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CPTU-based settlement modelling for Kujala test embankments on overconsolidated silty clay: A probabilistic approach

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Abstract. Sounding resistance, such as CPTU cone tip resistance, is often used to estimate the settlement of the subsoil when designing ground-supported structures. This paper utilizes CPTU-based probabilistic modelling to estimate the settlement of two test embankments located in Finland. The model is based on an empirical correlation between CPTU net cone resistance and constrained modulus ($M_0$), described by factor $\alpha$. Besides the existing correlation (prior distribution of $\alpha$), this paper also considers the site-specific observations via Bayes’ theorem. By means of Monte Carlo simulation, the settlement histograms were produced and the results compared to the observed settlements. Lastly, the effect of dry crust on the embankment load transmission was considered in order to decrease the model bias. The results indicate that for the other embankment the effect of dry crust has to be considered in order to acquire satisfactory results. It is concluded that this CPTU-based settlement model might be suitable for stiff, overconsolidated silty clay soils in Finland. However, model calibration using site-specific oedometer test data is recommended. On the other hand, such simplified model with a singular modulus should not be used for normally consolidated clay subsoils.

1. Introduction

Recently, cone penetration tests with pore pressure measurement (CPTUs) are increasingly utilized in soil characterization. Accordingly, many researchers have suggested empirical correlations (i.e., transformation models) linking the constrained modulus to CPTU tip resistance (e.g., [1,2]). Since the oedometer tests on soil samples represent only a small portion of the subsoil, and may be subject to disturbance effects, the CPTU-based estimation of compressibility properties may provide more reliable and accurate results [3]. Despite its potential usefulness, CPTU-based settlement estimation is very rarely used in Finland. CPTU-based settlement modelling in Finnish soil conditions has been studied in a few master theses; for instance, according to Toikka’s [4] study the settlement estimates based on cone tip resistance were quite close to the observed settlements at four silty subsoil sites in Lahti, Finland.

This paper aims to evaluate the suitability of such CPTU-based settlement modelling for Kujala test embankments located in Lahti, Finland. At these test embankments, the observed settlements of the overconsolidated (OC) silty clay subsoil were very small, indicating that most of the compression occurs at OC state. Hence, a singular modulus ($M_0$) estimated from CPTU cone tip resistance could be used to acquire a preliminary estimate of the total settlement. However, since the empirical correlation between modulus and cone resistance is characterized by notable transformation uncertainty, a probabilistic
approach is applied in order to consider the underlying uncertainty in a more robust manner (e.g., Phoon [5]). By means of Monte Carlo simulations, this paper aims to evaluate the reliability and accuracy of this CPTU-based preliminary settlement modelling for Kujala test embankments. Lastly, the probabilistic model is utilized to calibrate the embankment load distribution altered by the thick dry crust topsoil.

2. Correlations for constrained modulus

Constrained modulus \( M \) is defined as the change in effective vertical stress \( \sigma' \) divided by the change in vertical strain \( \varepsilon_v \). In the case of clay compressibility, the constrained modulus is nonlinear and stress-dependent. In Swedish practice (e.g., Larsson [6]), the constrained modulus at in-situ stress state \( (\sigma'_0) \) is referred to as \( M_0 \) modulus. It is defined from the initial part of stress-modulus -curve from constant rate of strain (CRS) oedometer test. This modulus \( M_0 \) is used to calculate the strains during the stress increase from effective in-situ stress \( \sigma'_0 \) to the preconsolidation pressure \( \sigma'_p \). In the case of Scandinavian clays, such strain within the overconsolidated stress state is usually in the scale of 3-5 % (e.g., [6,7]).

