

# Model based stratification for sampling of soil carbon stock changes

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# Introduction

- ▶ Models provide a tempting way to monitor soil carbon stock changes.
- ▶ However, ultimate verification of predictions comes from measurements
- ▶ Measuring soil carbon stock changes is notoriously laborious
- ▶ Stratification of sampling improves the precision

## Aim

- ▶ To seek an answer to hypothesis:
  - Model-predicted changes of soil carbon stock may be used to improve the precision of soil sampling

# Content

- ▶ Stratification method
- ▶ Simulations for stratifications
- ▶ Results: simulations, stratification, relative precision, within site variability, comparison of stratification
- ▶ Conclusions

# Stratification method

- ▶ Stratification may be used to improve the precision of sampling
  - More similar sites to same strata
- ▶ Approximately optimal stratification by dividing the cumulative distribution function of  $\sqrt{PDF}$  to equal intervals (Cochran, 1977)
  - Here, the PDF is for measured soil carbon stock change
  - I.e. stratification by soil carbon stock change
- ▶ Measured soil carbon stock changes unknown -> **use simulated values**

# Sampling from strata

- ▶ Simple random sampling (SRS) without replacement
- ▶ Sampling from strata by
  - Equal allocation of samples
  - Proportional allocation
  - Neyman allocation
- ▶ With/Without within site variability and measurement error

# Simulations for stratification

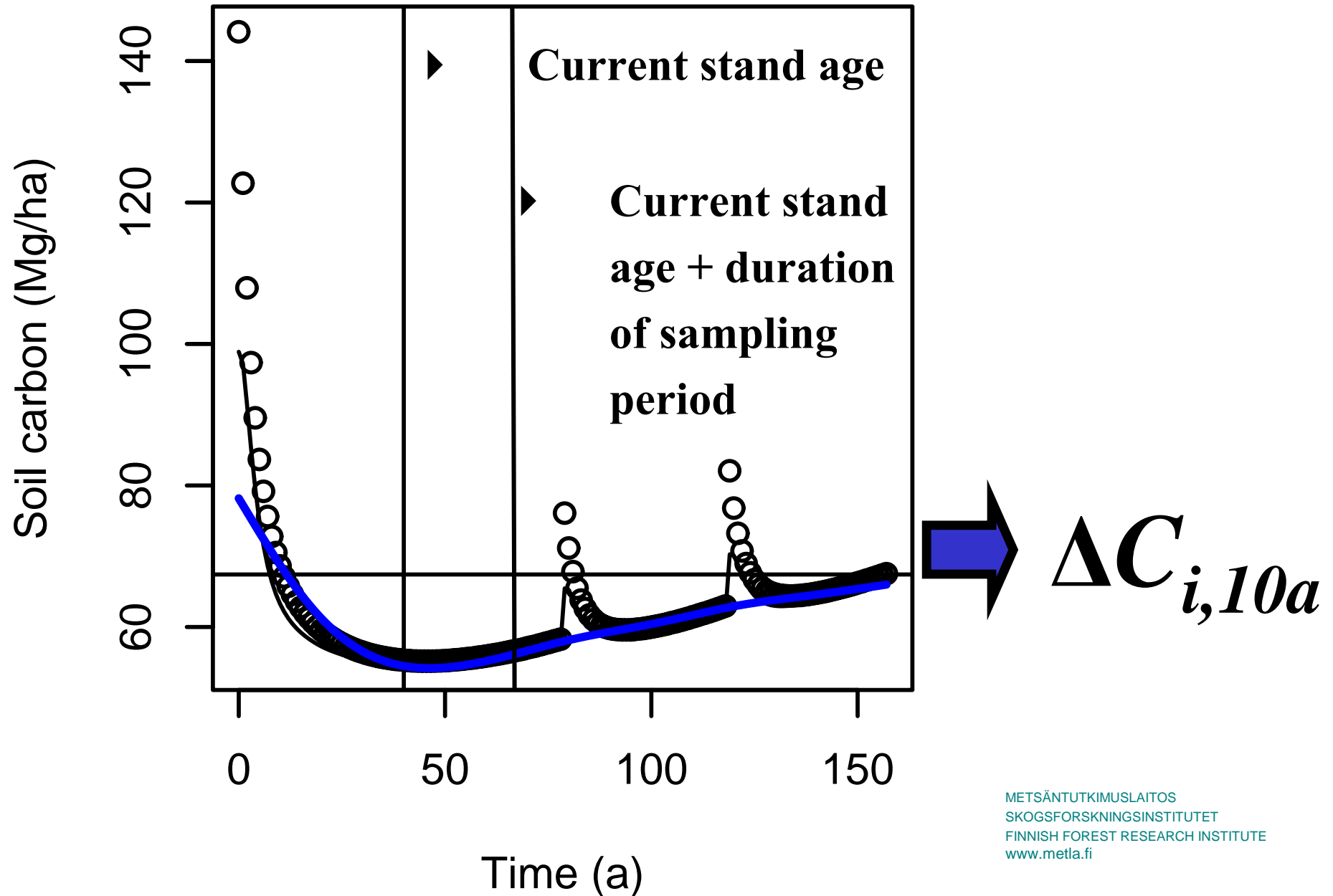
- ▶ What is the predicted change of soil carbon on each plot during the sampling period (e.g 10 a)?
- ▶ Motti stand model + Yasso soil model
  - Typical management scenario assumed
    - most of the Finnish forests are managed
    - Old-growths stand --> the change estimate from last years of the rotation

