

Bayesian strain-rate correction for marine sediments characterization

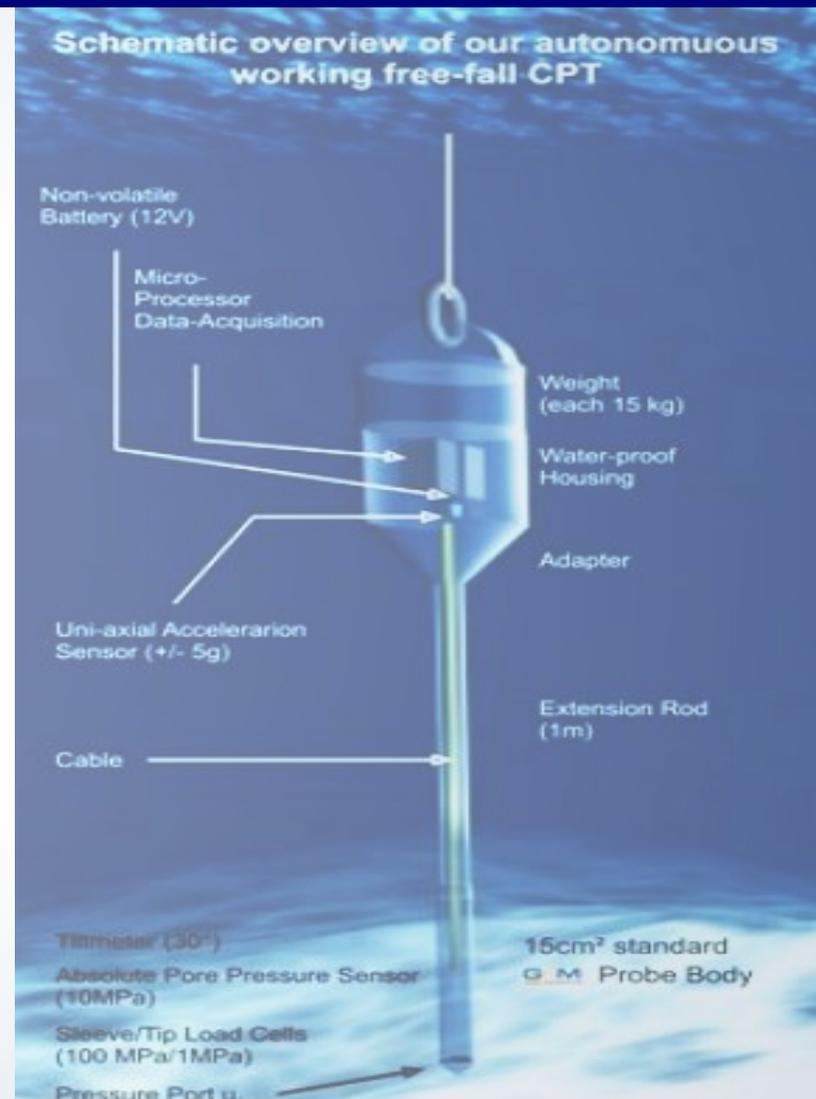
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Dynamic Cone Penetration Test

- Introduction
- Main uncertainty in Dynamic data process: Strain-Rate effect
- How to improve conversion accuracy from Dynamic to Static measurements?
(Further predict of geotechnical design parameters (e.g. C_u , friction angle))



Dynamic Cone Penetration Test (SW-CPTu)

Shallow water Dynamic CPTu (500m water depth)

Application:

- Marine geosciences (e.g. slope stability research);
- Engineering practice (e.g. exploration of cable tracks).

Main advantages vs Static CPTu:

- Relative light-weight;
- Fast test execution;
- Less sensitive to bad weather conditions (Stegmann 2006).

Main drawback:

- Penetration velocity \gg static test (constant velocity)
- Data process requires rate effect quantification.



Figure 1. Submarine application of Dynamic CPTu

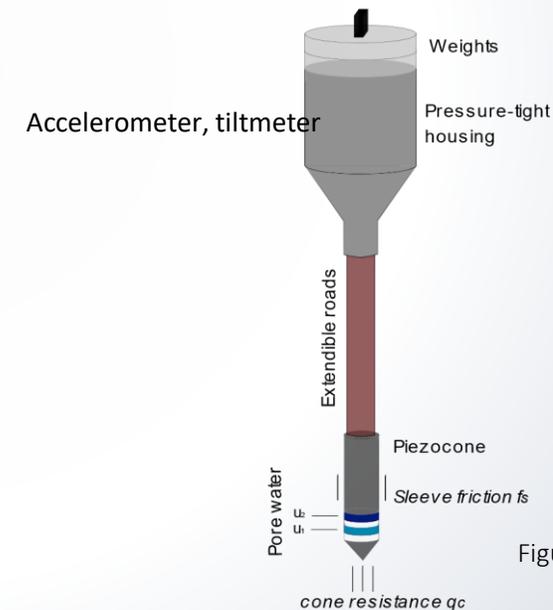


Figure 2. Sketch of Dynamic CPTu equipment

Introduction: Strain-Rate effect

Key Point: Soil properties depend on the rate at which soils are deformed (strain rate).

Rate effect in fine grained soils:

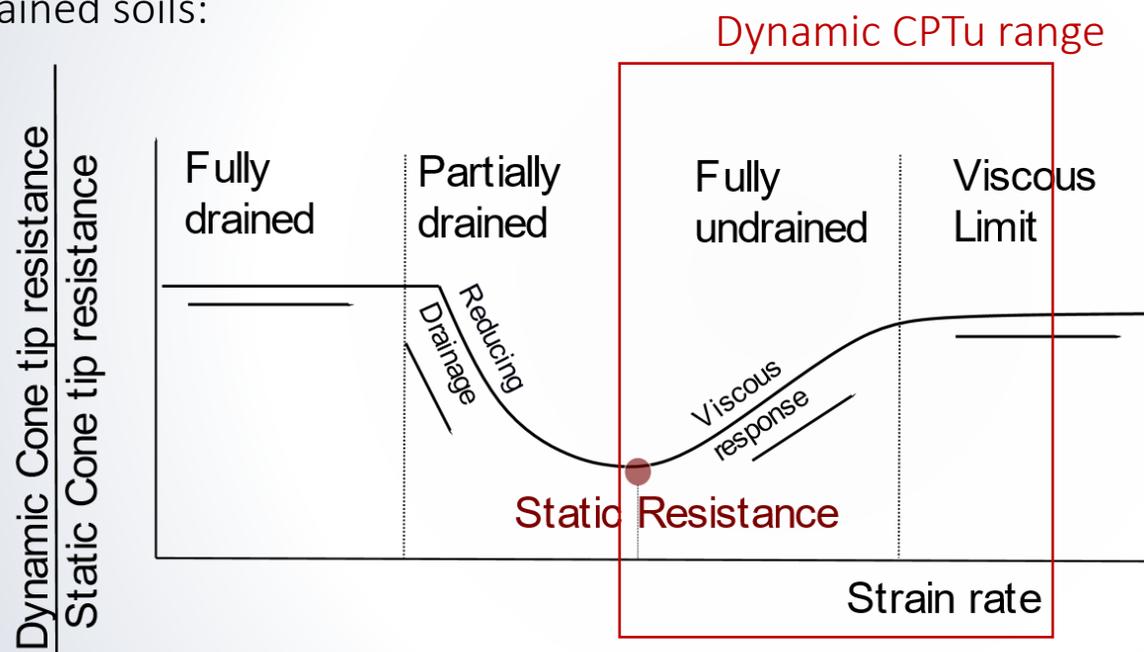


Figure 3. Rate effect variation at varying strain level (Quinn and Brown, 2011, Lehane et al., 2009)