The empirical correlation between in-situ stress constrained modulus \( M \) and CPT cone tip resistance has been studied by many researchers (e.g., [1–3,8–11]). In this transformation model, the constrained modulus can be taken as the in-situ (initial) modulus \( M_0 \) [12], thus giving equation (1) for the correlation:

\[
M_0 \approx \alpha (q_t - \sigma_{v0}) = \alpha q_{net}
\]  

(1)

In which \( q_t \) is the corrected cone tip resistance, \( \sigma_{v0} \) is the total in-situ stress, \( q_{net} \) is the net cone tip resistance and \( \alpha \) is the constant (correlation factor). However, it should be noted that some of the established correlations link the modulus to the measured cone tip resistance \( q_t \) instead of \( q_{net} \).

Various suggestions for the value of factor \( \alpha \) has been given for different soils. Mayne [2] collected a diverse group of soils ranging from sands, silts, organic clays and intact inorganic clays (see figure 3). Based on this summary, Mayne [2] suggested some representatives values for \( \alpha \) factor: \( \alpha \approx 5 \) for soft to firm "vanilla clays" and normally consolidated “hourglass sands”, \( \alpha \approx 1-2 \) for organic plastic Swedish clays and \( \alpha \approx 10-20 \) for cemented clay. In this data, \( \alpha = 28 \) is the maximum value (OC Ticino sand).

Regarding clays in Finland, Di Buò et al. [12] concluded (based on five clay sites in Finland) that \( \alpha = 5-10 \) captured majority of the data points. However, the authors emphasize that \( \alpha = 10 \) captured the mean trend of the sites that were characterized by the highest sample quality. However, it should be noted that the Finnish clays studied by Di Buò et al. [12] were all very soft; the range of reported \( q_{net} \) values was only 50-250 kPa. Toikka [4] studied the correlation with the measured cone tip resistance \( q_t \) for Lahti silty soil, and found the factor \( \alpha \) to vary between 2 and 10 with mean value around \( \alpha = 5 \). The tip resistance in Toikka’s data varied between 200 kPa and 5000 kPa.

Besides CPTU, other sounding methods have also been used to estimate the constrained modulus. In Swedish practice (e.g., Larsson [6]), the initial constrained modulus \( M_0 \) for clays is often estimated from empirical correlations utilizing undrained shear strength \( s_u \) \( (M_0 = 250 s_u) \) or preconsolidation pressure \( \sigma'_p \) \( (M_0 = 50 \sigma'_p) \). Other common transformation models include the correlation between \( N \) value of standard penetration test (SPT) and modulus, but this correlation for clays is not very strong (e.g., [11]).

3. Probabilistic settlement model

3.1. Dimensions and settlement monitoring at the test embankments

Kujala test embankments were built as a part of Kujala interchange construction works, a subproject related to Lahti Southern Ring Road highway project in Finland. This paper uses the term “test embankment” to refer to the test fills constructed during the geotechnical design phase. These two test fills, located 250 meters from each other, remained as the final ramp embankments connected to an interchange bridge. The remaining road structures adjacent to the test fills were constructed at the end of the monitoring period. In the following sections, the higher test embankment (test fill) will be referred to as “Kujala-5” and the lower as “Kujala-4”. 
At Kujala-5, the embankment height at the studied section was approximately 5.5 m, however the final embankment height varied to some extend due to changes in ground surface level. The crest width was 18.8 m, and the length of the embankment (before the adjacent embankment structures were finalized) was 22 m. The slopes and the slopes were built into 1:1.5 slope. At Kujala-4, the average height of the embankment was 3.8 m with 1:1.5 slopes. The crest width was 18 m and the length was 23.5 m. In the Monte Carlo simulations, the embankment heights were modelled as triangle distributions (see table 1). In both embankments, the fill material was crushed rock with 0-90 mm grain size. The unit weight of the fill was modelled as a normal distribution with mean value of 20 kN/m$^3$ and standard deviation SD=1 (corresponding to coefficient of variation COV=5%).

The settlements were monitored via settlement plates for 17 months. The settlements stabilized to 64-74 mm at Kujala-5 and 103-115 mm at Kujala-4 (see figure 1). The shapes of the time-settlement graphs imply that these values are rather close to final settlement values. Overall, the observed settlements were less than 120 mm, which would correspond to 4% strain in the case of three meters thick compressing soil layer. The thickness of silty clay subsoil is much greater than this at both sites (which would result in even smaller strain), so one can assume that most of the strains occur mostly in overconsolidated state. Thus, for this simplified settlement calculation, initial modulus $M_0$ can be used.

