

1 **Neutrons on Rails – trans-regional monitoring of soil**
2 **moisture and snow water equivalent**

3 **M. Schrön¹, S. E. Oswald², S. Zacharias¹, M. Kasner¹, P. Dietrich^{1,3},**
4 **S. Attinger^{1,2}**

5 ¹UFZ – Helmholtz Centre for Environmental Research GmbH, Leipzig, Germany

6 ²Institute of Environmental Science and Geography, University of Potsdam, Potsdam, Germany

7 ³Center of Applied Geoscience, Eberhard Karls University of Tübingen, Tübingen, Germany

8 **Key Points:**

- 9 • Cosmic-ray neutron detectors in trains respond to spatial patterns of water con-
10 tent.
- 11 • First experiments and analysis provide a proof of concept.
- 12 • Using the railway system for regular environmental monitoring could extend the
13 measurement capability to trans-regional and nationwide scales.

Corresponding author: Martin Schrön, martin.schroen@ufz.de

Abstract

14 Large-scale measurements of the spatial distribution of water content in soils and snow
15 are challenging for state-of-the-art hydrogeophysical methods. Cosmic-ray neutron sens-
16 ing (CRNS) is a non-invasive technology that has the potential to bridge the scale gap
17 between conventional in-situ sensors and remote-sensing products in both, horizontal and
18 vertical domains. In this study we explore the feasibility and potential of estimating wa-
19 ter content in soils and snow with neutron detectors in moving trains. Theoretical con-
20 siderations quantify the stochastic measurement uncertainty as a function of water con-
21 tent, altitude, resolution, and detector efficiency. Numerical experiments demonstrate
22 that the sensitivity of measured water content is almost unperturbed by train materi-
23 als. And finally three distinct real world experiments provide a proof of concept on short
24 and long-range tracks. With our results a trans-regional observational soil moisture prod-
25 uct becomes a realistic vision within the next years.
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Plain Language Summary

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28 Large-scale measurements of the spatial distribution of water content in soils and
29 snow are challenging for state-of-the-art hydrogeophysical methods. Cosmic-ray neutron
30 sensing (CRNS) is a mobile and non-invasive technology that has the potential to bridge
31 the scale gap between conventional in-situ sensors and satellite measurements in both,
32 the horizontal and the vertical domain. In this study we explore the feasibility and po-
33 tential of estimating water content in soils or snow with neutron detectors in trains. The-
34 oretical considerations estimate how train velocity and recording period influence the un-
35 certainty of such measurements and demonstrate that train materials hardly affect the
36 neutron response to water content. Three distinct experiments provide a proof of con-
37 cept on short and long-range tracks. With our results a trans-regional observational soil
38 moisture product becomes a realistic vision within the next years.

1 Introduction

Water stored in soils and snow controls the energy and water exchange between the terrestrial surface and the atmosphere (Vogel et al., 2018), impacts regional weather, and shapes the development of hydrometeorological extremes like heat waves, droughts, floods, or avalanches (e.g., Lehning et al., 1999; Douville & Chauvin, 2000; Liang & Yuan, 2021). Therefore, a solid estimation of land surface water at relevant spatiotemporal scales is of utmost importance.

Satellite-based remote sensing platforms aim at global estimations of water in soils and snow at resolutions of several kilometers with the prospect of finer resolutions using new instrumentation and algorithms (Chan et al., 2016; Mattia et al., 2018; Foucras et al., 2020). However, major limitations are the shallow measurement depth (\sim cm), long return frequencies (\sim days), and low performance during complex weather conditions, under vegetation cover, and in complex terrain (Fang & Lakshmi, 2014; Lawford, 2014). Ground-based in-situ sensors have been developed to measure water content in different soil depths with high spatiotemporal precision and extent from the point to smaller field scales (Bogena et al., 2010; Romano, 2014; Ochsner et al., 2013; Dorigo et al., 2021). However, the sensors require regular maintenance, the spatial representativity is low (\sim m), and the concept is not scalable to regional scales (Pan & Peters-Lidard, 2008; Schelle et al., 2013).