Introduction: Strain-Rate effect

Example:

Homogeneous clay samples

1st issue:

Strain-rate effect q_t

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Strain-rate effect f_s

2nd issue:

Strain-rate (Impact Velocity, depth, Soil Type)

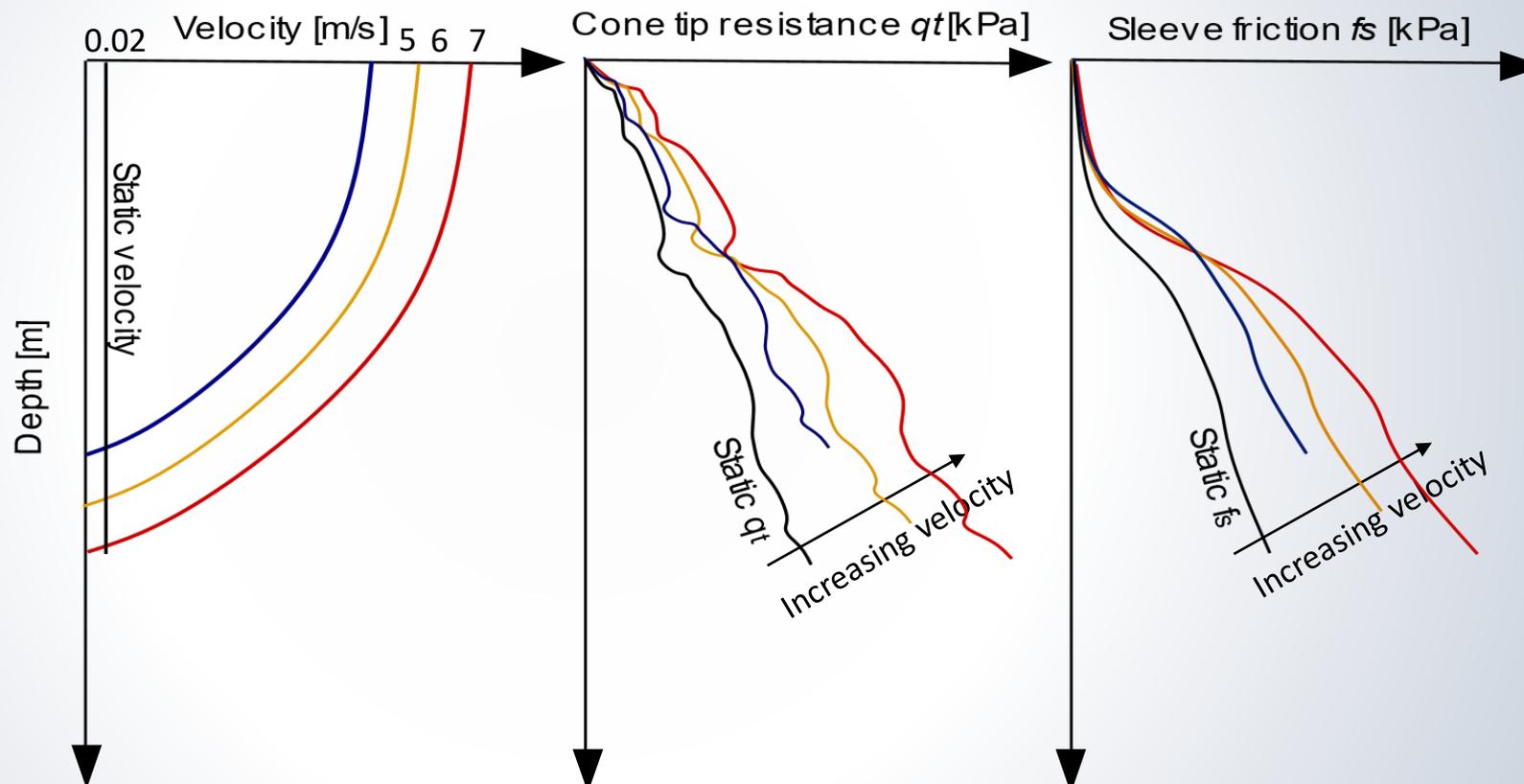


Figure 4. Example of strain-rate effect on cone tip resistance and sleeve friction
Each CPTu measurements must be converted to static ones

Introduction: Strain-Rate Coefficient

SRF = Strain-Rate Factor

$$SRF = \left(\frac{v_{dyn}}{v_{stat}} \right)^{\beta}$$

v_{dyn} = penetration velocity of SW- CPTu;
 v_{stat} = penetration velocity of static CPTu;

Known parameters

β = strain-rate coefficient

Simplistic model:

$$\text{Converted } q_{t \text{ static}} = q_{t \text{ dyn}} / SRF$$

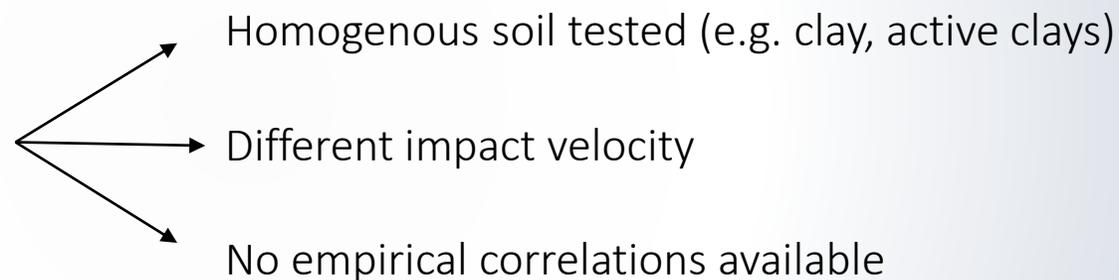
Introduction: Strain-Rate Coefficient

General data acquisition at a site: Dynamic CPTU + Gravity cores

General approach: β estimated from Literature: average value from experimental studies

Literature: empirical estimation of β :

(Steiner et al., 2012, Chow et al. 2017, Lehane et al. 2009)



Possible misleading conversion of in situ CPTu

How to improve β prediction?

$$\text{Minimizing: } q_{t\text{-stat}} - q_{t\text{ dyn}} / \left(\frac{v_{dyn}}{v_{stat}} \right)^\beta = 0$$

$$\mu_{(q_{t\text{-stat}})} - \mu_{(q_{t\text{ dyn}} / \left(\frac{v_{dyn}}{v_{stat}} \right)^\beta)} = 0$$