### Figure 1. Settlement plate measurements: Kujala-4 (left) and Kujala-5 (right).

#### 3.2. Field and laboratory investigations

Besides monitoring, extensive field and laboratory testing program was initiated. The undisturbed samples were tested at two university laboratories: Incrementally loaded (IL) oedometer tests were performed at Aalto University and CRS oedometer tests at Tampere University. In the clay-silt -layers below the dry crust, the water content is 30-90 %, and the liquid limit is approximately equal or slightly greater than the water content. The clay fraction is 30-70 %. The undrained shear strength is in the scale of 20-50 kPa and the overconsolidation ratio (OCR) is mostly in the scale of 1.5-2. The sensitivity is mostly around 10, but in some silt-rich layers, the sensitivity is as high as 60. In dry crust layers, the undrained shear strength is up to 200 kPa and OCR is 4-6.

Two CPTU soundings were performed at both test embankments (sounding points 7406 and 7401 at Kujala-5, and points 7407 and 7326 at Kujala-4). The measured (uncorrected) cone tips resistance $q_c$ and measured pore pressures $u_2$ together with estimated hydrostatic pressures are presented in figure 2a. Pore pressure piezometers were utilized in estimating the hydrostatic pressure. The depth of groundwater level was assumed to be 0.8 m and 1.0 m at Kujala-5 and Kujala-4, respectively (mode values table 1).

At Kujala-5 site, the dry crust layer seems to extend into a depth ($z$) of six meters (light grey area in figure 2a). Furthermore, based on the pore pressure measurement of the CPTU, there seems to be a highly permeable layer in $z=4.5-5.5$ m (darker grey area); this layer is probably silty soil with sand layers. On contrary, at Kujala-4, there is no such highly permeable middle layer. In addition, the dry crust is thinner ($z=4$ m). The depth of the silty clay basin in Kujala-4 is more shallow ($z=14$ m) than in Kujala-5 ($z=19$ m).

The initial modulus $M_0$ was estimated for the selected oedometer tests (figure 3). At Kujala-5 site, only one CRS test provided an estimate for $M_0$. Due to sample disturbance, the modulus curve was
almost a straight line for many silty samples. (In fact, a sample quality criteria based on the ratio of the peak initial modulus to the smallest modulus after the peak has been proposed due to this phenomena [7].) The incrementally loaded oedometer tests were of better quality, but the determination of $M_0$ is more unreliable due to the lack of continuous stress-strain curve. The value of factor $\alpha$ was calculated considering the net cone resistance at the same depth. Observed minimum $\alpha$ is 6 and maximum is $\alpha=23.5$. At Kujala-4, the initial modulus $M_0$ could be estimated for one oedometer test only; CRS test in the depth of $z=10.4$ m with $\alpha=4$. However, since the distance between the sites was only 250 m, one can assume that the $\alpha$ values defined for both Kujala sites represent the same “site-specific” transformation model.

![Figure 2a](image_url) Measured cone tip resistances $q_c$ and pore pressures $u_2$ at the test embankments.

![Figure 2b](image_url) Net cone resistance ($q_{net}$) and unit weight ($\gamma$) profiles used in the settlement calculation.

3.3. Probabilistic transformation model for constrained modulus (prior $\alpha$)

The probabilistic transformation model ($\alpha$ factor) for $M_0$ was defined first based on the data collected by Mayne [2]. This “prior distribution” for $\alpha$ was defined by visual fitting. A lognormal distribution with mean $\mu=9$ and standard deviation SD=5.5 (COV=61 %) was selected as the prior distribution based on its upper and lower 95 % confidence limits (2.55 and 23.17); as shown in figure 3, almost all
the prior data is within these limits (except for organic plastic clays, which were considered irrelevant for Kujala subsoil conditions). The first Monte Carlo simulations were performed using this prior $\alpha$.

