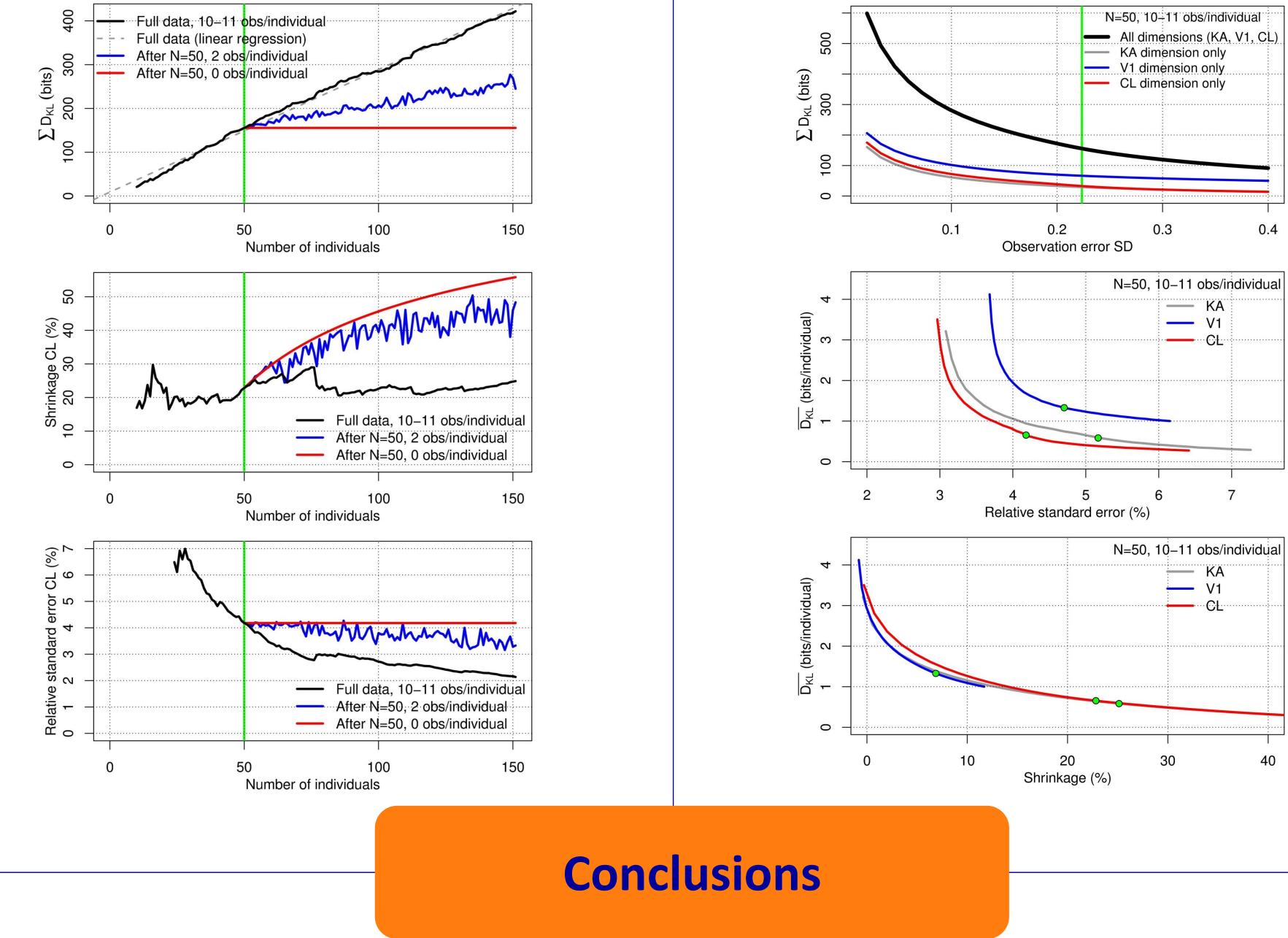
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 $KA = 0.6 \cdot \left(WGT / 70kg\right)^{-0.25} \cdot \exp(\eta 1)$  $V1 = 20 \cdot (WGT / 70kg) \cdot \exp(\eta 2)$  $CL = 1 \cdot (WGT / 70kg)^{0.75} \cdot \exp(\eta 3)$  $IPRED = C_{V1} \cdot (1 + \varepsilon)$ 