Mean value Mean value

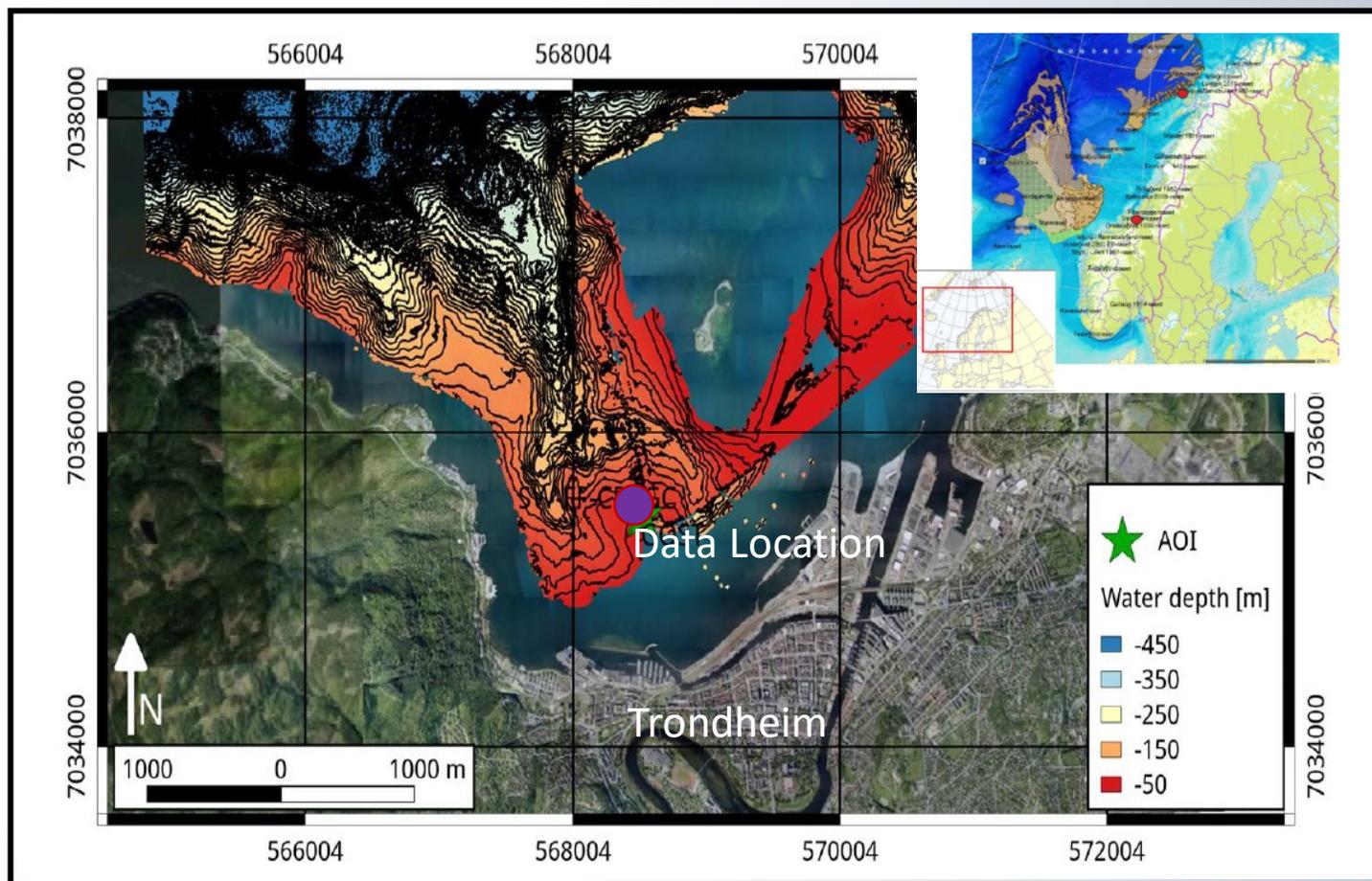
$$f(\mu_{(q_{t\text{-stat}})}, \beta) = 0$$

Unknown: $\mu_{(q_{t\text{-stat}})}, \beta$

Study area:

Location: Trondheim, Norway.

Recent studies: Roskoden 2015-Phd Thesis, L'Heureux, J. et al., 2010).



Roskoden (2015), PhD Thesis.

Gravity cores

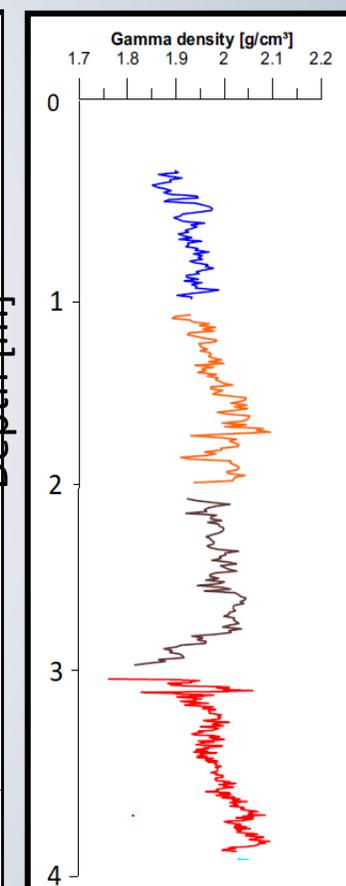
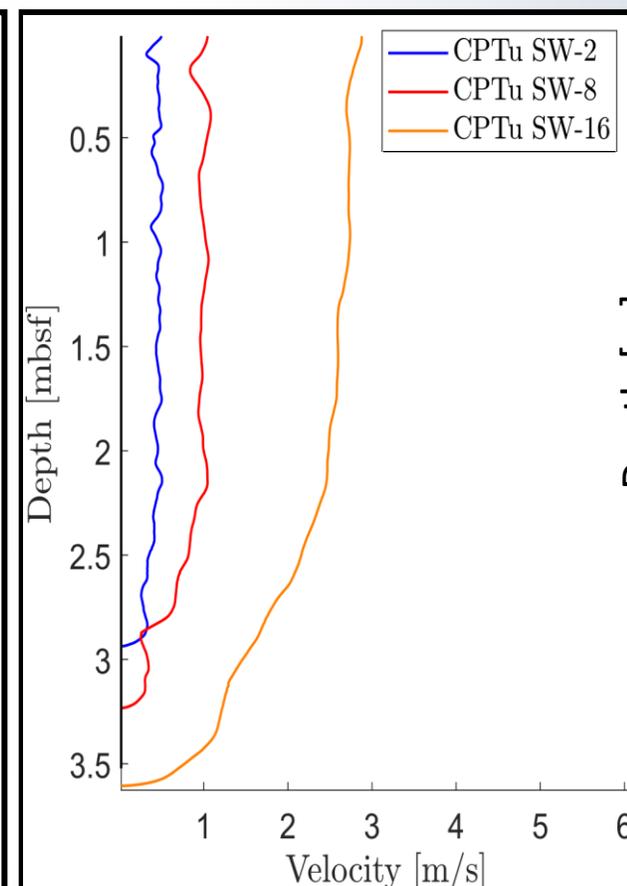
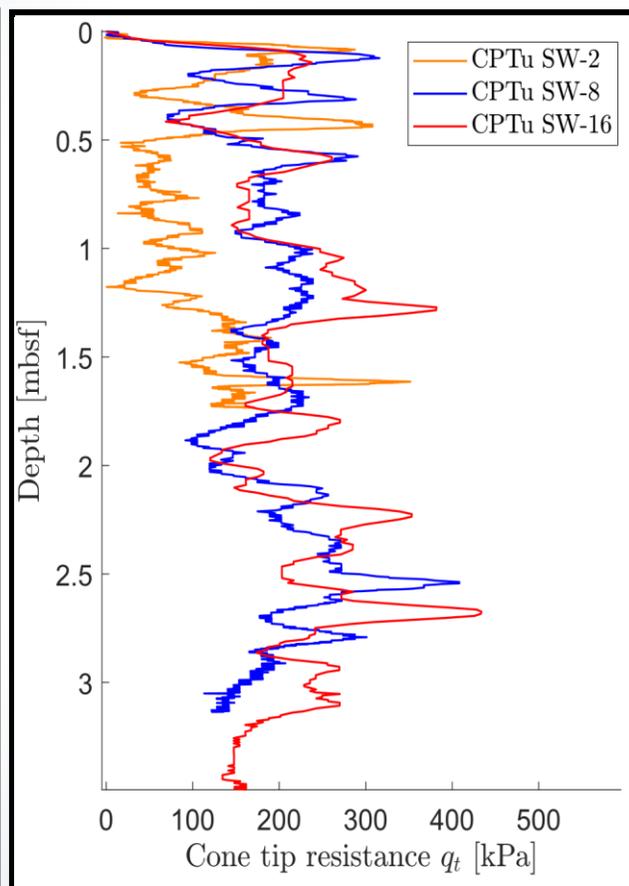
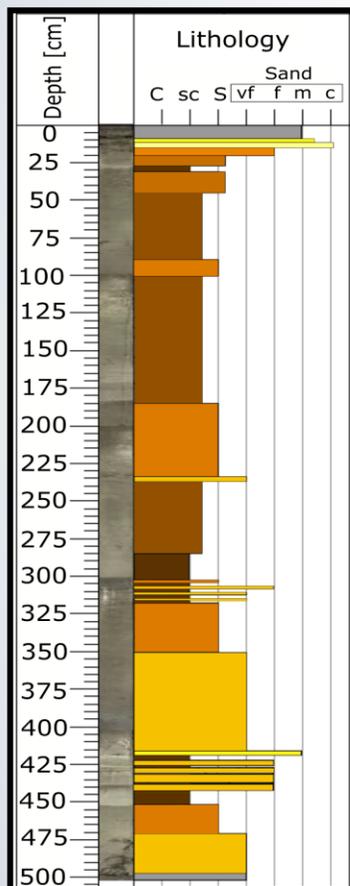
Dynamic CPTu