However, when site-specific data is available, the distribution of global $\alpha$ values can be updated via Bayes’ theorem (e.g., [13]) to acquire a posterior distribution marked with less uncertainty. In other words, the prior distribution of $\alpha$ based on global data (Mayne [2]) was combined with the site-specific observations (observed $\alpha$ at both Kujala sites). The mean value of 12 site-specific $\alpha$ factors was $\mu=11.7$ (SD=5.1, COV=44%). The updated, posterior lognormal distribution for $\alpha$ has a mean value of $\mu=10.7$ and SD=1.23. This corresponds to a much smaller COV=11 %.

![Figure 3. Correlation between the net cone resistance and constrained modulus (data after Mayne [2]).](image)

![Figure 4. Definition of modulus $M_0$ for the Kujala-5 samples.](image)

3.4. Probabilistic settlement modelling (Monte Carlo simulation)

For the settlement analysis, one CPTU profile was selected for each test embankment (marked with black line in figure 2a). Only one was selected since there was very little difference between the two CPTUs at each site. At Kujala-4, point 7326 was right below the embankment and therefore selected. At Kujala-5, point 7401 was closer to two settlement plates with consistent measurements (64 mm and 66 mm) and thus selected. Figure 2b presents the net cone resistances and the unit weight profile of both embankments. The unit weigh values in dry crust layer were estimated from the water content of disturbed samples. The CPTU measurements were made every 2 cm, but for the analysis, the mean value of 10 cm section was used (“Model” in figure 2b). However, it should be noted that in the settlement analysis, the mean value of corrected cone resistance (instead of net value) was used in order to account for the uncertainty related to the total in-situ stress (or rather, the unit weight of soil).

The corrected cone resistance $q_t$ was modelled as a random variable as well via an error term $\varepsilon_{qt}$ applied to its measured value ($q_{t,measured}$) as defined in equation (2):

$$q_t = q_{t,measured} \varepsilon_{qt}$$  \hspace{1cm} (2)

The distribution for the error term $\varepsilon_{qt}$ was a lognormal distribution with mean value of 1 and COV = 8 %, based on the average COV observed by Phoon and Kulhawy [14] regarding the inherent variability of corrected cone tip resistance $q_t$. Parameters for all random variables are listed in table 1.

The inherent variability of subsoil unit weight was modelled in a similar manner, with normally distributed error term $\varepsilon_{\gamma}$ with COV = 5 % applied to the mean value (figure 2b, “Model”). The value of COV was selected as a cautious estimate based on observations by Löfman and Korkiala-Tanttu [15]: the range of inherent variability of unit weight of soil (for Finnish clays and silts) was COV = 1-4 %.
The Monte Carlo simulation was performed using the @RISK software, an add-in for Microsoft® Excel [16]. The adopted sampling method was Latin Hypercube sampling. The number of simulations was 50,000, and the random variables were assumed to be independent. The variance reduction due to averaging was not considered (variance reduction factor was assumed to be 1) because the settlements were calculated for each 100 mm section. In other words, the averaging length is less than the vertical autocorrelation length $\theta$ (for $q_t$, Phoon and Kulhawy [14] found $\theta=0.2-0.5 \text{ m}$); hence, the variance reduction factor can be taken as 1 (e.g., [17]).