## Material for simulations

- ▶ NFI permanent sample plots
  - Representative for Finnish forests
  - Subset of upland mineral soils used
- ▶ Data on: location, temperature sum, fertility, dominant tree species, stand age
- ▶ Typical seedling stands for each species

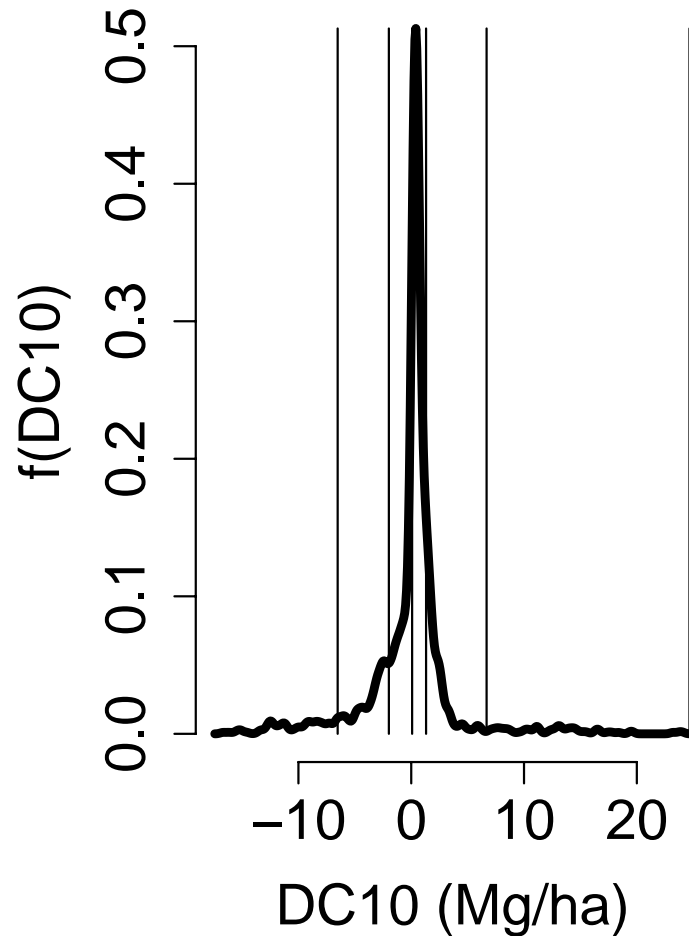