Penetration velocity [m/s]

Lab. Data

Data acquired:

- Gravity cores
- 16 Dynamic CPTu
- Lab. Data
(Soil unit weight)



Unknown: $\mu_{(q_{t-stat})}$, β

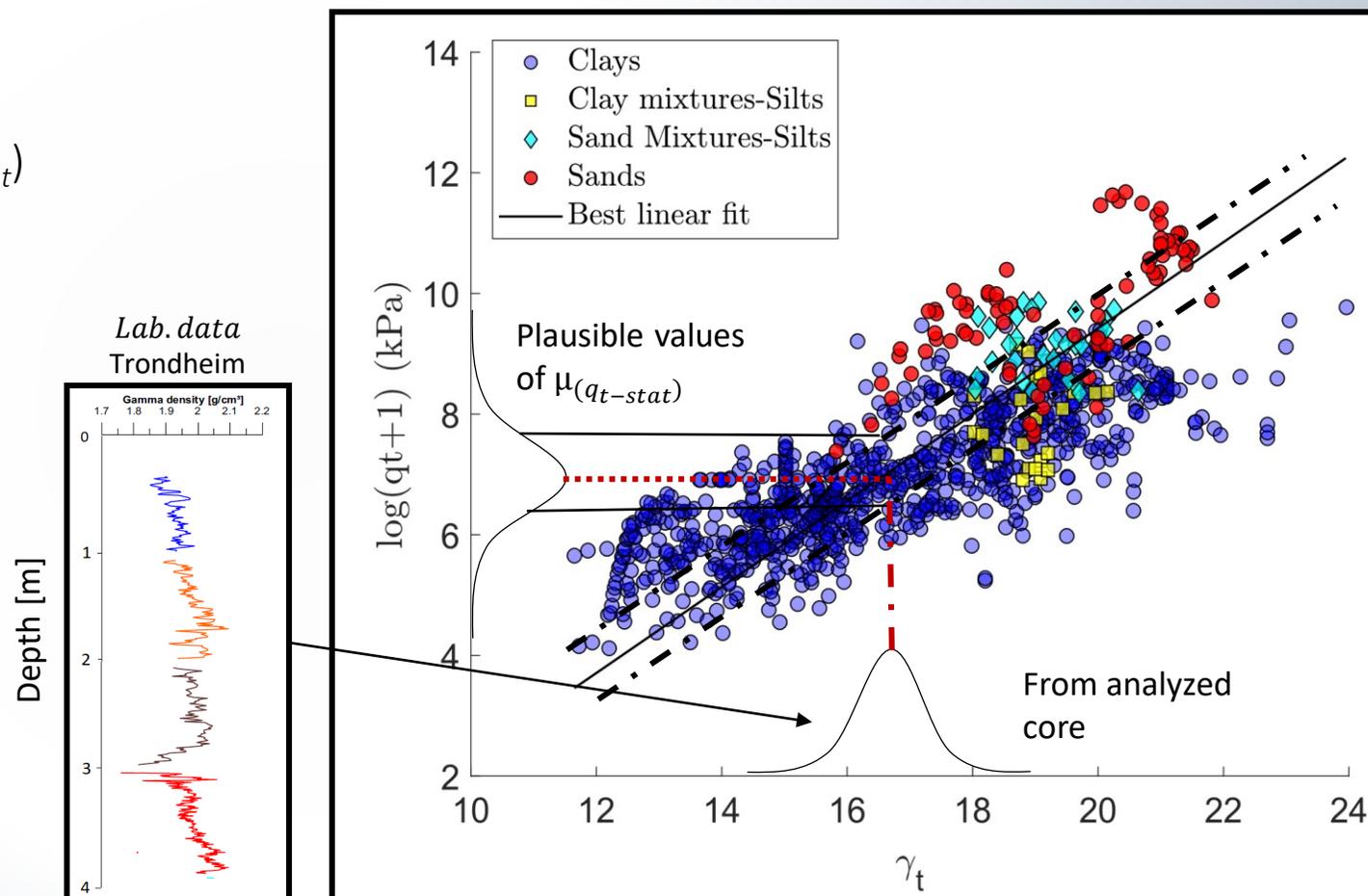
$\mu_{(q_{t-stat})}$: Empirical correlations

Soil unit weight- static CPTu parameters (e.g. q_t)

