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An integrated shear-wave velocity model for the Groningen gas field, The Netherlands

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Abstract A regional shear-wave velocity (\(V_S\)) model has been developed for the Groningen gas field in the Netherlands as the basis for seismic microzonation of an area of more than 1000 km\(^2\). The \(V_S\) model, extending to a depth of almost 1 km, is an essential input to the modelling of hazard and risk due to induced earthquakes in the region. The detailed \(V_S\) profiles are constructed from a novel combination of three data sets covering different, partially overlapping depth ranges. The uppermost 50 m of the \(V_S\) profiles are obtained from a high-resolution geological model with representative \(V_S\) values assigned to the sediments. Field measurements of \(V_S\) were used to derive representative \(V_S\) values for the different types of sediments. The profiles from 50 to 120 m are obtained from inversion of surface waves recorded (as noise) during deep seismic reflection profiling of the gas reservoir. The deepest part of the profiles is obtained from sonic logging and \(V_P\)–\(V_S\) relationships based on measurements in deep boreholes. Criteria were established for the splicing of the three portions to generate continuous models over the entire depth range for use in site response calculations, for which an elastic half-space is assumed to exist below a

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clear stratigraphic boundary and impedance contrast encountered at about 800 m depth. In order to facilitate fully probabilistic site response analyses, a scheme for the randomisation of the $V_S$ profiles is implemented.

**Keywords** Shear-wave velocity · Site response analysis · Geology · Randomisation · Surface-wave inversion · Microzonation

## 1 Introduction

The province of Groningen in the Netherlands (Fig. 1) is experiencing induced earthquakes due to the exploitation of a large onshore gas field. The largest induced earthquake to date was the Huizinge event of August 2012 with a local magnitude $M_L$ of 3.6 (moment magnitude $M$ = 3.4). This earthquake initiated the development of a comprehensive probabilistic seismic hazard and risk model for the region (Bourne et al. 2015), covering an area of about $35 \times 45$ km.

Rather than using proxy parameters such as the time-averaged 30 m shear-wave velocity, $V_{S30}$, to capture site effects, the aim has been to more faithfully model the dynamic effects of the specific profiles in the field, which are overlain by a particularly thick deposit of soft soils. To this end, a seismic microzonation of the field has been developed, the starting point of which is the model of shear-wave velocity ($V_S$) profiles described herein. The field-wide $V_S$ profiles are subsequently used to define a reference rock horizon—at a depth of about 800 m—and then used in a large number of site response analyses to obtain nonlinear frequency-dependent amplification factors, as described in Rodriguez-Marek et al. (2017). The final microzonation defines 161 zones. Bommer et al. (2017) explain the derivation of ground-motion prediction equations (GMPEs) for

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**Fig. 1** Location of the Groningen gas field (in red) in the northern part of the Netherlands
accelerations at the deep rock horizon and the process for combining the predicted rock motions with the nonlinear site amplification factors.

The work presented herein adds to the body of shared knowledge and experience of seismic microzonation in Europe, which includes, among many others, the EUROSEISTEST project in Volvi, Greece (Pitilakis et al. 1999), the characterisation of the Grenoble basin in France (Guéguen et al. 2007) and the microzonation of the city of Basel in Switzerland (Havenith et al. 2007). Each of these projects has made use of different types of data and measurements. The Groningen microzonation project marks an important contribution to the field since it uses innovative approaches. The project is also of broad interest since it applies to a much larger area than has been the focus for previous seismic microzonation projects. Another factor that makes the project relevant is that it corresponds to a region that is effectively aseismic in terms of tectonic earthquakes—the motivation for development of the Groningen microzonation is entirely due to induced seismicity.

This paper describes the development of the \( V_S \) model for the Groningen field. The \( V_S \) profiles have been constructed using a unique combination of \( V_S \) models over three separate depth ranges that collectively cover the full range down to about 800 m. The shallowest depth range, to 50 m below Dutch Ordnance Datum (NAP, which is approximately Mean Sea Level), is based on the high resolution 3D geological model GeoTOP (Stafleu et al. 2011; Stafleu and Dubelaar 2016), combined with representative \( V_S \) distributions with depth for the sediments. The \( V_S \) model for the intermediate depth range, from NAP-50 m to \( \sim \) NAP-120 m, comes from the inversion of Rayleigh waves data (surface waves) using the Modal Elastic Inversion method (Ernst 2013) on the extensive reflection seismic survey data. The \( V_S \) model for the third depth range, starting at \( \sim \) NAP-70 m to about NAP-800 m is based on the PreStack Depth Migration (PSDM) velocity model derived from sonic logs that is used to image the reservoir. These three models, all with different spatial resolutions, are spliced to obtain one \( V_S \) model over the full depth range.

In order to obtain fully probabilistic estimates of the ground shaking hazard at the surface, the site amplification characteristics are modelled in a probabilistic framework (Bazzurro and Cornell 2004a, b). Therefore, \( V_S \) profiles were constructed by sampling from statistical distributions of \( V_S \) that were obtained from site-specific field measurements. The resulting randomized \( V_S \) profiles capture the variability and uncertainty in the amplification factors. The randomisation scheme for \( V_S \) includes confining stress dependent relations and correlations within and between units in the profile.

This paper first gives a short description of the geological setting. This is followed by a detailed description of the three different \( V_S \) models. The subsequent sections elaborate on the construction of \( V_S \) profiles, including splicing of the three \( V_S \) models, layering and randomisation. Thereafter, the zonation of the field necessary for the aggregation of the probabilistic amplification factors is described. For illustration, the shallow \( V_S \) model is used to construct a \( V_{S30} \) map. The paper concludes with a discussion of potential improvements to the \( V_S \) model that may be addressed in future work.