### Table 1. Random variable parameters in the Monte Carlo simulation.

<table>
<thead>
<tr>
<th>Parameter description</th>
<th>Symbol</th>
<th>Distribution type</th>
<th>Min</th>
<th>Mode$^a$</th>
<th>Mean</th>
<th>Max</th>
<th>SD$^b$</th>
<th>COV$^c$</th>
</tr>
</thead>
<tbody>
<tr>
<td>Error term for cone resistance $q_t$</td>
<td>$\epsilon_{qt}$</td>
<td>Lognormal</td>
<td>0</td>
<td>0.991</td>
<td>1.0</td>
<td>$+\infty$</td>
<td>0.080</td>
<td>8.0 %</td>
</tr>
<tr>
<td>Net cone resistance $q_{net}$ of the top soil (kPa)</td>
<td>$q_{net_top_soil}$</td>
<td>Triangle</td>
<td>1500</td>
<td>3000</td>
<td>3000</td>
<td>4500</td>
<td>612.4</td>
<td>20 %</td>
</tr>
<tr>
<td>Error term for unit weight $\gamma$</td>
<td>$\epsilon_{\gamma}$</td>
<td>Normal</td>
<td>$-\infty$</td>
<td>1.0</td>
<td>1.0</td>
<td>$+\infty$</td>
<td>0.050</td>
<td>5.0 %</td>
</tr>
<tr>
<td>Factor $\alpha$ for modulus $M_0$</td>
<td>$\alpha_{prior}$</td>
<td>Lognormal</td>
<td>0</td>
<td>5.591</td>
<td>9.0</td>
<td>$+\infty$</td>
<td>5.500</td>
<td>61 %</td>
</tr>
<tr>
<td>Groundwater depth (m)</td>
<td>$GW_{Kujala-5}$</td>
<td>Triangle</td>
<td>0.6</td>
<td>0.8</td>
<td>0.867</td>
<td>1.2</td>
<td>0.125</td>
<td>14 %</td>
</tr>
<tr>
<td></td>
<td>$GW_{Kujala-4}$</td>
<td>Triangle</td>
<td>0.3</td>
<td>1.0</td>
<td>1.0</td>
<td>1.7</td>
<td>0.286</td>
<td>29 %</td>
</tr>
<tr>
<td>Unit weight of the fill material (kN/m$^3$)</td>
<td>$\gamma_{fill}$</td>
<td>Normal</td>
<td>$-\infty$</td>
<td>20</td>
<td>20</td>
<td>$+\infty$</td>
<td>1.000</td>
<td>5 %</td>
</tr>
<tr>
<td>Embankment height (m)</td>
<td>$H_{Kujala-5}$</td>
<td>Triangle</td>
<td>5.3</td>
<td>5.5</td>
<td>5.5</td>
<td>5.7</td>
<td>0.0817</td>
<td>1.5 %</td>
</tr>
<tr>
<td></td>
<td>$H_{Kujala-4}$</td>
<td>Triangle</td>
<td>3.6</td>
<td>3.8</td>
<td>3.8</td>
<td>4.0</td>
<td>0.0817</td>
<td>2.2 %</td>
</tr>
</tbody>
</table>

$^a$The value that appears most often (i.e., the most frequent value).

$^b$Standard deviation.

$^c$Standard deviation divided by the mean ($100\%$).

### 3.5. Modelling of the embankment load and the effect of dry crust

The first analyses (section 4.1) were conducted using the vertical load component of the embankment load ([18]) in the middle line of the embankment. Almost identical load distribution (in the middle of the embankment) is acquired by applying a rectangular vertical loading ([18]). In these first analyses, higher stiffness of the dry crust layer was considered via higher cone resistance ($q_{net\_top\_soil}$ in table 1), but its possible effects on the embankment load distribution was not modelled.

In sections 4.2 and 4.3 however, the effect of a stiffer dry crust layer on the load distribution is studied via Odemark’s [19] method, also known as the equivalent thickness method. In this method, the upper layer’s thickness ($h$) is transformed into an equivalent thickness which corresponds to the Young’s modulus $E_2$ of the deeper (less stiff) layer via equation (3) [19,20]:

$$h_e = \delta \left[\frac{E_2(1-\nu_2^2)}{E_1(1-\nu_1^2)}\right]^\frac{1}{2} h \approx 0.9 \left(\frac{E_1}{E_2}\right)^\frac{1}{3} h$$

where $E_i$ is the upper layer’s Young’s modulus ($E_i>E_2$), $\nu_1$ and $\nu_2$ are the Poisson’s ratios of upper and deeper layers, respectively. Correction factor $\delta$ can be taken as 0.9 [20] and if the Poisson’s ratios are assumed equal, the equation simplifies into the latter form, which was used in the analyses. The equivalent thickness $h_e$ is then used to define the modified depth $z'$ according to equations (4a) and (4b) [20]:

$$z' = \frac{h_e z}{h}, \text{ when } z \leq h$$  \hspace{1cm} (4a)

$$z' = \frac{h_e}{h} z + h - h, \text{ when } z > h$$  \hspace{1cm} (4b)
The modified depth $z'$ is then used to calculate the embankment load (modelled as rectangular vertical loading) in order to acquire the “corrected” vertical load at real depth $z$ from the ground surface. Examples of this correction are presented in section 4.2.

For Kujala-5, the selected value for upper layer thickness was $h = 6$ m (see figure 2a). Besides the dry crust clay layer, this thickness includes the permeable (and probably stiff) sandy silt intermediate layer as well. In the case of Kujala-4, $h = 4$ m was estimated to be suitable. Since there is no CPTU data for the uppermost top soil, the ratio $E_1/E_2$ was varied. Nevertheless, the possible range for this ratio can be estimated; the net cone resistance in the deeper layers is in the scale of 500 kPa (figure 2b), and the upper layer’s net cone resistance was estimated to lie within the range 1500-4500 kPa (table 1), leading to probable range of $E_1/E_2 = 3.9$.

4. Results

4.1. Probabilistic settlement analysis with embankment load

The Monte Carlo simulation was performed with both prior $\alpha$ and updated, posterior $\alpha$. According to the produced histograms (probability densities in figures 5 and 6), using the posterior distribution of $\alpha$ (instead of prior) decreases the uncertainty in the settlement prediction greatly (i.e., standard deviation decreases). In the case of Kujala-4, the uncertainty is smaller than at Kujala-5 in both analyses (prior and posterior $\alpha$); this difference is explained by the fact that the compressing soil layer is thicker in Kujala-5, thus accumulating more sources of uncertainty to the total settlement.

![Figure 5. Settlement histogram for Kujala-5 using prior $\alpha$ (left) and posterior $\alpha$ (right).](image)

![Figure 6. Settlement histogram for Kujala-4 using prior $\alpha$ (left) and posterior $\alpha$ (right).](image)

In the case of Kujala-4 (figure 6) the calculated and observed settlements are in good agreement in both analyses. In Kujala-5 however, the observed settlement is significantly lower than estimated (outside of...
The effect of dry crust stiffness was studied by varying the ratio $E_1/E_2$. In this analysis, the settlement calculation was deterministic, i.e., the analysis was performed using constant parameters. For each random variable, the most probable (mode) value was used (table 1). For factor $\alpha$, the posterior value (mode = 10.52) was used. In addition, the settlement was estimated using the highest value $\alpha=27$ in the prior data (figure 3). The results of these deterministic analyses and examples of the final stresses are presented in figure 7. In the bias calculation, the observations made at the settlement plates situated closest to the selected CPTU profile were used (111 mm for Kujala-4 and 66 mm for Kujala-5). The model bias was defined as the calculated settlement divided by the observed value.

For Kujala-5, the model bias (with $\alpha=10.5$) with converges closer to the no-bias-line as the stiffness ratio $E_1/E_2$ increases. When $E_1/E_2=2$, model bias is approximately 3, whereas increase in stiffness ratio to $E_1/E_2=10$ decreases the model bias to 2. In other words, the accuracy of the model increases when higher dry crust stiffness is assumed. However, it should be noted that ratios $E_1/E_2$ much greater than 9 might not be applicable in actual in-situ conditions due to too large tensile forces [20]. When the maximum value of $\alpha$ is used ($\alpha=27$), the no-bias-line is met at $E_1/E_2=6-9$. Since this stiffness ratio might be more realistic, the actual $\alpha$ that applies to Kujala-5 might be greater than the posterior ($c$).