# Example of simulation on plot *i*

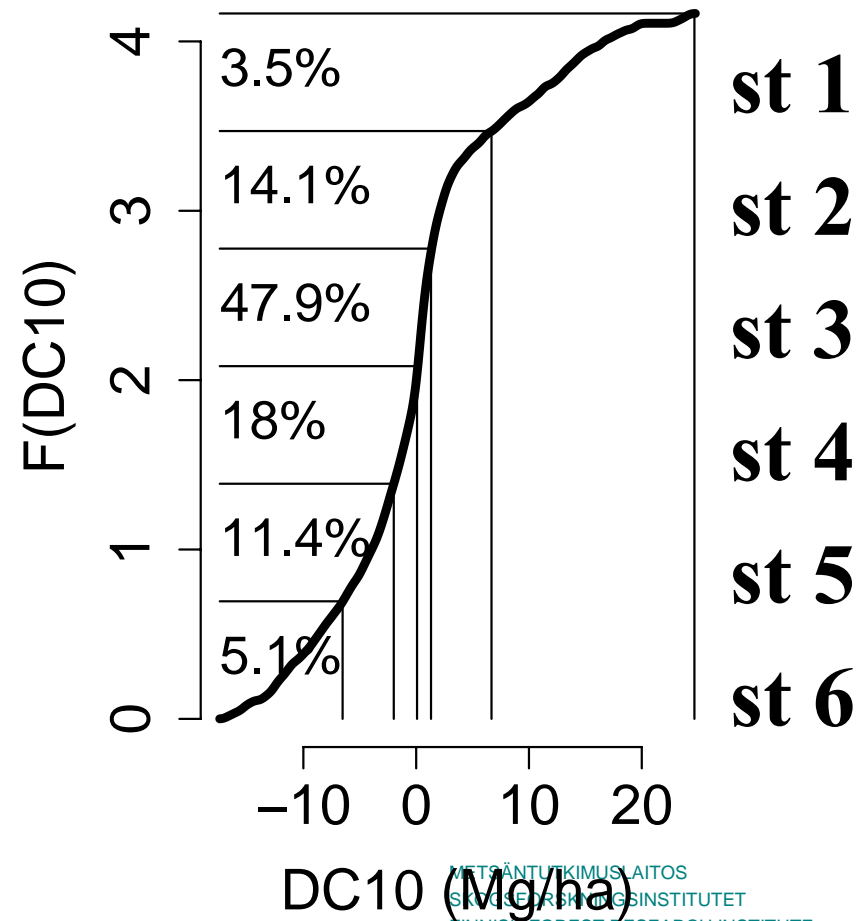


# Stratification, 6 strata

Probability Density Function

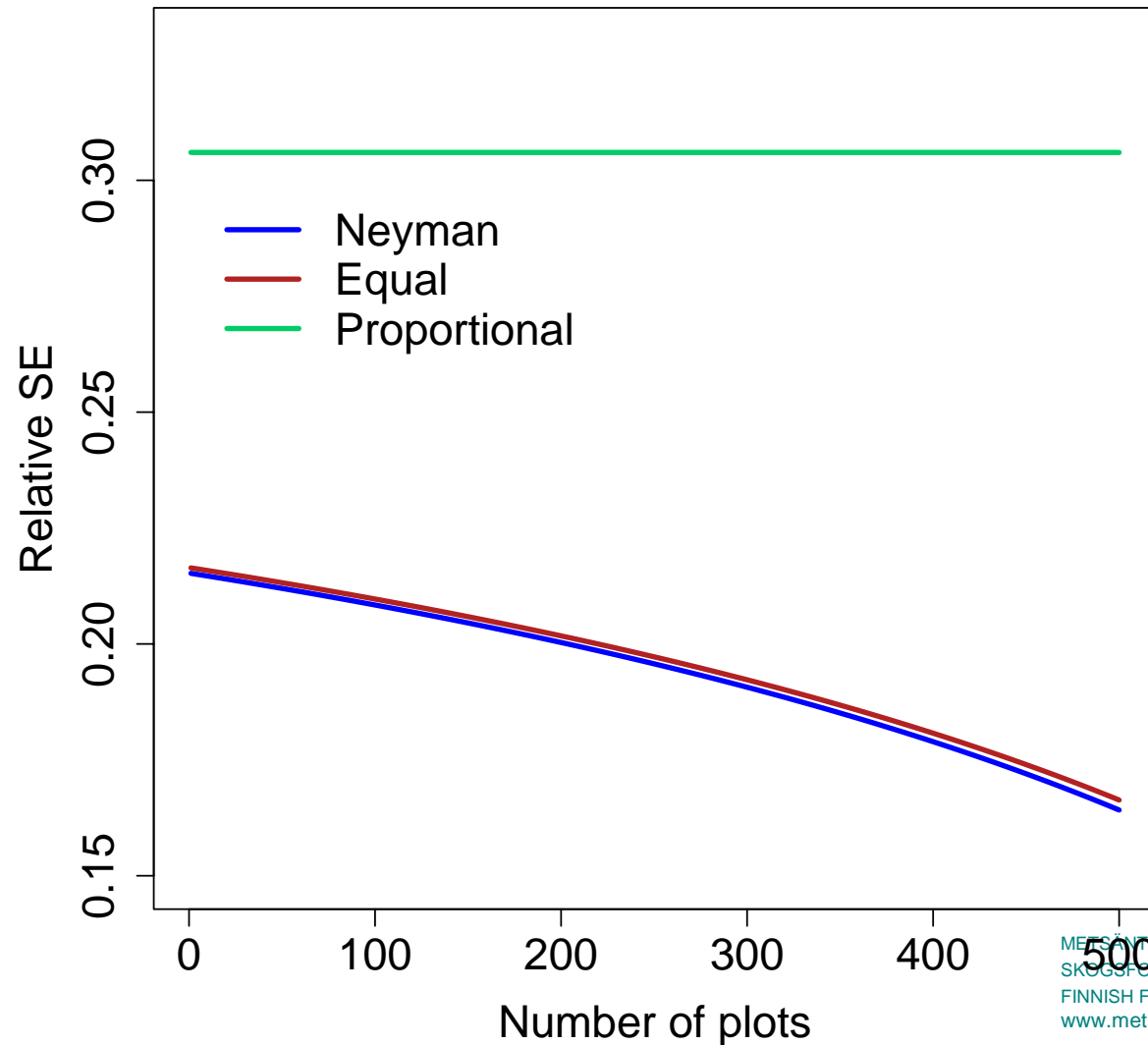


Cumulative Distribution Function



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# Sampling from strata: precision relative to Simple random sampling