2 Geological setting

The area of Groningen consists of a mainly flat, low-lying area with an average surface level close to mean sea level. The geological history of the region includes the deposition of fluvial braid plain sands, ice sheet loading, erosion during Pleistocene ice ages, and an
infill of Holocene shallow marine (intertidal) and terrestrial deposits consisting of soft clays, sands and the formation of peat. As a result, a thick layer of unconsolidated deposits (over 800 m thick) with a large degree of heterogeneity is present over the entire Groningen region. The Cretaceous limestones of the Chalk Group were selected as the reference horizon for site response analyses (Rodriguez-Marek et al. 2017). Hence the formations considered for the site response analysis are the Paleogene, Neogene and younger deposits overlying the Chalk Group (de Mulder et al. 2003; Vos 2015). The lowermost deposits are the Lower North Sea Group, which consists of an alternation of primarily marine grey sands and sandstones and clays of Late Paleocene to Middle Eocene age. On average, the base of the Lower North Sea Group is situated at a depth of 840 m. The predominantly clayey marine formations of the Oligocene (Middle North Sea Group) are found between about 450 and 350 m depth, corresponding to the base of the overlying marine deposits of the Breda Formation (Miocene). This also corresponds to the base of the Upper North Sea Group. The Breda Formation consists of open marine clays, sandy clays and loam. The overlying sediments belong to the Oosterhout Formation (Pliocene), consisting of marine delta slope deposits of clay, fine sand and loam. The combination of the Lower, Middle and Upper North Sea Group is also known as the North Sea Supergroup (https://www.dinoloket.nl/en/nomenclature-deep). Stratigraphically, the overlying Pleistocene and Holocene formations belong to the North Sea Supergroup. The deposits of these formations are described separately below, owing to their heterogeneity, which is reflected in the variation in their geomechanical properties.

The uppermost deposits of Pleistocene and Holocene age are influenced by the last three ice ages and associated sea level fluctuations. During the Pleistocene, the sedimentary sequence is characterized by a succession of fluvial (Peize Formation, Appelscha Formation and Urk Formation), glacial (Peelo and Drente Formation) and marine deposits (Eem Formation) following the penultimate glacial stage. The Elsterian glaciation produced deep subglacial features (‘tunnel valleys’), which were filled with sands and clays of the Peelo Formation and were buried by younger sediments. The second glaciation, the Drente Substage of the Saalian glacial, produced the till sheet that constitutes the Drente plateau. The ridge-and-valley topography is still present in the landscape stretching from the city of Groningen towards the South-East (Hondsrug). The region was not covered by ice-sheets during the last ice-age (Weichselian). During this period, a widespread superficial blanket of eolian sand formed that in many places marks the top of the Pleistocene deposits (the so-called cover sands). The northern part of the Netherlands borders the North Sea. During interglacial periods, a large part of Groningen became a coastal plain. The Holocene succession consists of alternations of shallow marine intertidal deposits (Naaldwijk Formation) and peat (Nieuwkoop Formation). Two distinct peat layers can be recognized: Basal peat and Holland peat. The Holocene succession reaches several tens of metres in the north (maximum thickness of 28 m in the far north) and is absent in the south of the region. A representative geological cross section of the top 25 m is shown in Fig. 2. This section clearly shows that the subsurface displays a large degree of heterogeneity due to the channel sands and the clayey deposits in the intertidal basin. An effort was made to combine all available geological and geomechanical data into a stochastic model that captures the spatial variability of these deposits. This is because the vertical position, thickness, lateral extent of soft layers (i.e. clay and peat) and the impedance contrasts at layer boundaries are dominant factors in site amplification.
3 Description of \( V_S \) models

The integrated model of \( V_S \) from the surface to the base of the North Sea Supergroup is a combination of three different \( V_S \) models, each with its own depth range. The top part of the model ranges from the surface to NAP-50 m and is constructed from the combination of the high resolution 3D geological voxel model GeoTOP constructed by TNO—Geological Survey of the Netherlands and Groningen specific \( V_S \) data. The \( V_S \) model for this depth range has been derived from seismic cone penetration tests (SCPT) that were linked to the geological units of the GeoTOP model. This \( V_S \) model is referred to as the GeoTOP \( V_S \) model.

The intermediate depth interval ranges from approximately NAP-40 m to NAP-120 m. Extensive reflection seismic surveys were conducted in the 1980s for imaging purposes of the reservoir. The legacy data were reprocessed using surface waves information to retrieve a \( V_S \) model based on the Modal Elastic Inversion (MEI) method (Ernst 2013). This model is referred to as the MEI \( V_S \) model.

The deepest depth interval ranges from approximately NAP-70 m to the base of the North Sea Supergroup, providing overlap with the MEI \( V_S \) model. The \( V_S \) model for this depth range is based on the pre-stack-depth-migration model (PSDM) of compression-wave velocity (\( V_P \)) used to image the reservoir. The \( V_P \) model is based on 70 sonic logs and well markers in 500 wells. The \( V_P \) model is converted to a \( V_S \) model using
relationships for the $V_p/V_s$ ratio based on Groningen-specific data. This model is referred to as the Sonic $V_s$ model. The following sections describe each of these models in more detail.

### 3.1 GeoTOP $V_s$ model

GeoTOP describes the subsurface in voxels measuring 100 by 100 by 0.5 m (x, y, z) to a maximum depth of NAP-50 m. The model provides estimates of stratigraphy and lithology, including sand grain-size classes. The estimates are calculated using Sequential Gaussian Simulation (SGS) and Sequential Indicator Simulation (SIS) (Goovaerts 1997; Chilès and Delfiner 2012). These stochastic techniques allow the construction of multiple, equally probable 3D subsurface models as well as the evaluation of model uncertainty (Stafleu et al. 2011). The “most likely” subsurface model was determined from the multiple subsurface models, using the averaging technique described by Soares (1992). This “most likely” model is used in the construction of the GeoTOP $V_s$ model. The GeoTOP model (version 1.3) is publicly available at [https://www.dinoloket.nl/en/subsurface-models](https://www.dinoloket.nl/en/subsurface-models) (Stafleu and Dubelaar 2016).

The GeoTOP model of the north-eastern part of the Netherlands, including the Groningen region, was constructed using some 42,700 digital borehole descriptions from DINO, the national Dutch subsurface database operated by the Geological Survey. The largest part of these boreholes consists of manually-drilled auger holes collected by the Geological Survey during the 1:50,000 geological mapping campaigns. Most of the other borehole data comes from external parties such as groundwater companies and municipalities. Because of the large share of manually-drilled boreholes, borehole density decreases rapidly with depth.

An example of a cross section through the GeoTOP model is presented in Fig. 3, showing the stratigraphic units in the top panel and the lithological classes in the bottom panel. From top to bottom there are in this example Holocene Naaldwijk clays, the Holland and Basal Peat, sands of the Boxtel Formation and clays of the Peelo Formation.