The critical scale gap between in-situ and remote sensing techniques could be closed by Cosmic-Ray Neutron Sensing (CRNS) (Zreda et al., 2012; Andreasen et al., 2017). The proximal sensing technique is based on the passive and non-invasive detection of cosmic-ray neutrons. Their intensity strongly depends on the hydrogen abundance in the surrounding environment and thereby responds to water. According to Köhli et al. (2015), the signal intrinsically covers large areas (\sim 15 ha) and soil depths (\sim 10–70 cm), making it a "fundamental breakthrough ... in the monitoring ... of soil moisture status" (Romano, 2014). Stationary CRNS has been successfully applied in research, agriculture, and water resource management, usually at single locations, sometimes integrated in small or national networks (Evans et al., 2016; Andreasen et al., 2017; Fersch et al., 2020). To overcome this spatial limitation, the detector can be used in a mobile mode similar to the so-called car-borne CRNS rover. This mobile variant can capture spatially resolved information of root-zone soil moisture (or snow) from the field to catchment scale *en pas-*

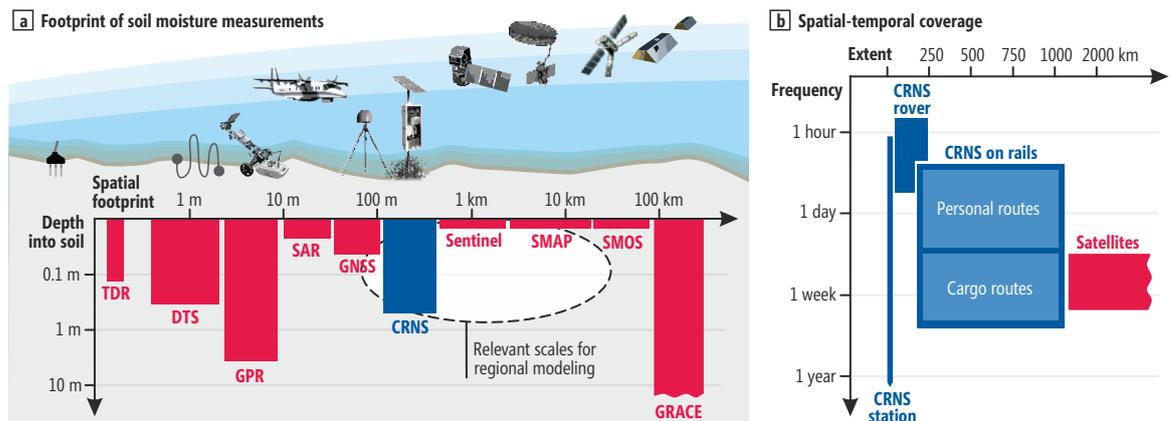


Figure 1. CRNS on rails may close the gap for soil moisture products at relevant scales. (a) Horizontal and vertical footprint of soil moisture measurement techniques (log scale). (b) Spatial and temporal extent of stationary CRNS, mobile CRNS, and satellite systems.

71 *sant* (Dong et al., 2014; McJannet et al., 2014; Schrön, Rosolem, et al., 2018; Fersch et
72 al., 2018; Vather et al., 2019). However, this car-borne CRNS roving is a demanding method
73 in terms of personnel and time constraints. A monitoring concept to support a trans-
74 regional observational product would require significantly larger scales and more frequent
75 measurements (see Fig. 1). One option for a new mobile platform is the use of existing
76 transport infrastructure, such as public or cargo transportation (buses, trucks, or trains).

