# **Supplemental Material**

Estimating nephron number from biopsies: impact on clinical studies

Darya Morozov<sup>1</sup>, Neda Parvin<sup>1</sup>, Mark Conaway<sup>3</sup>, Gavin Oxley<sup>4</sup>, Edwin J. Baldelomar<sup>1</sup>, Aleksandra Cwiek<sup>4</sup>, Kim deRonde<sup>2</sup>, Scott C. Beeman<sup>5</sup>, Jennifer R. Charlton<sup>+2\*</sup>, and Kevin M. Bennett<sup>1+\*</sup>

<sup>1</sup>Mallinckrodt Institute of Radiology, Washington University School of Medicine, St. Louis, MO

<sup>2</sup>Department of Pediatrics, University of Virginia

<sup>3</sup>Department of Public Health Sciences, University of Virginia

<sup>4</sup>University of Virginia

<sup>5</sup>School of Biological and Health Systems Engineering, Arizona State University

+ Equal contribution

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#### **Statistical analysis:**

Reliability of needle and virtual biopsies:

The analysis was based on a 'replication reliability study' (Fleiss, The Design and Analysis of Clinical Experiments, 1986, Section 1.3) (1). These analyses help establish the reliability referring to the reproducibility of biopsy to estimate the whole kidney N<sub>glom</sub>, and reveal the effect of measurement error on statistical power for comparing groups.

<u>Needle Biopsy:</u> Based on the statistical model of reliability (1), let  $T_i$  be the whole kidney  $N_{glom}$  measurement from person i, i = 1,...,N. This value is measured imperfectly by  $Y_{ij}$ , the  $j^{th}$  needle biopsy from kidney i, i = 1,...,n and j = 1,...,m. The  $N_{glom\_NB}$  and whole kidney  $N_{glom}$  values are related by the model in Eq. 5:

$$Y_{ij} = \beta + T_i + \varepsilon_{ij}$$
 Eq.5

with  $E(T_i) = \mu$ ,  $Var(T_i) = \sigma_T^2$ . The  $\varepsilon_{ij}$  are assumed independent with mean 0 and variance  $\sigma_e^2$ .

The accuracy of N<sub>glom\_NB</sub> estimated from needle biopsies compared to "gold standard" N<sub>glom</sub> from CFE-MRI is based on the following model:

$$Y_{ii} - T_i = \beta + \varepsilon_{ii}$$
 Eq. 6

where  $T_i$  taken as a fixed and observed whole kidney  $N_{glom}$  for the  $i^{th}$  kidney. The bias in the needle biopsy-based  $N_{glom}$  measurement is  $\beta$  and the variance is  $\sigma_e^2$ . Estimates of the bias and variability parameters can be obtained in SAS 9.4 using PROC GLM or PROC MIXED.

### Virtual Biopsy:

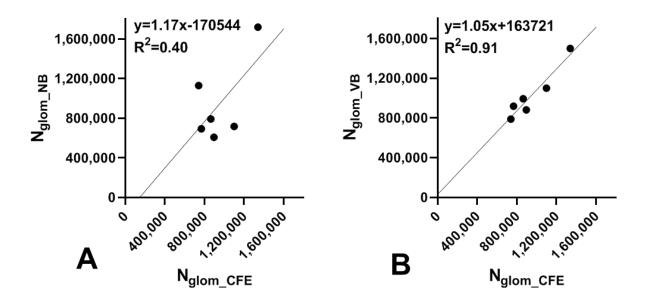
The statistical model for reliability of Nglom VB is:

$$Y_{ijkl} = T_i + \beta_{j(i)} + \gamma_{k(ij)} + \varepsilon_{ijkl}$$
 Eq. 7

$$E(T_i) = \mu$$
,  $Var(T_i) = \sigma_T^2$ ;  $E(\beta_{j(i)}) = \theta_j$ ,  $Var(\beta_{j(i)}) = \sigma_s^2$   
 $E(\gamma_{k(ij)}) = 0$ ,  $Var(\beta_{j(i)}) = \sigma_c^2$ ;  $E(\varepsilon_{ijkl}) = 0$ ,  $Var(\varepsilon_{ijkl}) = \sigma_e^2$ 

Where *i* refers to kidney, *j(i)* refers to site within kidney, *k(ij)* refers to cluster and *l* is the measurement within a cluster in a site in a kidney. As above, this model can be fit in SAS 9.4 PROC GLM or PROC MIXED. Using PROC MIXED has some advantages in testing the assumption of this model. The 'group' and 'local' options in the 'repeated' and 'random' statements in SAS PROC MIXED allow for testing, for example, whether the variability due to sites within the kidney differs by site.

## **Linear regression plots:**



**Figure 1S.** Linear regression plots between (A) mean  $N_{glom}$  measured from needle biopsies ( $N_{glom\_NB}$ ) and  $N_{glom}$  measured by CFE-MRI ( $N_{glom\_CFE}$ ) or (B) mean  $N_{glom}$  measured from virtual biopsies ( $N_{glom\_VB}$ ) and  $N_{glom\_CFE}$ .

## References:

1. **Fleiss JL**. The Design and Analysis of Clinical Experiments. . In: *Biometrical JournaWiley*, New York – Chichester – Brislane – Toronto – Singapore 1986, 432 S., 1986.