Regression introduce uncertainty $\mu_{(q_{t-stat})}$

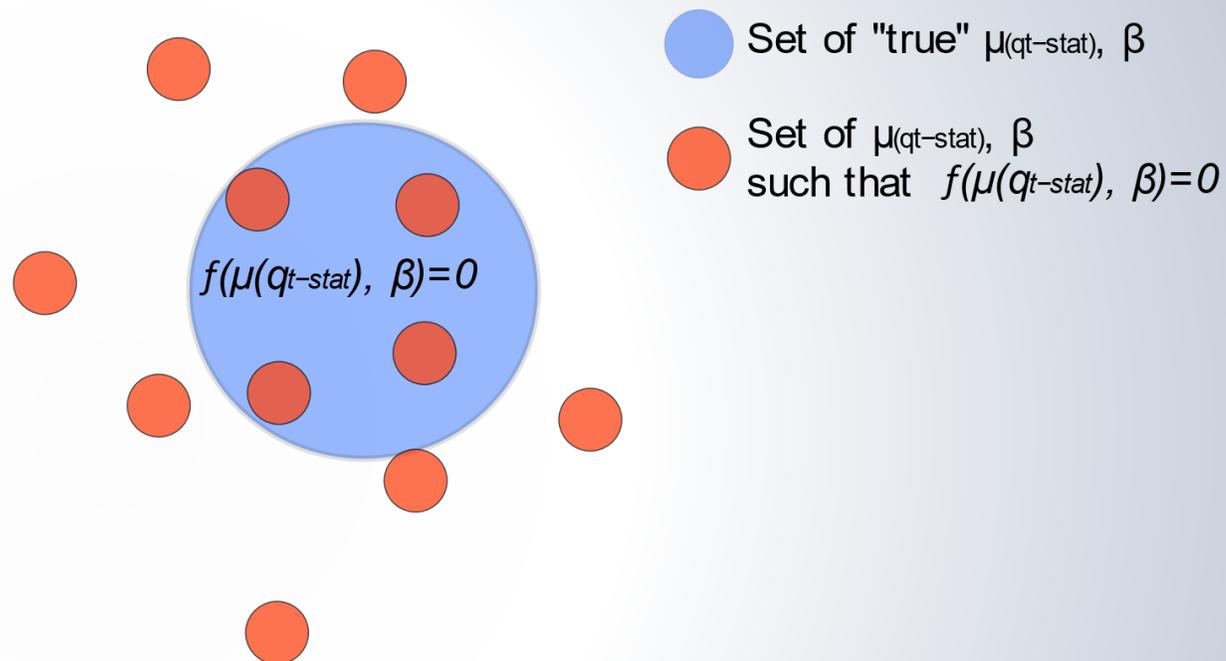
$$f(\mu_{(q_{t-stat})}, \beta) = 0$$

Probabilistic approach



$$\mu(q_{t-stat}) - \mu(q_{t-dyn} / \left(\frac{v_{dyn}}{v_{stat}}\right)^\beta) = 0$$

Least square method: ~~poorly condition~~
(Observations & ~~exponential coefficient~~ -unknown)

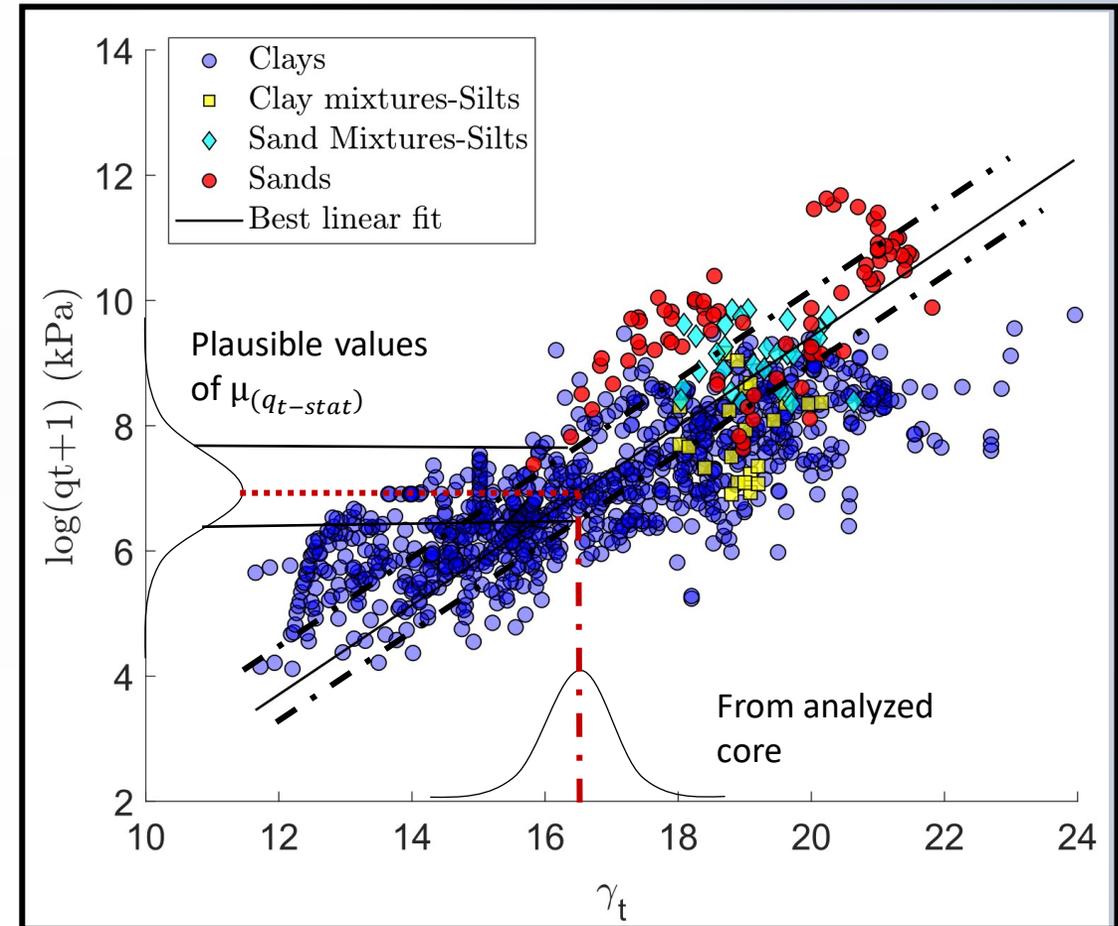


Bayesian Equivalent samples approach

(Zhang et al., 2012)

Methodology: Bayesian approach.