The GeoTOP $V_s$ model associates each of the voxels of the GeoTOP model to a $V_s$ value. The various constituents of the Groningen subsoil have different geological histories, as described above, and consequently have different geomechanical characteristics. A Holocene clay will have a different $V_s$ than a Pleistocene clay that has experienced loading by ice sheets. Therefore, different $V_s$ statistical distributions were derived for each of the stratigraphic and lithological combinations that are found in the Groningen field. In the following, the combination of stratigraphy and lithology is referred to as “unit”.

A data set of 88 SCPTs in Groningen provided the input for these statistical distributions. The $V_s$ measurements from the SCPTs at each depth were associated with a stratigraphy, inferred from GeoTOP, and lithological class, inferred from cone resistance and friction ratio of the accompanying cone penetration test (CPT). The effective isotropic confining stress, $\sigma_0$ (i.e., the average of the vertical and two horizontal components of the effective stress) for each depth is computed using the unit weight of the overlying sediments and assuming a mean water table of 1 m below the surface. For brevity, the effective isotropic confining stress is hereafter simply referred to as “confining stress”. Next, $V_s$ values from the SCPTs were clustered for each unit and $\ln V_s$ was plotted versus the natural logarithm of the confining stress. The clustered $V_s$ values were then used to develop models to assign $V_s$ values to each of the GeoTOP voxel-stacks. A voxel-stack is a vertical sequence of voxels at a particular (x,y)-location in the GeoTOP grid. This approach
enhances the inclusion of geological information in the $V_S$ profiles generated for this depth range.

Generally, $V_S$ increases with confining stress (e.g. Hardin 1978; Jamiolkowski et al. 1991; Yamada et al. 2008). Therefore, we checked for confining stress dependence within each group of $V_S$ data. A typical model for $V_S$ dependence on confining stress is:

$$\ln V_S = \ln V_S^1 + n \ln \left( \frac{\sigma_o}{P_a} \right)$$

(1)

where $\sigma_o$ is the confining stress, $P_a$ is atmospheric pressure, $\ln V_S^1$ is a parameter that represents the shear-wave velocity at a confining stress equal to one atmosphere, and $n$ is the slope that defines confining stress dependence (Sykora 1987). The parameters $n$ and $\ln V_S^1$ and their statistics were determined for each unit. Shear-wave velocity values are assumed to be log-normally distributed; hence the $\ln$-mean and the standard deviation fully define the distribution. The development of the confining stress-dependent $V_S$ models considered three different cases. The first case is when the confining stress-dependence is fully defined by the SCPT data. A second case is for units that do not show confining stress dependence. The last case is for units where the data is insufficient to define the confining stress-dependence, but such dependence is to be expected based on analogy to similar

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**Fig. 3** Cross section through a small part of the GeoTOP model, showing the stratigraphic units (top) and lithological class (bottom)
sediments elsewhere in the field. For these units, the parameters of Eq. 1 are based on existing literature and expert judgment.

Two examples of $V_S$ data for typical units with numerous observations are shown in Fig. 4. The mean $V_S$ at a certain confining stress is described by the slope $n$ and the intercept $\ln V_S$, while the standard deviation depends on the number of observations, mean $\ln(\sigma_0/p_a)$, the sum of squares of $\ln(\sigma_0/p_a)$ and the total variance of $V_S$ (Montgomery et al. 2011). As indicated previously, the $V_S$ profiles also need to be randomised (described in Sect. 4.1), for which the above described mean $V_S$ and standard deviation will be used. There were sufficient observations (a minimum of 20) for 10 units to derive a confining stress-dependent relation based on SCPT data. The parameters describing the confining stress dependence defined by the SCPT data are given in Table 1. The values of the slope $n$ for Groningen clay (including sandy clay and clayey sand) range from 0.18 to 0.43 with an average of 0.28. This compares well with literature values for clay, which are generally given as $n = 0.25$ (Hardin 1978; Jamiolkowski et al. 1991; Yamada et al. 2008). The slope $n$ depends on the type of sediment (Fig. 4): clays of the Peelo Formation (Pleistocene glacial deposits) have a stronger confining stress dependence (larger $n$) than clays of the Naaldwijk Formation (Holocene tidal deposits) which is generally present at much shallower depths and thus lower confining stresses.

For several units, confining stress dependence is not apparent in the SCPT data, showing an $n$ close to 0 or even slightly negative. Figure 5 shows $V_S$ data for medium sand from the Boxtel Formation. Since the slope in this case is very close to 0 (0.07), no confining stress dependence was imposed for this unit. In some other cases, the geological history implies that confining stress dependence is not expected. For example, the clay from the Drente Formation formed under varying glacial conditions and the effect of spatially varying loading is much larger than the confining stress dependence. The distributions of these constant $V_S$ units are defined by the mean and standard deviations of $\ln V_S$. These are summarised in Table 2. A minimum standard deviation of 0.2 is imposed. This lower limit was used in the past as measurement uncertainty in $V_S$ profiles (Coppersmith et al. 2014). A standard deviation 0.27 is imposed on all peats, based on the observations from the SCPT data set for Nieuwkoop Holland Peat.

The last class of $V_S$ consists of units for which confining stress dependence of $V_S$ is to be expected, but there are not enough data in the SCPT data set to constrain this

Fig. 4 Example of numerous $V_S$ observations in the SCPT data set, for clays from the Peelo Formation (left) and Naaldwijk Formation (right). The solid line describes the regression while dotted lines indicate 95% confidence intervals.
relationship. In that case, we estimate $n$ from literature. We use $n = 0.25$ for clay, for all lithoclasses within Nieuwkoop Basal Peat and for peats within Pleistocene Formations following Hardin (1978), Jamiołkowski et al. (1991) and Yamada et al. (2008). For sand, we use measured coefficients of uniformity $C_u$ from Groningen to estimate $n$ using Menq (2003). This results in values for $n$ varying between 0.25 and 0.29. In this case, no average