# Within site variability and measurement precision, preliminary assumption

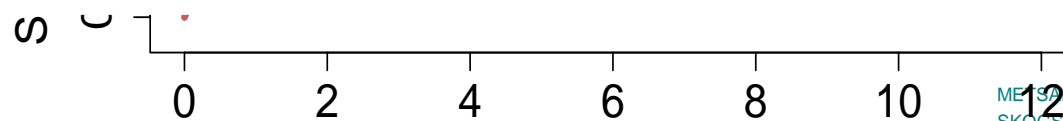
12) 6

If two independent random samplings are assumed for a plot :

$$\sigma_{\Delta C}^2 = \sigma_{C1}^2 + \sigma_{C2}^2$$

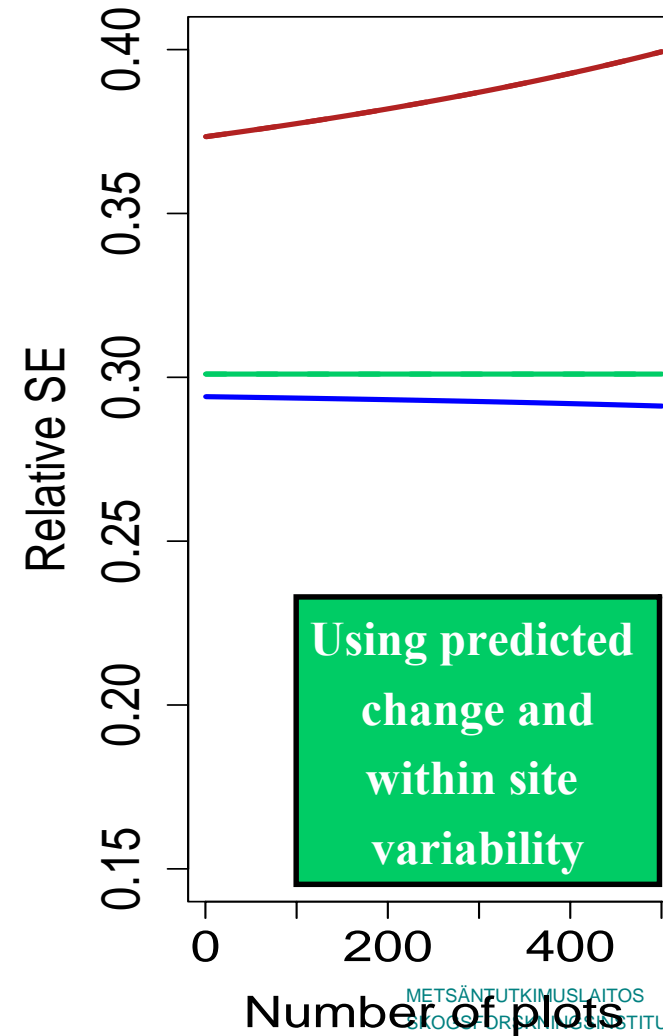
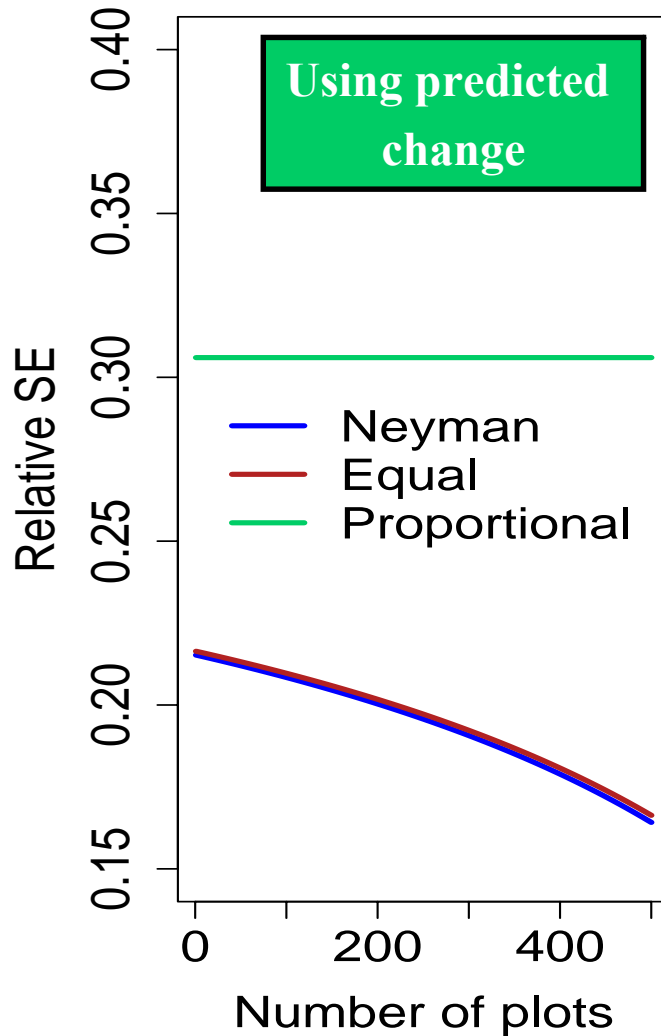
$$SE_{\Delta C} = \sqrt{\frac{\sigma_{\Delta C}^2}{N}}$$