77 While railway infrastructure is already attracting some interest in the field of ap-
78 plied geophysics (Lavoué et al., 2021; Izvolt et al., 2016), the present feasibility study
79 specifically pushes the current limitation of mobile CRNS towards trans-regional mon-
80 itoring. We will begin with theoretical considerations about the expected measurement
81 uncertainty and spatial resolution, followed by a sensitivity study on the influence of ve-
82 hicles and track beds using numerical simulations. We will then showcase three exper-
83 iments in real trains to demonstrate the feasibility of train-based CRNS roving to trans-
84 regionally monitor water content of soils and snow.

2 Basic considerations on the general feasibility

The CRNS method relies on the detection of the natural neutron radiation in the energy range of 1–1000 eV. The key features of this method are:

- neutrons are mostly insensitive to most materials other than hydrogen (Zreda et al., 2012),
- the detector receives signals from all directions, integrating over a footprint area of ~ 100 – 300 m radius (Köhli et al., 2015),
- neutron intensity provides information regarding soil moisture from depths of up to 70 cm (Franz et al., 2013; Köhli et al., 2015).

CRNS detectors usually consist of shielded proportional gas counters to detect passing epithermal neutrons (Zreda et al., 2012; Schrön, Zacharias, et al., 2018; Köhli et al., 2018). Their count rates are typically measured in counts per hour (cph) or counts per minute (cpm).

2.1 From neutrons to water

The measured neutrons respond to all static and dynamic hydrogen pools in the footprint, such as water in soils (Scheffele et al., 2020), vegetation (Jakobi et al., 2018), and snow (Schattan et al., 2019). They are corrected for the effect of variable air pressure, air humidity, and incoming cosmic rays (Zreda et al., 2012; Hawdon et al., 2014; Schrön et al., 2016) using direct measurements or data from weather services. The relationship between corrected neutrons, N , and the total water θ contained in the various pools has been described by Desilets, Zreda, and Ferré (2010) and adopted by most authors in the last decade. According to analysis from Köhli, Weimar, Schrön, and Schmidt (2021) it can be reformulated as:

$$\theta(N) \approx p_0 \frac{1 - N/N_{\max}}{p_1 - N/N_{\max}}, \quad (1)$$

where N_{\max} is the maximum neutron flux under dry conditions which mainly depends on the individual detector sensitivity, but has also been suspected to reflect site-specific conditions (Zreda et al., 2012). Parameters $p_0 = -0.115$, $p_1 = 0.346$, and $N_{\max} = 1.075 N_0$ can be derived from the parameters used so far in the Desilets equation (Desilets et al., 2010) which most of the previous research studies refer to. Here, N_{\max} (or N_0) represents site- and detector-specific efficiency and is often used as a calibration parameter. In order to generate gravimetric and volumetric soil moisture products, we account for soil bulk density, soil lattice water, and organic material as described in Dong et al. (2014); McJannet et al. (2017); Schrön, Rosolem, et al. (2018) using the soil grids database (Hengl et al., 2017). Road-effect corrections are only applied for car-borne roving following Schrön, Rosolem, et al. (2018) and should be adapted to railways in future research. Those and other factors could introduce systematic uncertainty to a specific derived product as has been quantified by Baroni et al. (2018) and Jakobi et al. (2020). In addition, Fersch et al. (2018) suggested a method to reduce uncertainties from landscape heterogeneity passed by the moving detector. Nevertheless, all measurements have a lower limit of stochastic uncertainty in common due to the random nature of neutron detection.

2.2 Stochastic measurement uncertainty and spatial resolution

High count rates are essential to ensure a sufficiently high signal-to-noise ratio and reliable data products. Mobile measurements require a particularly high sensor efficiency to reduce the data accumulation period and the corresponding areal coverage (i.e., foot-

129 print length). This balance is a major challenge for large-scale CRNS applications that
 130 inextricably links measurement quality with spatial resolution.