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**Table 1** Look-up table summarising parameters for the confining stress dependent $V_S$ (Eq. 1) from SCPTs

| Formation     | Lithological class                        | # Observations | Slope $n$ | Intercept $V_{S1}$ (m/s) | Mean $\ln\left(\sigma_0/\rho_u\right)$ | Sum of squares $\ln\left(\sigma_0/\rho_u\right)$ | Total variance of $\ln(V_S)$ |
|---------------|------------------------------------------|---------------|-----------|--------------------------|----------------------------------------|-------------------------------------------------|-----------------------------|
| Boxtel        | Sandy clay and clayey sand               | 43            | 0.20      | 217.0                    | 0.10                                    | 5.67                                            | 0.039                         |
| Boxtel        | Fine sand                                | 260           | 0.11      | 247.2                    | -0.057                                  | 64.41                                           | 0.054                         |
| Naaldwijk     | Clay                                     | 303           | 0.18      | 135.6                    | -1.20                                   | 107.49                                          | 0.11                          |
| Naaldwijk     | Sandy clay and clayey sand               | 245           | 0.28      | 190.6                    | -0.98                                   | 59.65                                           | 0.066                         |
| Naaldwijk     | Fine sand                                | 166           | 0.36      | 247.2                    | -0.78                                   | 34.05                                           | 0.099                         |
| Nieuwkoop Basal Peat | Peat                                 | 22            | 0.57      | 156.0                    | -0.77                                   | 3.70                                            | 0.19                          |
| Peelo         | Clay                                     | 455           | 0.33      | 194.4                    | 0.39                                    | 41.89                                           | 0.033                         |
| Peelo         | Sandy clay and clayey sand               | 41            | 0.43      | 181.3                    | 0.66                                    | 2.59                                            | 0.033                         |
| Peelo         | Fine sand                                | 222           | 0.10      | 265.1                    | 0.54                                    | 16.26                                           | 0.022                         |
| Peelo         | Medium and coarse sand, gravel and shells | 72            | 0.24      | 265.1                    | 0.61                                    | 3.04                                            | 0.020                         |

During random sampling, a minimum standard deviation of 0.2 is imposed.

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**Fig. 5** Example of $V_S$ independent of confining stress, $n = 0.07$ based on the data for medium sand of the Boxtel Formation. For this data set, a slope of $n = 0$ is chosen. The solid line describes the regression while dotted lines indicate 95% confidence intervals.
VS estimates were available from the SCPT data set. Therefore, we used judgement to infer average VS for these units. Next, the intercept ln $V_{S1}$ was determined such that the estimate of VS occurs at the average depth of occurrence in the region and consistent with the slope $n$. In the Groningen region, the intercept at \( \ln \left( \frac{\rho_c}{\rho_w} \right) = 0 \) corresponds to a depth of approximately 13 to 14 m. The parameters describing the confining stress dependence in this fashion are given in Table 3. A standard deviation of 0.27 for ln $V_S$ is imposed for all peats, and a value of 0.20 for all other lithologies, consistent with the values from Table 1 and Table 2.
Not all units present in the Groningen region are represented in the SCPT data set. In those cases, either representative relations from similar units or expert estimates were used (e.g. Table 3). For example, all members from the Naaldwijk Formation were represented by the Naaldwijk Vₜ distributions in Table 1 and 3. Additionally, the Vₜ distributions of Nieuwkoop Holland Peat are assumed to be representative for all Holocene peats.

### 3.2 MEI Vₜ model

Three-dimensional seismic reflection data was acquired in the 1980s by NAM for the purpose of imaging and characterisation of the Groningen gas reservoir. This legacy data set was used to constrain the Vₜ model in the intermediate depth range. Surface waves are generally regarded as noise in the process of seismically imaging deep reflectors. Therefore, they are attenuated during acquisition of seismic data and suppressed during processing. The surface (and guided) waves, however, propagate along the surface and therefore contain useful information of the elastic properties of the near-surface. Survey techniques have been designed that use these types of waves (e.g. Park et al. 1999). Hence, inversion of these surface waves was used to derive a Vₜ model.

| Formation                              | Lithological class                       | Slope \( n \) | Intercept \( V_{S1} \) (m/s) | Standard deviation \( \ln(V_S) \) |
|----------------------------------------|------------------------------------------|---------------|-------------------------------|---------------------------------|
| Appelscha, Boxtel, Drachten, Eem, Peelo, Urk-Tynje member | Peat                                     | 0.25          | 122.7                         | 0.27                            |
| Appelscha                              | Clay                                     | 0.25          | 267.7                         | 0.20                            |
| Boxtel                                 | Clay                                     | 0.25          | 177.7                         | 0.20                            |
| Drachten                               | Clay                                     | 0.25          | 146.9                         | 0.20                            |
| Drachten                               | Sandy clay and clayey sand               | 0.25          | 221.4                         | 0.20                            |
| Drente                                 | Fine sand                                | 0.25          | 225.9                         | 0.20                            |
| Drente                                 | Medium sand                              | 0.25          | 239.8                         | 0.20                            |
| Drente                                 | Coarse sand, gravel and shells           | 0.26          | 237.5                         | 0.20                            |
| Drente, Gieten member                  | Fine sand                                | 0.25          | 278.7                         | 0.20                            |
| Drente, Gieten member                  | Medium sand                              | 0.29          | 295.9                         | 0.20                            |
| Drente, Gieten member                  | Coarse sand, gravel and shells           | 0.26          | 295.9                         | 0.20                            |
| Eem                                    | Clay                                     | 0.25          | 194.4                         | 0.20                            |
| Naaldwijk                              | Medium and coarse sand, gravel and shells| 0.25          | 308.0                         | 0.20                            |
| Nieuwkoop Basal Peat                   | Clay                                     | 0.25          | 141.2                         | 0.27                            |
| Nieuwkoop Basal Peat                   | Sandy clay and clayey sand, fine, medium and coarse sand, gravel and shells | 0.25          | 170.7                         | 0.27                            |
| Urk, Tynje member                      | Clay                                     | 0.25          | 167.3                         | 0.20                            |
| Urk, Tynje member                      | Sandy clay and clayey sand               | 0.25          | 194.4                         | 0.20                            |
| Urk, Tynje member                      | Fine sand and medium sand                | 0.26          | 219.2                         | 0.20                            |
| Urk, Tynje member                      | Coarse sand, gravel and shells           | 0.26          | 265.1                         | 0.20                            |
The Modal Elastic Inversion (MEI) method was used for the elastic near-surface model building. In essence, the MEI method is an approximate elastic Full Waveform Inversion method, in which the elastic wavefield is approximated by focusing on waves that propagate laterally through the shallow subsurface. These waves include the fundamental mode of the Rayleigh wave, its higher modes and guided waves. A limited number of horizontally propagating modes, characterized by lateral propagation properties and depth-dependent amplitude properties, are taken into account to represent the near-surface elastic wavefield (Ernst, 2013). The objective in the MEI approach is to find a model that minimizes the difference between the observed data (Rayleigh waves) and the forward modelled data.