Say, N= 100

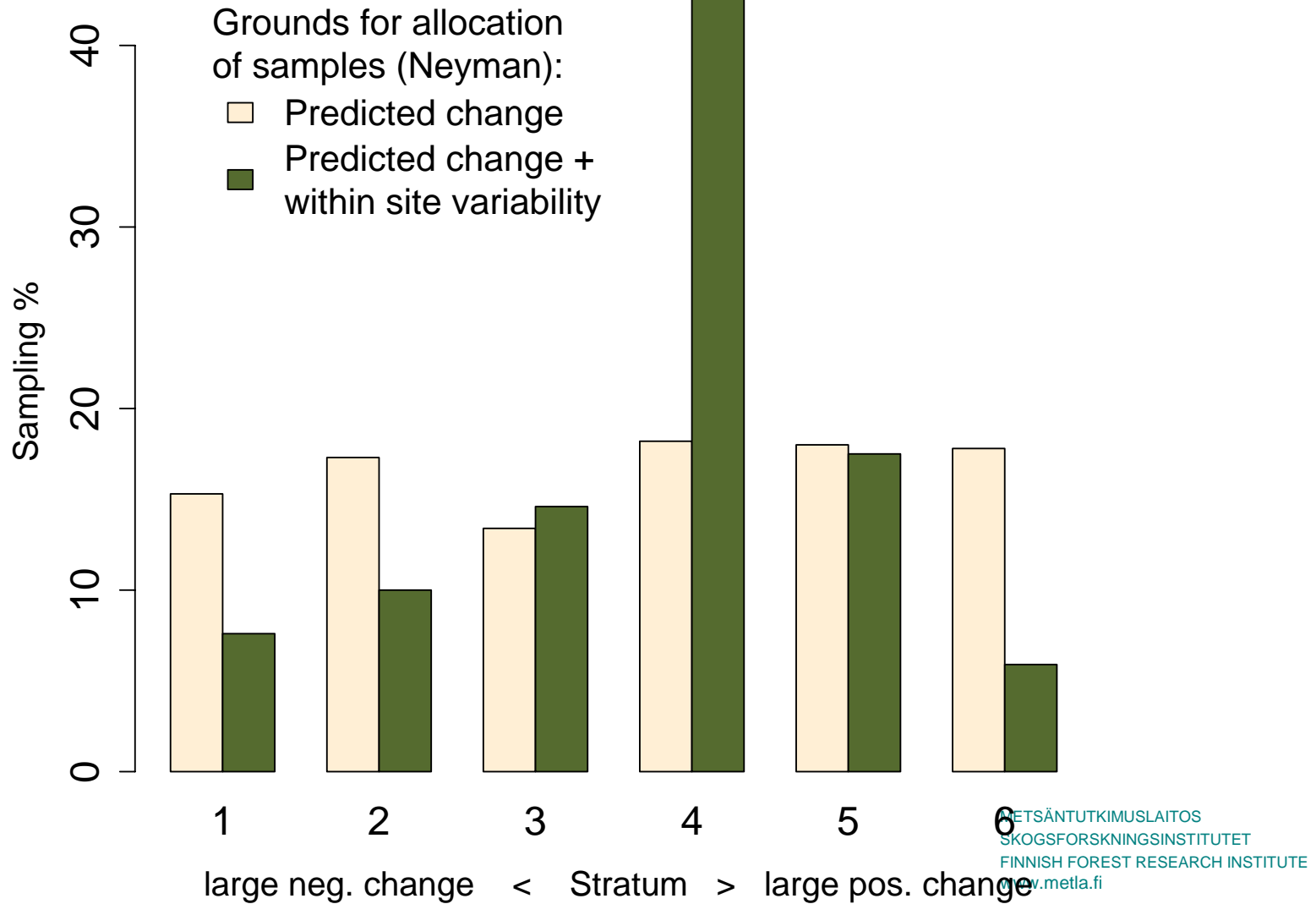


Avg. carbon density of a site (kg/m<sup>2</sup>)

# Variability reduces the benefits of stratification



# Allocation of samples



# Conclusions

- ▶ Within site variability changes the allocation of samples to strata completely from that of using only model predictions
- ▶ Hypothesis:
  - true, but only if effects of within site variability can be controlled.
  - Why: Large within site variability makes it difficult to say to what strata this site belongs to
- ▶ If stratification will be used in soil sampling it would be useful to have information on within site variability