#### $\eta 1$ , $\eta 2$ , $\eta 3$ inter-individual variance 0.05 ε observation error standard deviation 0.2236 Simulated WGT varied from 54.6-86.4 kg and there were 10-11 samples per individual

## Varying number of individuals and data richness

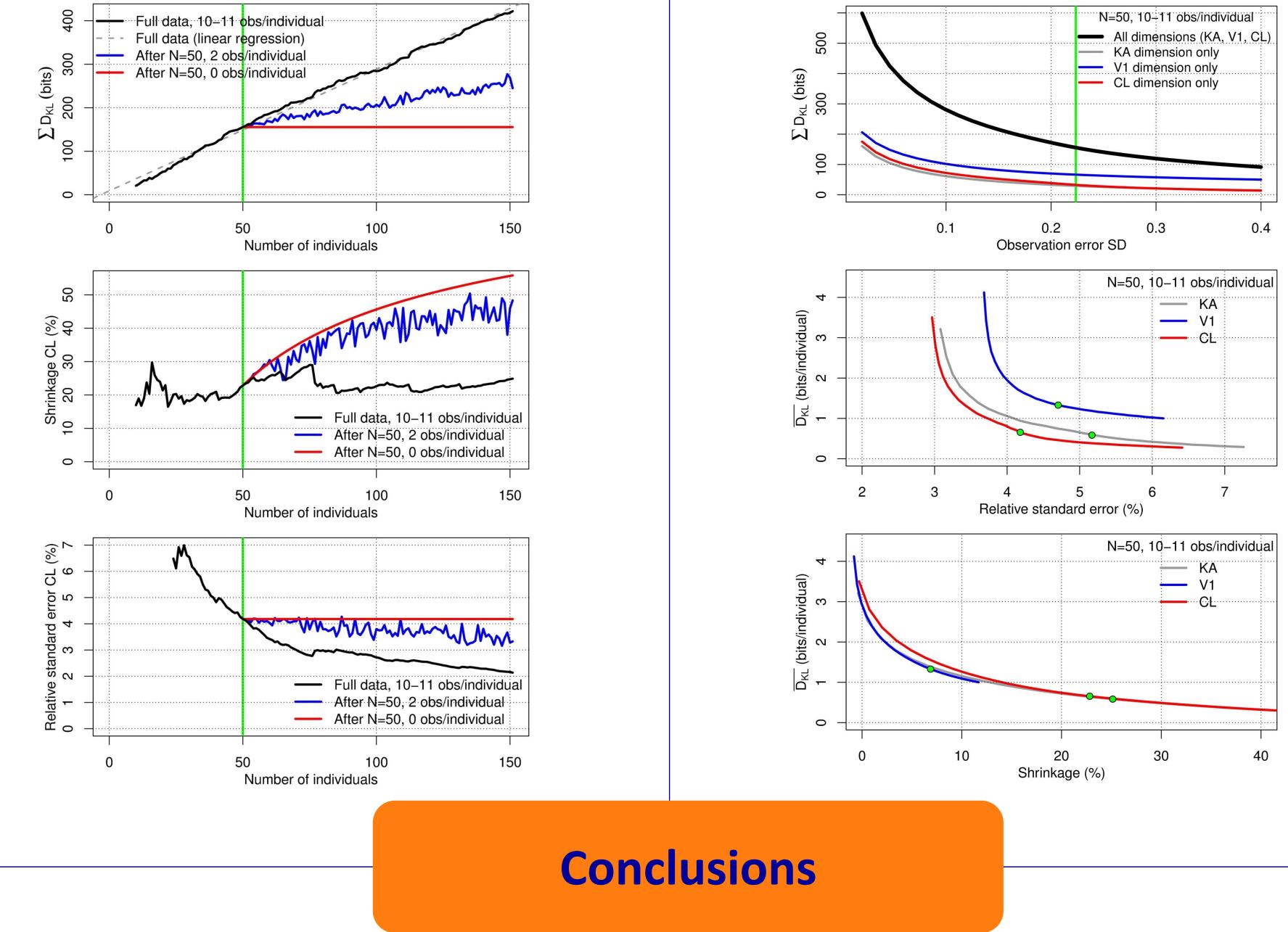
- For constant observations/individual,  $\Sigma D_{\kappa l}$  is  $\bullet$ proportional to the number of individuals
- $\Sigma D_{\kappa}$  increased with number of  $\bullet$ observations/individual
- $\Sigma D_{\kappa}$  constant with non-informative individuals
- Shrinkage reflects average information content and creating interpretation difficulties with unequal-data-richness across individuals