The pre-processing applied to the data prior to Modal Elastic Inversion was restricted to applying a high-cut filter and data selection of those data traces that contain the Rayleigh waves. For efficiency, the data set was split in large overlapping rectangular areas, which were inverted independently. Within one area, all data are inverted simultaneously. The resulting $V_S$ models are merged afterwards. The starting model was a laterally invariant vertical gradient, which was subsequently updated during the inversion. All lateral variations in the resulting $V_S$ model were introduced by the inversion. Generally, the uncertainty in $V_S$ values in the resulting $V_S$ model is estimated to be 5–10%.

The vertical resolution of the resulting $V_S$ model is limited and the maximum depth range to which $V_S$ in this case can be reliably estimated is approximately 120 m below the surface. This is due to the seismic data acquisition design and consequently the narrow frequency band in which the surface waves are unaffected and still present in the data. The surface seismic data was acquired in 1988 with mostly (buried) dynamite sources and to a lesser extent with vibroseis sources (in cities) or airgun sources (in lakes and offshore), and recorded with 10 Hz vertical geophones. The seismic data acquisition was designed for deep imaging of the Groningen reservoir with a typical orthogonal geometry with line spacing of 250–500 m and group spacing of 50 m. The receiver group arrays were designed to suppress and distort Rayleigh waves with wavelengths less than approximately 80 m. The effect of the geophone arrays on the surface waves is illustrated in Fig. 6, in which a typical seismic record is displayed with full frequency band and with a frequency high-cut filter applied to 3 Hz. Above 3 Hz the receiver arrays have distorted and aliased the Rayleigh waves, and below 1 Hz the Rayleigh waves have become too weak to be observed on the seismic records. The application of the MEI method to the data was

![Typical input seismic record from the Groningen 3D seismic reflection data with full frequency band (left) and high-cut filter at 3 Hz (amplified ×10, right). The numbers on the x-axis indicate trace numbers.](image)
therefore restricted to the bandwidth from 1 to 3 Hz, and to those recorded traces on which the Rayleigh waves were recorded.

The narrow temporal bandwidth results in a narrow range of wavelengths (roughly between \( \sim 70 \) m and \( \sim 500 \) m). The penetration depth of the Rayleigh wave depends on the wavelength: the short wavelengths are sensitive to the shallow subsurface velocities, whereas the long wavelengths are more sensitive to the deeper velocities. The narrow range of wavelengths and especially the lack of short wavelengths therefore results in limited resolving power for the very shallow subsurface velocities. A typical depth resolution kernel of the fundamental mode of the Rayleigh wave is shown in Fig. 7. The high frequencies (right side of the plot) are more sensitive to the shallow layers, while the low frequencies (left side of the plot) are more sensitive to the deeper layers. The maximum penetration depth is \( \sim 120 \) m and there is a limited resolving power of velocities in the shallow layers of the model (0–20 m).

The convergence of the inversion is verified using the normalized root-mean-square (RMS) misfit between the model and the data for each seismic shot (Fig. 8). The normalized RMS misfit ranges from 0 (excellent convergence) to 100 (very poor convergence). Generally, Fig. 8 shows that the normalized RMS misfit is good, but there are several areas with large misfits (denoted by red outlines). These larger misfits are linked to the source types used during the seismic acquisition. In and around cities or highways vibroseis sources were used and airguns were used in lakes. Both source types do not contain the low frequencies required for the inversion. The estimation of the shear-wave velocity model in these areas might be hampered by these conditions. In other areas, the seismic records were acquired using buried dynamite sources and show a much better

**Fig. 7** Representative depth sensitivity kernel for the fundamental mode of the Rayleigh wave as a function of depth and frequency. Red indicates low sensitivity, blue indicates high sensitivity.
RMS. The area in the north is characterized by a high ambient noise level that cannot be modelled and therefore shows up as a relatively large RMS misfit.

The inversion resulted in a $V_S$ model over the area where 3D seismic data was available, and therefore does not cover the full extent of the area of interest. Horizontally, the $V_S$ model is gridded on the same 100 m × 100 m grid as the GeoTOP model. Vertically, the $V_S$ model is defined at 10 m depth intervals. An example of a depth slice at 65 m depth is shown in Fig. 9. The MEI $V_S$ model shows distinct zones of relatively high and relatively low $V_S$ values in patterns that resemble geological features, such as buried valleys. These structures can also be recognized in a cross-section from West to East in the centre of the field (Fig. 10). The cross-section also shows the vertical smoothness of the model.

### 3.3 Sonic $V_S$ model

For larger depths, $V_S$ is derived from the seismic data that was collected to image the reservoir. One component of the processing of seismic data for imaging is the application of pre-stack depth migration (Yilmaz 2001), which among others moves dipping reflectors to their true positions. This procedure requires a velocity model, the so-called Pre-Stack Depth Migration Velocity model (PSDM velocity model). There are more than 500 wells in the Groningen field. Data from these wells were available for this project. Sonic logs, providing $V_P$, were measured in 70 of them. In several wells, $V_S$ was measured as well over a limited depth range. In two wells both $V_P$ and $V_S$ were measured over the entire North Sea Supergroup. Sonic logs and well markers for key horizons are used to construct a depth-calibrated, high-resolution P-wave ($V_P$) model over the entire field. There is sufficient coverage of sonic logs for depths larger than 200 m, but for shallower depths, the accuracy of the $V_P$ model is reduced.

The PSDM velocity model is used as input $V_P$ model for the North Sea Supergroup. Site response calculations, however, require information in terms of $V_S$ instead of $V_P$. Hence,
the PSDM $V_P$ values are converted to $V_S$ using $V_P/V_S$ relations from the two well logs where both $V_P$ and $V_S$ were measured over the entire North Sea Supergroup (Fig. 11). The measurements of $V_P$ and $V_S$ start below the depth of the conductor in the well, at $\sim 60–75$ m. The ratio between $V_P$ and $V_S$ shows a linear decrease with depth in the Upper North Sea Group, while it is more or less constant in the Lower North Sea Group (Fig. 11).