- ▶ Need for precise measurements on each plot

## Continuation of work

- ▶ Check the variance-mean relation on plot level
- ▶ Replace independent repeated sampling on a plot level to produce better estimates of N and variance of mean
- ▶ Run tests with refined assumptions
- ▶ Stratification for permanent monitoring of sites instead of one repeated sampling
  - combination of permanent and temporary plots

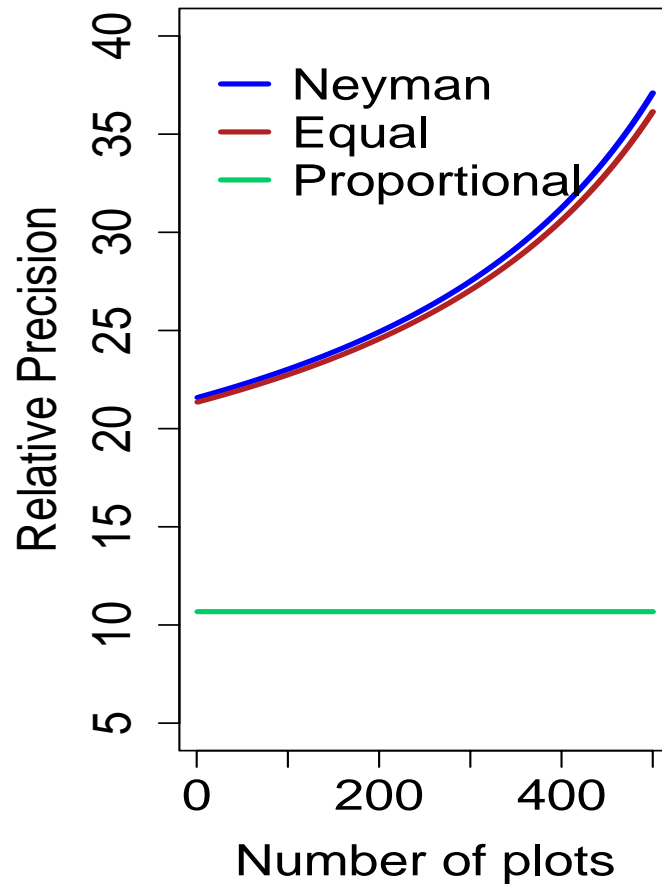


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**Thank you**

# Variability reduces the benefit of stratification

Using predicted change



Using predicted change and within site variability