131 Due to the non-linearity of Eq. 1, the propagated water content uncertainty, $\pm\sigma_\theta =$
 132 $\theta(N) - \theta(N \mp \sigma_N)$, is highly asymmetric. For simplicity, it can be estimated by a sym-
 133 metrical approximation approach suggested by Jakobi et al. (2020):

$$\sigma_\theta \approx \sigma_N \frac{p_2 N_0}{(N - p_3 N_0)^4} \sqrt{(N - p_3 N_0)^4 + 8 \sigma_N^2 (N - p_3 N_0)^2 + 15 \sigma_N^4}, \quad (2)$$

134 where the count rate $N(\theta)$ follows from Eq. 1, $\sigma_N \sim \sqrt{N}$ is its Gaussian uncertainty,
 135 N_0 represents the detector-specific efficiency, and $p_2 = 0.0808$, $p_3 = 0.372$. The more
 136 neutrons N are collected per measurement, the lower are the signal-to-noise ratio σ_N/N
 137 and consequently the lower the uncertainty of the soil moisture product, σ_θ . N depends
 138 primarily on the water content, but it is ultimately constrained by detector type, aggre-
 139 gation window, and background radiation intensity. To express these dependencies, we
 140 introduce a new practical quantity, the so-called *base counts*:

$$N_{0,\text{base}} = N_0 \times a \times b, \quad (3)$$

141 where the aggregation factor a (in minutes) is the rolling filter window or record period
 142 over which neutron counts are aggregated (Schrön, Zacharias, et al., 2018), and $b \sim \exp(\beta P)$
 143 is the barometric factor which accounts for the cosmic radiation dependency on air pres-
 144 sure P at various altitudes (see Hendrick & Edge, 1966; Desilets et al., 2006, and sup-
 145 plement material S1). The detector used in this study typically runs at $N_0 \approx 200$ cpm
 146 at sea level, which corresponds to base counts of 200–1200 for temporal resolutions of
 147 $a = 1$ –6 minutes, respectively.

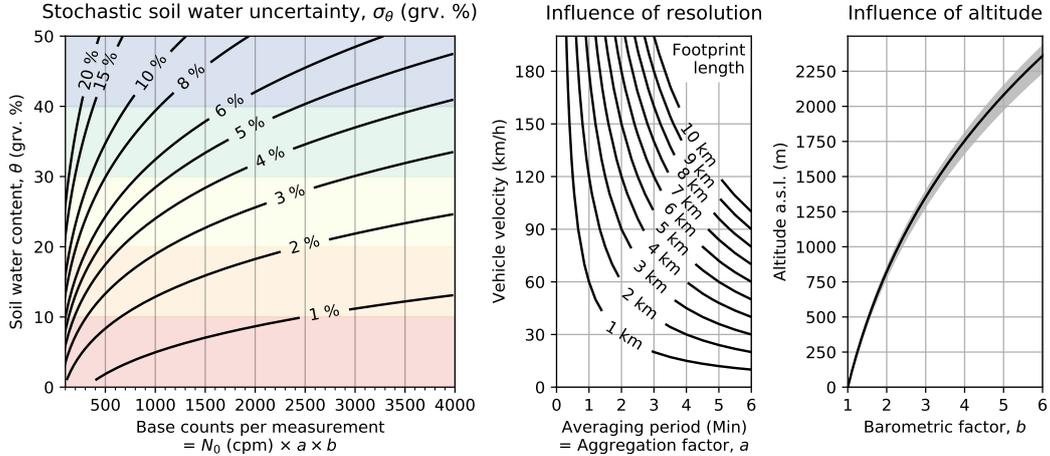


Figure 2. Stochastic measurement uncertainty σ_θ (in absolute gravimetric percent) based on actual soil moisture θ , the type of detector used, target spatiotemporal resolution, and typical topographic elevation. Base counts per measurement denotes the potential total number of counts collected during a given aggregation period at a certain altitude. For example, a detector with characteristic $N_0 = 500$ cpm at sea level could generate soil moisture products with 1–18% uncertainty at 1 minute resolution. With a 3-minute moving average filter (factor $a = 3$) at 820 m altitude (factor $b = 2$), this would be equivalent to 3000 base counts and 0.5–6.1% uncertainty. The temporal aggregation, however, also stretches out the footprint from 1 to 3 km at 60 km/h speed. Shaded area of the barometric line indicates the parameter range of $\beta = 135 \pm 5$ hPa.