For Kujala-4, the model bias decreases below 1 quickly as the stiffness ratio is increased. The initial result (with posterior $\alpha=10.5$), without any dry crust correction, was already in agreement with the settlement observations (figure 6). Hence, for Kujala-4, taking into account the dry crust stiffness does not increase the model accuracy.

4.3. Calibrated settlement model for Kujala-5

Without the dry crust effect, the observed settlement for Kujala-5 is outside of 90 % confidence limits even when the maximum prior value of $\alpha=27$ is adopted (figure 8). In other words, it would very unlikely that such small settlements could be caused by the full embankment load. Consequently, in the case of Kujala-5, the effect of dry crust to the embankment load distribution has to be considered.

Therefore, the Monte Carlo simulation is performed again with constant $\alpha$ and with the effect of dry crust. For $\alpha$, the greatest site-specific value observed was adopted ($\alpha=23.5$). Similarly, for the dry crust
effect, the greatest realistic value was used \((E_1/E_2=9)\). Finally, after this calibration, the agreement between observed and calculated is satisfactory (figure 9).

![Figure 8. Settlement histogram for Kujala-5 with constant \(α=27\), without dry crust effect.](image1)

![Figure 9. Settlement histogram for Kujala-5 with constant \(α=23.5\) and dry crust effect \((E_1/E_2=9)\).](image2)

4. Discussion

The CPTU-based settlement prediction performs very differently in these two test embankments. The possible explaining factors for this difference are: (i) the compressing clay-silt basin is thicker in Kujala-5, (ii) the dry crust is thicker and probably stiffer in Kujala-5 (effect of dry crust is greater), and (iii) in Kujala-4, some of the compression occurs in normally consolidated state.

Regarding the last point, the field and laboratory test results (figure 7, right) indicate that the preconsolidation pressure in Kujala-4 profile is indeed smaller than in Kujala-5. Given that the final stress in Kujala-4 is probably also greater (due to thinner dry crust), it seems probable that initial modulus \(M_0\) cannot model all the compression at Kujala-4 (since the final stress is probably larger than preconsolidation pressure). The authors expect that the good agreement for Kujala-4 in the first analyses was partly caused by the usage of inaccurate \(α\). Based on Kujala-5 analyses (section 4.3), more accurate value for \(α\) would be closer to the greatest observed value \((α=23.5)\). The “incorrectness” of posterior \(α\) may have been caused by the bias created by sample disturbance. However, it should be noted that the actual final stress could also be smaller than estimated here (in section 4.3) since the dry crust effect required assumptions that cannot be verified (such as the stiffness ratio \(E_1/E_2\)).

5. Conclusions

This paper defined a probabilistic settlement estimation model based on an empirical correlation between CPTU net cone resistance and constrained modulus \(M_0\) (described by factor \(α\)). The transformation uncertainty related to factor \(α\) was quantified for both the existing global correlation (prior \(α\)) and for the updated transformation model (posterior \(α\)) which considered also the site-specific oedometer test results. By means of Monte Carlo simulation, settlement histograms for the two test embankments were produced and the results compared to observed settlements. Lastly, the model for the for the higher embankment, Kujala-5, was calibrated to account for the stiff dry crust’s effect.

The results demonstrate that for Kujala-5, the effect of dry crust stiffness on the embankment load has to be considered. In addition, the actual factor \(α\) seems to be higher than average. For Kujala-5, using the maximum observed value \((α=23.5)\) together with the dry crust effect led to the smallest model bias. For the lower embankment (Kujala-4), the posterior distribution for \(α\) produced rather good results (without any calibrations). However, this result may be caused by different subsoil conditions and sample disturbance bias rather than better suitability of the model.

It can be concluded that this CPTU-based preliminary settlement calculation might perform rather well at overconsolidated (OC) silty soils in which the compression occurs at OC state (i.e., the final
stress does not surpass the preconsolidation pressure). However, it is advisable to calibrate the value of factor $\alpha$ by means of site-specific oedometer test. Finally, it should be emphasized that such singular modulus-based settlement model should not be applied to a normally consolidated clay subsoil.

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