The linear relationship to convert $V_P$ into $V_S$ for the Upper North Sea Group is given by:

$$V_S = \frac{V_P}{4.7819 - 0.0047Z}$$

where $Z$ is the depth in metres. The corresponding Poisson’s ratio in the Upper North Sea Group generally varies between 0.45 and 0.47. The constant relation to convert $V_P$ into $V_S$ for the Lower North Sea Group is given by:
This corresponds to a Poisson’s ratio of 0.446.

The Sonic $V_S$ model was discretised in layers of 25 m thickness and on a grid identical to the 100 m $\times$ 100 m cells of the GeoTOP model. A cross section of the sonic $V_S$ model through the centre of the field is shown in Fig. 12. The $V_S$ inversion which is present in the Lower North Sea Group at depths of $\sim$500 m is caused by the Brussels sand. Locally, this sand is cemented, leading to high $V_S$.

![Fig. 12 Cross section through the Sonic $V_S$ model, from west to east at the centre of the field. The vertical scale is exaggerated. The base of the Upper North Sea Group is indicated by the black line; the base of the Lower North Sea Group by the thin yellow line](image)
4 Generation of V\textsubscript{S} profiles

4.1 Splicing of the final V\textsubscript{S} models

The three V\textsubscript{S} models are spliced in order to obtain V\textsubscript{S} profiles from the surface to the reference baserock horizon at the base of the North Sea Supergroup for each location in the field. The top part, from the surface to NAP-50 m, consists of the GeoTOP V\textsubscript{S} model. The MEI V\textsubscript{S} model is appended between NAP-50 m and a maximum of NAP-120 m. The layer thicknesses in the MEI V\textsubscript{S} depth range are taken from the geological scenarios of the geological model below 50 m (Sect. 5). The maximum thickness of layers in this depth range is 10 m.

The extent of the MEI model is smaller than the extent of the area of interest, comprising of the Groningen field with a 5 km buffer. For regions outside the MEI range, the average MEI V\textsubscript{S} for a depth slice was selected as an estimate of V\textsubscript{S}. In effect, this yielded an increasing V\textsubscript{S} model with depth, but without detailed channel-like structures.

The transition between the MEI and the sonic V\textsubscript{S} model is chosen at the depth where the two V\textsubscript{S} profiles intersect. This choice avoids velocity inversions at the transition. The layer thicknesses in the sonic V\textsubscript{S} range are taken from the geological scenarios of the geological model below 50 m (Sect. 5) with a maximum thickness of 25 m. The reference rock horizon is represented by the base of the North Sea Supergroup. At this level, there is an impedance contrast as V\textsubscript{S} jumps from ~600 m/s (on average) just above this level to 1400 m/s (on average) just below this level. Examples of typical V\textsubscript{S} profiles constructed from the three V\textsubscript{S} models are shown in Fig. 13. In total, approximately 140,000 V\textsubscript{S} profiles were generated.

4.2 Randomisation of V\textsubscript{S} profiles

The spatial variability within each geological zone (Sect. 5) needs to be captured in the site response analyses. To this end, randomisation was applied in the site response calculations to the soil composition, input motions (Rodriguez-Marek et al. 2017) and to the V\textsubscript{S} profiles in the GeoTOP depth range. Randomisation of soil composition is achieved by assuming that the collection of GeoTOP voxel-stacks within one geological zone represent the likely

![Fig. 13 Examples of V\textsubscript{S} profiles over the full depth range, with sampled and mean V\textsubscript{S} in the top 50 m and MEI and Sonic V\textsubscript{S} below 50 m. The sampling is described in Sect. 4.2. The V\textsubscript{S} at reference rock horizon (1400 m/s) is out of the horizontal scale](image)
successions of units within that zone. The procedure to develop randomised GeoTOP $V_S$ profiles is shown in Figs. 14 and 15. Shear-wave velocity profiles were calculated for each voxel-stack in the GeoTOP model. For each unit in the voxel-stack (Fig. 14a,b), the corresponding $V_S$ relation is selected from Tables 1, 2, or 3. A sensitivity study indicated that the 0.5 m layering of GeoTOP created unrealistic site response results. Therefore, the GeoTOP layers were resampled into layers of 1.0 m, by examining the two voxels within every metre and selecting one of the voxels at random (Fig. 14c–e). Next, consecutive voxels of the same unit were merged into one layer up to a maximum thickness of 3.0 m (Fig. 14f,g). The maximum thickness of 3.0 m was imposed to preserve the confining stress-dependence of $V_S$.

Within one voxel-stack and one unit of stratigraphy and lithological class we assume full correlation of $V_S$. This means that all layers of a given unit within one voxel-stack are based on one sample of $V_S$ from the $V_S$ distribution of this unit (i.e. from Tables 1, 2, or 3). Within one voxel-stack and between different units we assume a correlation coefficient $\rho$ of 0.5. In order to avoid $V_S$ profiles that have extremely low or extremely high (and therefore unrealistic) $V_S$ values, the distributions were truncated at two standard deviations. This truncation follows common practice in site response analyses of nuclear facilities (EPRI 2013). To compensate for the truncation, the $V_S$ values are sampled from a distribution with a standard deviation that is increased by 16%. This value corresponds to the value that would render a truncated distribution with the desired (target) standard deviation. A correlated sampling approach was implemented largely following Toro (1995). The $V_S$ distributions were standardized in order to be able to sample in a correlated way between units having different $V_S$ distributions (different average and standard deviation of $\ln V_S$). Truncation was implemented as follows:

Fig. 14 Example of GeoTOP voxel-stack processing. From left to right: (a) original GeoTOP stratigraphic units; (b) original GeoTOP lithological classes; (c) resampled lithological classes into layers of 1.0 m thickness; (d) resampled stratigraphic units into layers of 1.0 m thickness; (e) random selector used to select the upper (grey) or lower (black) voxel; (f) merged lithological classes with a maximum thickness of 3.0 m; (g) merged stratigraphic units with a maximum thickness of 3.0 m. For the clays (green) of the Peelo Formation (purple), the mean depth of the unit is indicated. Far right: bar graph of the sampled shear-wave velocity profile assigned to the voxels by applying the routine from Fig. 15 to voxel-stacks (f) and (g).
1. Draw a random sample $\ln(V_{\text{sample}})$ from a normal distribution with

$$\mu = \ln(V_{\text{mean}}) \quad \text{and} \quad \sigma^* = 1.16\sigma_{\ln V_S} \quad (4)$$