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For increasing a , the footprint gets stretched out along the track and thereby increases the spatial scale of the measurement:

$$\text{footprintlength} \quad L = v \cdot a / 60, \quad (4)$$

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With the velocity v (in km/h) and aggregation window a (in minutes) this leads to footprint lengths of several kilometers. This already indicates an imminent tradeoff between the vehicle's travel speed and spatial resolution.

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Figure 2 shows the uncertainty range of the derived soil water content (Eqs. 1 and 2) that can be obtained with different base counts $N_{0,\text{base}}$ (Eq. 3). Due to the non-linearity of Eq. 1, the method provides a range of uncertainties from dry to wet conditions. Higher base counts, e.g. determined by more efficient detectors, longer aggregation periods (middle panel), or higher topographic elevation (right panel), lead to a substantial decrease of σ_θ . In the case of regularly operated train routes, there is even a fourth option to reduce the uncertainty along a track by repeated measurements during the day. The number of repetitions would improve the measurement precision in a similar way as the aggregation factor a .

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2.3 Sensitivity analysis of neutron detection in a train wagon

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The hypothesis that neutron detectors can be used inside a train to measure water content in the surrounding area is not trivial, but has been supported by previous experiments using cars and by theoretical studies of neutron transport physics. In order to more specifically assess potential effects of train-like environments to the detection efficiency, we set up neutron transport simulations with the Monte-Carlo tool URANOS (Köhli et al., 2015, 2018). In the model we defined rails as 25x25 cm steel bars and a 3 or 9 m wide track bed as soil with 6% water content in accordance to data from Ižvolt et al. (2016). Two types of wagons have been modelled, an "open" wagon consisting of a 10 cm thick steel plate at the bottom, and a "closed" wagon with additional ceiling (10 cm

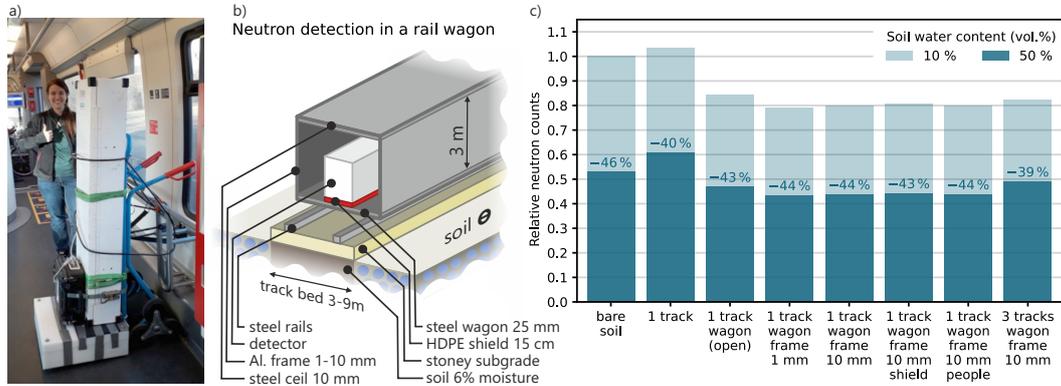


Figure 3. a) The CRNS Rail Rover system mounted on a dolly in a train. b) Simulation design of a rail track and wagon made of steel bottom, steel ceiling, and an aluminium frame. c) Simulation results show the dependency of the neutron count rate on dry and wet soil in various scenarios: 1 track (3 m), 3 parallel tracks (9 m), an open wagon without frames/ceiling, aluminium frames of different widths, 15 cm HDPE shielding below the detector (cmp. photograph), and the presence of people in adjacent wagons at 25 m distance from the (empty) detector wagon. A constant relative span between neutrons under dry and wet conditions (given in %) indicates constant measurement precision and general applicability of the $\theta(N)$ relationship independent of the housing.