1st: Proposal distribution of $\mu_{(q_t-stat)}$:
from Soil unit weight

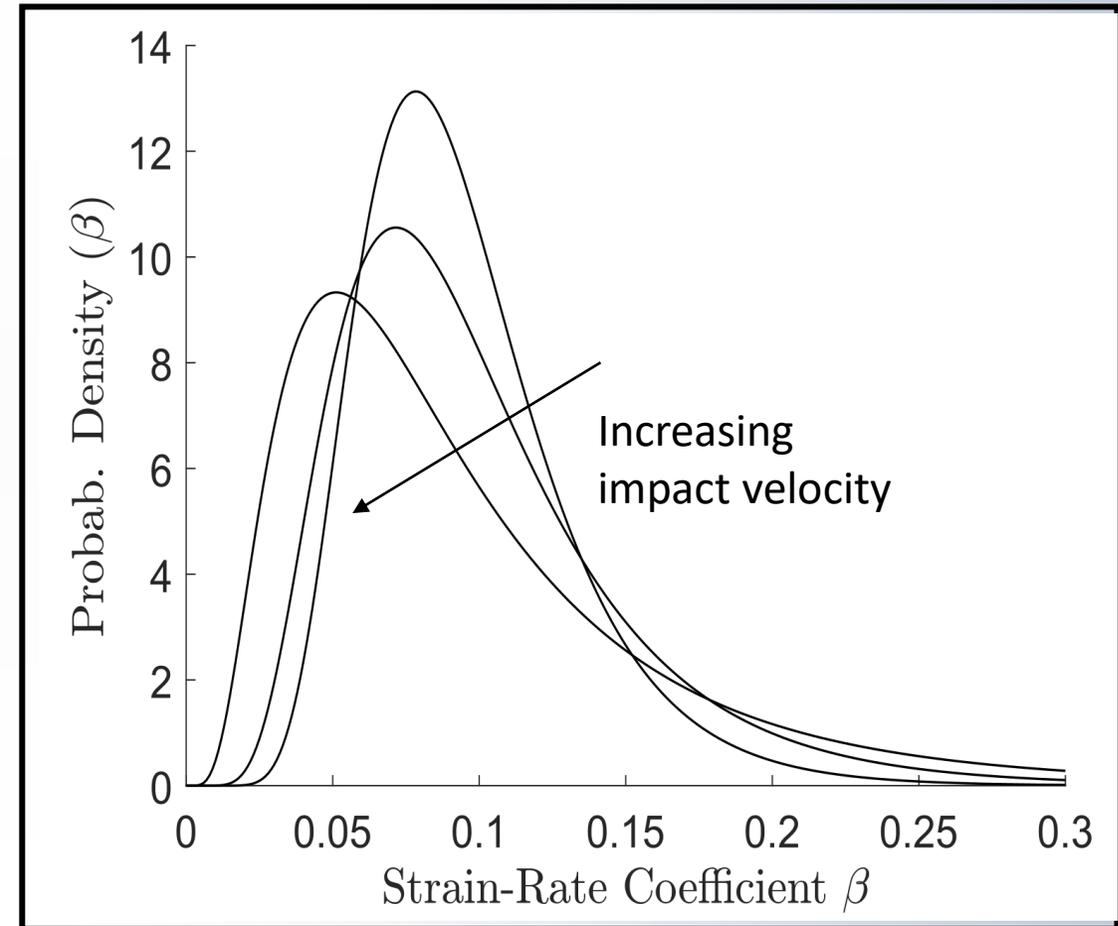


Methodology: Bayesian approach.

1st: Proposal distribution of $\mu_{(q_t-stat)}$:
from Soil unit weight

2nd: Proposal distribution of β

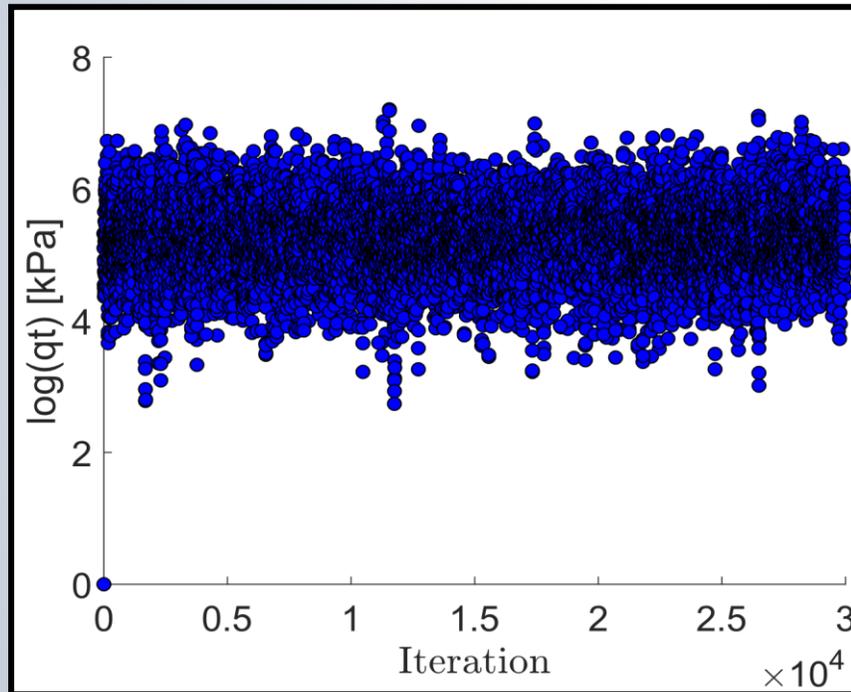
β (Impact velocity):
Lognormal distribution whose spread depend on
impact velocity.



3rd: $f(\mu_{(qt-stat)}, \beta) = 0$ - Hybrid Markov Chain Montecarlo simulation.

Generation of random $\mu_{(qt-stat)}, \beta$ samples from their respective proposal distribution

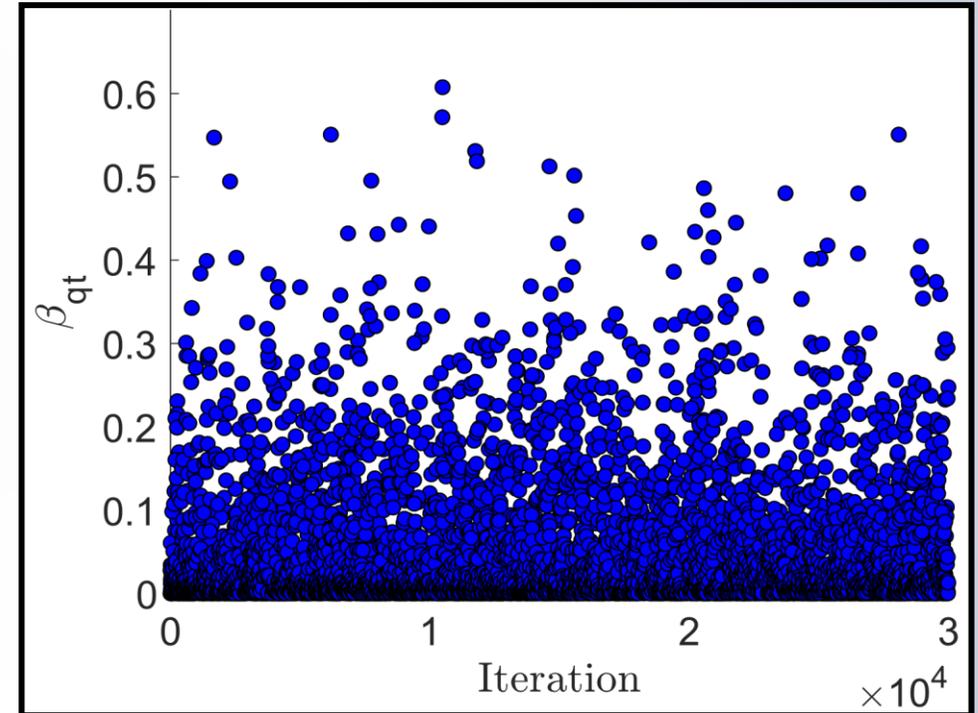
35000 *Equivalent samples*:



$$\sigma_{(\mu(qt))} = 0.7$$

$$\mu_{(\mu(qt))} = 5.41$$

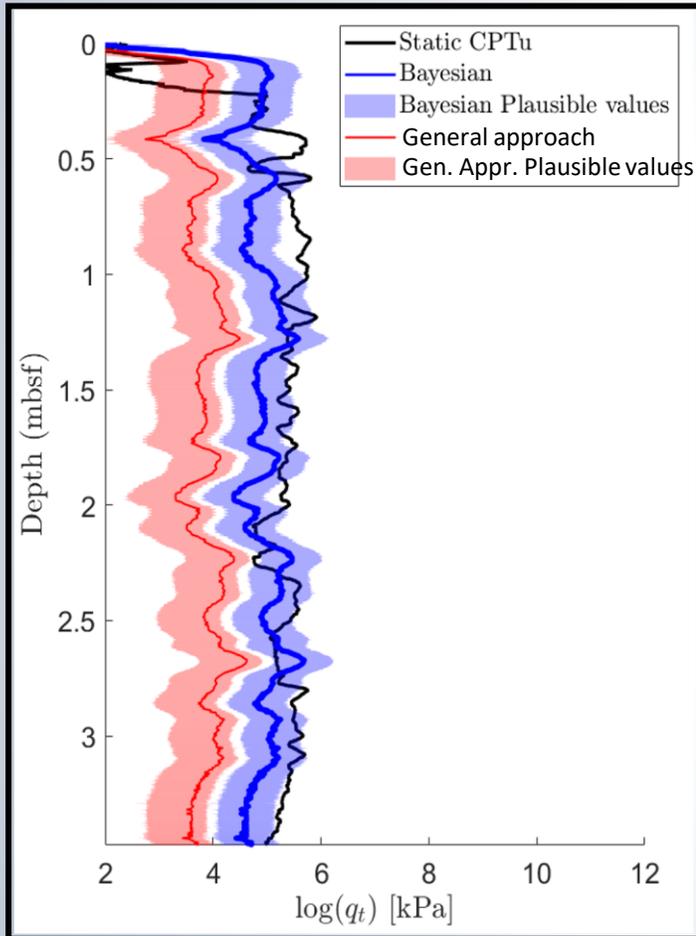
35000 *Equivalent samples*:



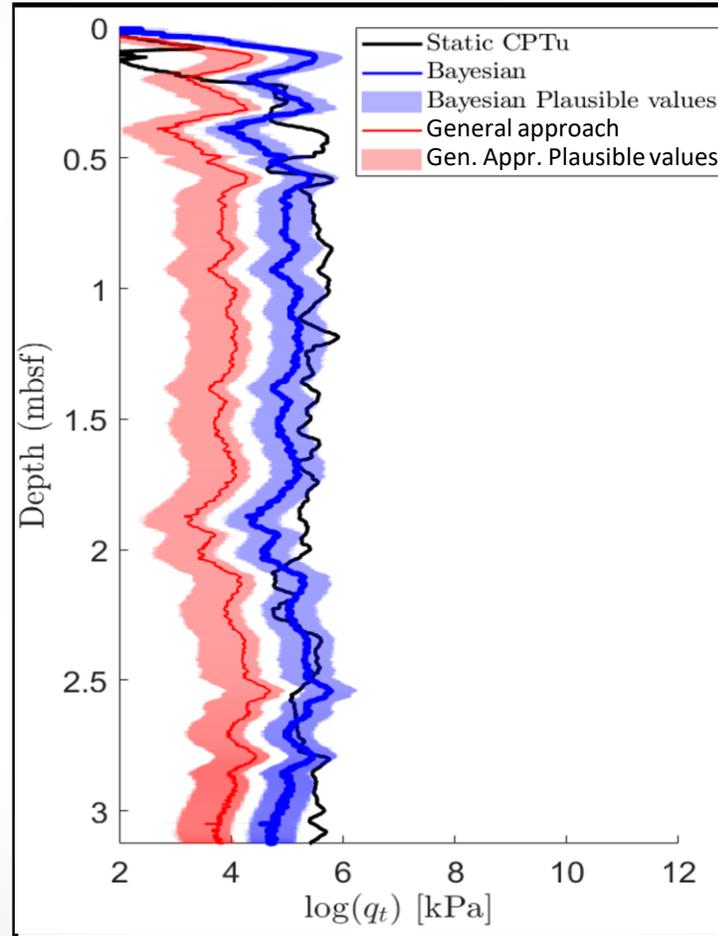
$$\sigma_{(\beta)} = 0.04$$

$$\mu_{(\beta)} = 0.03$$

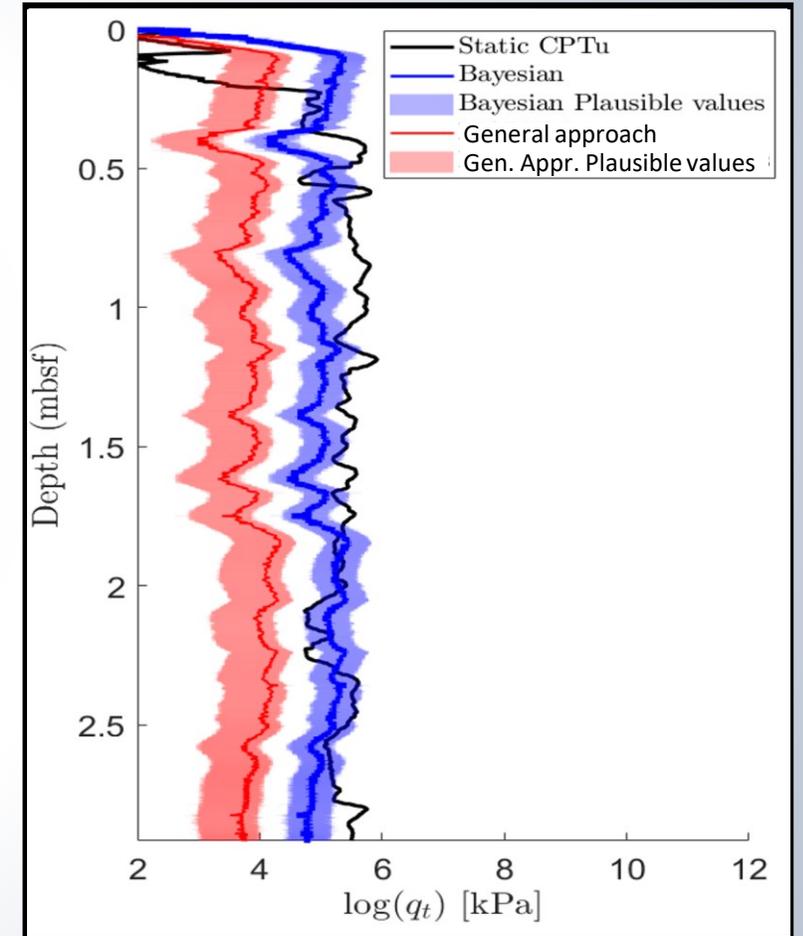
SW_16: Impact velocity 3 m/s



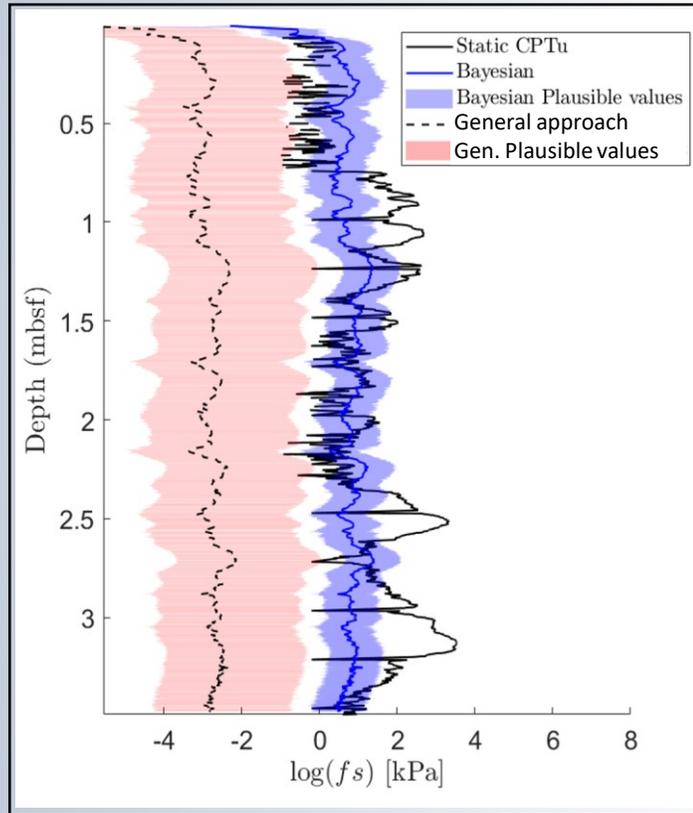
SW8: Impact velocity 1.1 m/s



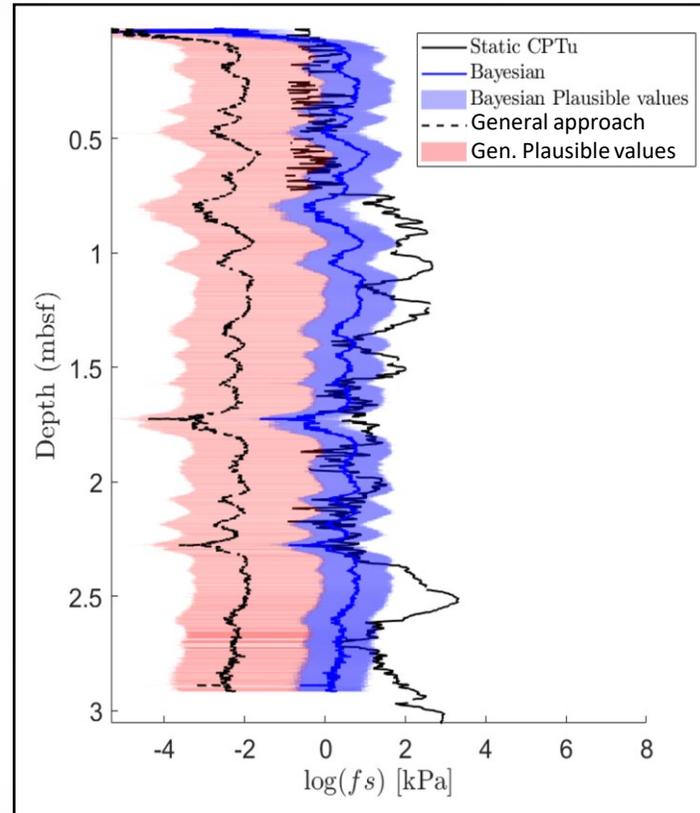
SW2: Impact velocity 0.5 m/s



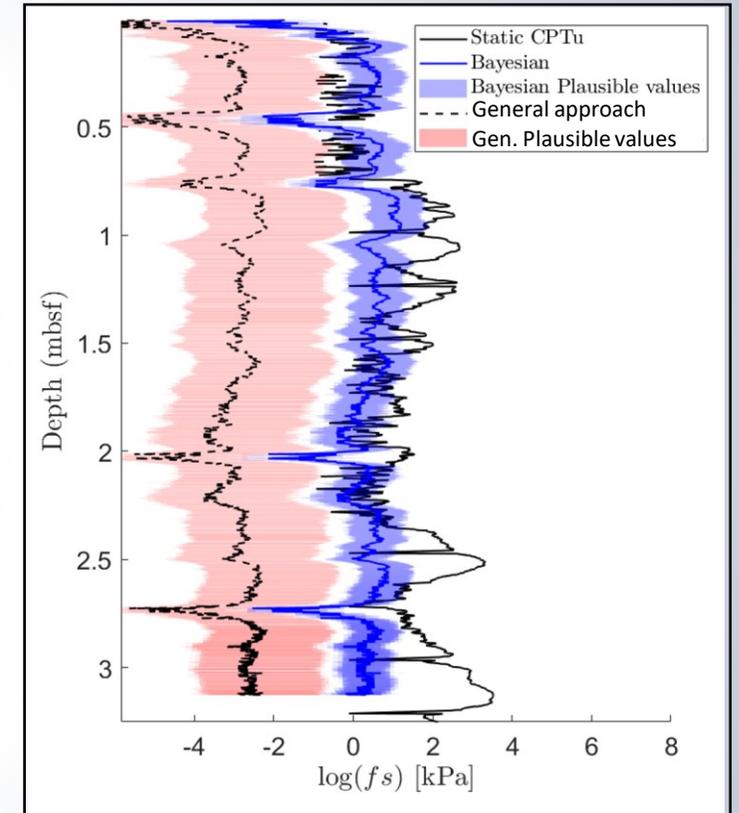
SW_16: Impact velocity 3 m/s



SW8: Impact velocity 1.1 m/s



SW2: Impact velocity 0.5 m/s



Conclusion:

- Good prediction of strain-rate coefficient β at different impact velocity from soil unit weight;
- The approach seems effective for simultaneously correction of dynamic cone tip resistance and sleeve friction;
- The approach is easily extendible for more complex model (e.g. adding overburden stress, drag force);

Possible development:

- Testing different strain-rate effect equations.
- Pore water pressure conversion.
- Direct conversion of undrained shear strength

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Acknowledgement

The first author's work has been supported by the funding from the European Union's Horizon 2020 research and innovation programme under the Marie-Sklodowska-Curie grant agreement No 721403.