2. Standardise to a distribution with $\mu = 0$ using

$$\ln\left( V_{\text{sample,standardized}} \right) = \frac{\left( \ln(V_{\text{sample}}) - \mu \right)}{\sigma^*} \quad (5)$$

3. Repeat steps 1 and 2 until

$$\left| \ln\left( V_{\text{sample,standardized}} \right) \right| < 2.0 \quad (6)$$

The random sample for each unit is taken at the average depth of occurrence of this unit in the voxel-stack. For the confining stress-dependent $V_S$ relations in Table 1 the standard deviation is related to the distance to average $\ln\left( \frac{\Delta S}{\Delta V} \right)$. In order to avoid sampling in the confining stress range either outside the range defined by the data, or always at the tails of the distribution which might result in relatively large standard deviation, the random sample $\ln(V_{\text{sample}})$ is taken at the average depth of occurrence of the particular unit, assuming that this is comparable to the average confining stress.

When moving to the next unit in the voxel-stack, correlated sampling is applied, again at the average depth of occurrence of the next unit. The correlated sampling is implemented as follows:

1. Draw an auxiliary variable $b$ (needed for standardized and truncated distribution) from a normal distribution with $\mu = 0$ and $\sigma = 1.16$.
2. Repeat step 1 until $|b| < 2.0$.
3. Calculate $\ln\left( V_{\text{sample,standardized}} \right)$ correlated to the previous layer using the correlation coefficient $\rho$ and auxiliary variable $b$ using:

$$\ln\left( V_{\text{sample,standardized}} \right) = \rho \ln\left( V_{\text{previous,layer,standardized}} \right) + b \sqrt{1 - \rho^2} \quad (7)$$

4. Transform $\ln\left( V_{\text{sample,standardized}} \right)$ to $\ln(V_{\text{sample}})$ using:

$$\ln(V_{\text{sample}}) = \mu + \sigma^* \ln\left( V_{\text{sample,standardized}} \right) \quad (8)$$

where $\mu$ is the mean $V_S$ value at that depth.

5. Use $\ln\left( V_{\text{sample,standardized}} \right)$ as $\ln\left( V_{\text{previous,layer,standardized}} \right)$ in Eq. (7) in the calculation of the next unit.

Using the above described procedure, the truncated and correlated $\ln V_S$ is sampled for each unit at one depth per unit. In order to determine the shear-wave velocities at other depths of this unit in the voxel stack, the updated intercept $\ln V_{S2}$ is determined using the slope $n$ of the corresponding distribution and $\ln(V_{\text{sample}})$ from Eq. 8 for this unit using:
\[
\ln V_{S2} = \ln V_{S_{\text{sample}}} - \left( n \ln \left( \frac{\sigma_o}{P_o} \right)_{\text{averaged depth}} \right)
\]

(9)

Finally, the \(\ln V_S\) values at all other depths (and thus confining stresses) within this voxel-stack of this unit are calculated using:

\[
\ln V_S = \ln V_{S2} + n \ln \left( \frac{\sigma_o}{P_o} \right)
\]

(10)

In effect this means that only \(\ln V_{S1}\) and not the slope \(n\) is randomized in Eq. 1.

Examples of randomized \(V_S\) profiles are shown in Fig. 16. For reference, the units and the profiles based on the mean \(V_S\) relation are included as well. For uniform units, the confining stress dependent increase in \(V_S\) is apparent (e.g. in the left panel of Fig. 16). The correlated sampling ensures that the jumps in \(V_S\) between units are not unrealistically large. In some cases, almost the entire profile is sampled in the low side of the mean (e.g. middle panel of Fig. 16). Because of a correlation coefficient of 0.5, jumps from relatively high sampled \(V_S\) to relatively low sampled \(V_S\) between units is still possible (e.g. right panel of Fig. 16).

5 Zonation and layering

The geological cross section of Fig. 2 shows that the subsurface in Groningen is heterogeneous. As a consequence, the properties that dictate the response to earthquake shaking will be spatially variable. However, for practical considerations the site response analyses were conducted for a finite number of zones within the Groningen field (Rodriguez-Marek et al. 2017). Hence, we defined zones of geologically similar build-up to accommodate for the heterogeneity of the subsurface. These zones were used as a starting point for the site response zonation (Rodriguez-Marek et al. 2017).

Several sources of information were available for the definition of the geological zones: the high resolution 3D geological model GeoTOP (Stafleu et al. 2011; Stafleu and Dubelaar 2016), 19,082 borehole descriptions from the DINO database (www.dinoloket.nl), 5674 cone penetration tests, the Digital Geological Model of the Netherlands (DGM, Gunnink et al. 2013), the REgional Geohydrological Information System II (REGIS II, Vernes and van Doorn 2005), the digital terrain model AHN (open data, www.ahn.nl) and paleogeographic maps (Vos and Knol 2015; Vos et al. 2011).

Because of the difference in resolution and depth range of the various geological sources, the region of Groningen has been divided into zones of similar geology for two depth ranges (Kruiver et al. 2015). Between the surface and the maximum depth of GeoTOP (NAP-50 m), the geological zones are defined based on characteristic successions. The model consists of a geological zonation map (Fig. 17) and the GeoTOP voxels with stratigraphic and lithoclass information. This zonation was applied to the site response results (Rodriguez-Marek et al. 2017), because of the large contribution of the shallowest deposits to the site response.
For each 1 - 3 m thick layer in the voxel-stack

First time strat-lith combination is encountered?

Yes

Determine mean depth of strat-lith combination

Convert mean depth to \( \ln(\sigma'_w/p_r) \)

Uppermost layer?

No

Eq. (4)

Draw random sample \( \ln(V_s) \) based on Table 1, 2, 3

Eq. (5)

Standardize to a distribution with \( \mu = 0 \)

Eq. (6)

|Standardized value| < 2.0 ?

|b| < 2.0 ?