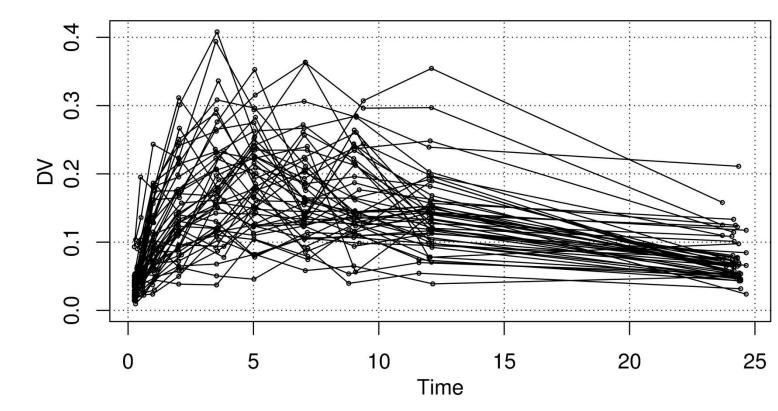


## Varying observation error

For N=50 and 10-11 observation/individual

- $\Sigma D_{\kappa_l}$ , and each dimension, decreases with observation error
- Relative standard error decreases with  $D_{\kappa}$  for each dimension
- In this equal-data-richness case,  $\overline{D_{\kappa_l}}$  for each dimension seems related to  $\eta$ -shrinkage





### **Kullback-Leibler divergence**

For multivariate normal distribution the  $D_{\kappa}(NO||N1)$  of  $NO(\mu 0, \Sigma 0)$  from  $N1(\mu 1, \Sigma 1)$  is:

$$= \frac{1}{2} \left\{ tr \left( \Sigma_1^{-1} \cdot \Sigma_0 \right) + \left( \mu_1 - \mu_0 \right)^T \Sigma_1^{-1} \left( \mu_1 - \mu_0 \right) - k + \ln \left( \frac{|\Sigma_1|}{|\Sigma_0|} \right) \right\}$$

where NO to is the individual estimated distribution and N1 to represent the estimated population distribution.  $D_{\kappa}$  represents the information gained when one revises one's beliefs from the prior N1 to posterior N0. We sum  $D_{\kappa I}$ over the individuals to obtain  $\Sigma D_{KL}$ . Dividing the equation above by *ln(2)* results in units of *bits*.

#### Simulations

- The number of individuals was N=10-150. Mixed data-richness was simulated by limiting the number of observations to 2 when N>50. Non-informative individuals had 0 observations
- Observation error  $\varepsilon$  was varied from 0.02-0.4.

Evaluating  $\Sigma D_{\kappa}$  satisfies the properties expected of a information measure. This allows a new quantification of information obtained by model estimation. For the reference PK model:

Parameter	η1 (KA)	η2 (V1)	η2 (CL)	Total
Estimated variance	0.0616	0.0793	0.0414	
Relative standard error (%)	5.17	4.70	4.18	
Shrinkage (%)	25.1	6.86	22.8	
ΣD <sub>KL</sub> (bits)	29.2	66.4	32.7	155.5

# **Questions for future work**

- How do we interpret the physical meaning of **bits**?
- How is the information obtained related to non-eta parameters?
- Does maximizing  $\Sigma D_{\kappa_l}$  lead to optimal study design?
- Is there a relationship (possibly approximate) between  $\Sigma D_{\kappa l}$  and parameter certainty (i.e. standard errors)?
- Are  $\Sigma D_{\kappa l}$  summaries useful to combine multiple studies of varying quality?