No

Eq. (7)

Calculate correlated, standardized \( \ln(V_s) \)

Yes

Eq. (8)

Transform to \( \ln(V_s) \)

Eq. (9)

Calculate intercept \( \ln(V_{s_2}) \)

Convert layer depth to \( \ln(\sigma'_w/p_r) \)

Calculate \( \ln(V_s) \) of the layer using slope and intercept

Vs profile
In addition to the $V_S$ profiles, site response analyses require the assignment of nonlinear properties to each layer. This assignment is based on the soil type for each layer (e.g. using Darendeli 2001 or Menq 2003). For the near surface layers, the stratigraphic and lithoclass information from the GeoTOP model is used to assign soil types to each layer. The deeper geological structures are different from the shallow structures, hence a different geological zonation map applies for the depth range between NAP-50 m and the base of the Upper North Sea Group. The layer and composition information for this depth range which are needed for assigning appropriate nonlinear properties to each layer is represented by characteristic scenarios for subsurface successions for each zone, including a probability of encountering each scenario in that zone. In the site response calculations, the Lower North Sea Group was considered to behave linearly (Rodriguez-Marek et al. 2017). Therefore, details on the composition of these layers, beyond its $V_S$ value, are not needed. The full layer profile for each location (X, Y coordinate on the GeoTOP grid) is a combination of the GeoTOP layers of stratigraphy and lithoclass, appended with one of the deeper scenarios while taking into account the probabilities of each scenario on the deeper geological zones.

6 $V_{S30}$ map

The time-averaged shear-wave velocity over the upper 30 m of a profile ($V_{S30}$) is used as an input to some of the components of the ground motion model for the Groningen region (Bommer et al. 2017). The GeoTOP $V_S$ model enables the determination of a $V_{S30}$ map. This map is also relevant for the new building codes in the Netherlands. A $V_S$ profile is calculated between the surface and 30 m depth following the scheme described in Sect. 4, except that the layers of 0.5 m thickness of GeoTOP have been preserved. The $V_{S30}$ is then calculated from the $V_S$ profile as the harmonic mean. This procedure was repeated 100
times with a different initial random sample for each voxel-stack, which provided 100 estimates of $V_{S30}$ for each GeoTOP grid cell.

The average and standard deviation of $V_{S30}$ values for each geological zone were calculated from all $V_{S30}$ values within that zone. For the entire field, the average $V_{S30}$ based on all $V_{S30}$ realisations is 207 m/s with a standard deviation of 48 m/s. The $V_{S30}$ maps with zonation are shown in Fig. 18 for mean and standard deviation of $V_{S30}$. The data are plotted in coloured bins in these figures; the $V_{S30}$ data per zone is available in Online Resource 1. The average and median $V_{S30}$ are very similar, with a maximum difference of 5 m/s. The maps show a distinct pattern in $V_{S30}$ that is related to geology. The northern part of the region contains the thickest layers of soft Holocene deposits, with overall low $V_{S30}$ values. Channel structures can be recognised, e.g. in the eastern part. The southern part consists of Pleistocene deposits which are generally stiffer (but still relatively soft soil), which is reflected in the higher $V_{S30}$ values. The Hondsrug (the sand ridge on which the city of Groningen is situated) stands out as a relatively high $V_{S30}$ zone in the south west. Immediately east of the ridge is a channel that is filled with soft Holocene deposits. This is reflected by the low $V_{S30}$ zone adjacent to the Hondsrug. The right panel of Fig. 18 shows that the standard deviation of $V_{S30}$ ranges from 25 to 54 m/s between zones. Although the $V_{S30}$ values are relatively low (soft soil), variation of $V_{S30}$ within zones, expressed by the standard deviation, can be significant. The degree of variation in $V_{S30}$ within the zones is not uniform across the entire field. In particular, within-zone variation is greater in the south where the average $V_{S30}$ values are higher.
In this paper a shear-wave velocity model is presented spanning the depth range from the surface to about 800 m depth as a starting point for site response analyses in the Groningen field. The combination of three different VS models (in terms of depth ranges) resulted in a model that is unique on this scale. Furthermore, a randomisation scheme for the shallow VS profiles (to 50 m depth) was developed, taking into account the geological characteristics of the subsurface. This randomisation is required to determine the probabilistic amplification factors that feed into the seismic hazard and risk analysis. To facilitate the randomisation of VS, we derived VS relations for each unit of stratigraphy and lithology that is present in the region. Full correlation of VS is assumed within one unit in a vertical profile. Between units, however, partial correlation is assumed using a correlation coefficient of 0.5. Additionally, a microzonation model was constructed to account for geological heterogeneity.

The VS relations for the shallow depth range were based on SCPT data measured in the region. For some units there was insufficient data to determine a confining stress-dependent relation. Future fieldwork campaigns will increase the number of SCPTs available for this analysis. Additionally, there is a large number of CPTs (~5700) available in the region. This provides the opportunity to design a region specific VS relation based on both SCPTs and CPTs. When the CPTs are converted to synthetic VS profiles using relations that are calibrated on the SCPT-CPT data set, more units can be quantified. Rather than using standard classification methods (e.g. Robertson 1990), the adapted scheme is needed to accommodate soils such as peats and glacial clays.
Additionally, the transition between the GeoTOP and MEI V_S models is currently defined at one fixed depth, i.e. the maximum depth of the GeoTOP model. An alternative choice might be a transition level that varies in depth and is based on the location of channel structures or other geological features. However, to implement this approach the exact locations of these features needs to be known and these are currently not known in sufficient detail. Future subsurface models by TNO—Geological Survey of the Netherlands may include the required level of detail. In the current model, scenarios of stratigraphy were used below NAP-50 m to accommodate the uncertainty in locations of key geological features.

The model building presented in this paper represents a unique exercise in which a comprehensive set of geological, geotechnical, and geophysical data was used to build an extensive 3D V_S model for site response analyses. The construction of this deep and detailed V_S model has enabled the incorporation of fully probabilistic, nonlinear site amplification functions into the estimation of surface ground motions (Rodriguez-Marek et al. 2017). In essence, an approach to including local site effects in seismic hazard analyses usually applied in site-specific studies for critical facilities can thus be applied to hazard and risk assessments for an entire region (Bommer et al. 2017).

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