172 steel of 1/10 density, i.e., effectively 1 cm thickness) and an aluminium frame at the sides
173 of various widths. The whole train box exhibits a width of 3 m and contains a 1.5×0.5 m
174 detector and an optional 15 cm thick polyethylene plate (HDPE) below. The latter is
175 an experimental method aimed at reducing potential local effects (known as the *road-*
176 *effect*, Schrön, Rosolem, et al., 2018). Humans in adjacent wagons (25 m distance from
177 the detector) may depict an additional source of hydrogen in the sensor’s footprint and
178 have been modeled by 1.5×0.5 m objects of humid gas with 30 kg/m^2 water content.

179 The simulation results in Fig. 3 show the overall count rate for dry soil (light blue)
180 relative to a bare soil scenario (first bar). They also emphasize the expected intensity
181 drop for wet soil (dark blue), which is a measure for the sensitivity to soil moisture changes.
182 The single track scenario (3 m width, second bar) increases the overall count rate, as ex-
183 pected from the known road effect, but simultaneously reduces the sensor sensitivity from
184 46 to 40 %. The steel bottom of the open wagon (bar 3) reduced the count rate by 18 %,
185 but also restocks the sensitivity to 43 %. This bottom material thereby reduces the lo-
186 cal effect of the track bed by half such that even the additional support of the much smaller
187 HDPE shield becomes negligible, as indicated by the sixth bar. In the closed wagon, the
188 steel ceiling also reduces the count rate by a few more percent, while the frame width
189 does not seem to have significant influence on the signal (bars 4–5). Crowded adjacent
190 wagons do not show substantial influence, but two additional rail tracks (total width of
191 9 m) again reduce the sensitivity to soil moisture.

192 It is important to note that measurement sensitivity, i.e. the neutron response to
193 soil moisture between 10 and 50 % volumetric water content, remains almost unchanged
194 throughout all scenarios. Wide track beds have the largest influence on the sensitivity,
195 but this effect is already well understood and correction approaches for roads exist (Schrön,
196 Rosolem, et al., 2018). The results indicate that the standard relationships to convert
197 neutrons to soil moisture (Eq. 1) can be applied for CRNS monitoring using train wag-
198 ons and probably need only very minor adaptations for specific vehicle configurations.

199 3 Experimental proof of concept

200 Experiments along various train tracks have been performed to verify the theoretical
 201 considerations with empirical evidence. The detector used in this study is similar
 202 to the one used by Schrön, Rosolem, et al. (2018) with an original record period of 10
 203 seconds and a 15 cm polyethylene shield at the bottom. It has been fixed in a vertical
 204 mode on a dolly to ensure easy movement by a single person (see Fig. 3).

205 3.1 Detecting landscape features with a regional train

206 The first experiment challenges the hypothesis of whether or not a detector inside
 207 a train is capable of detecting environmental changes of water content outside a moving
 208 train wagon. The railway from Leipzig to Berlin passes several different natural and
 209 urban landscape features at an altitude of ~ 50 m a.s.l. and 30–140 km/h. On a ride in
 210 January 2019 their influence on the neutron response has been observed and shown exemplarily
 211 in Fig. 4. Nearby lakes, swamps, and forests substantially decreased the count rate
 212 (depending on their distance to the railway) due to their higher amount of water content.
 213 An intensity increase has been observed particularly in urban areas and near
 214 highly artificial structures, such as road network infrastructure, where stones, drainages,
 215 and dry and dense soil have been assembled. Underpass stations and underground tracks
 216 led to an extraordinary drop of the count rates, which almost ceased in deeper tunnels.
 217 This expected behaviour is just natural due to the limited penetration depth of atmospheric
 218 cosmic-ray neutrons. Residual neutron intensity can be explained by the natural background
 219 radiation of soils (Missimer et al., 2019) and by the large scattering length
 220 of neutrons in air following underground pathways (Köhli et al., 2015).

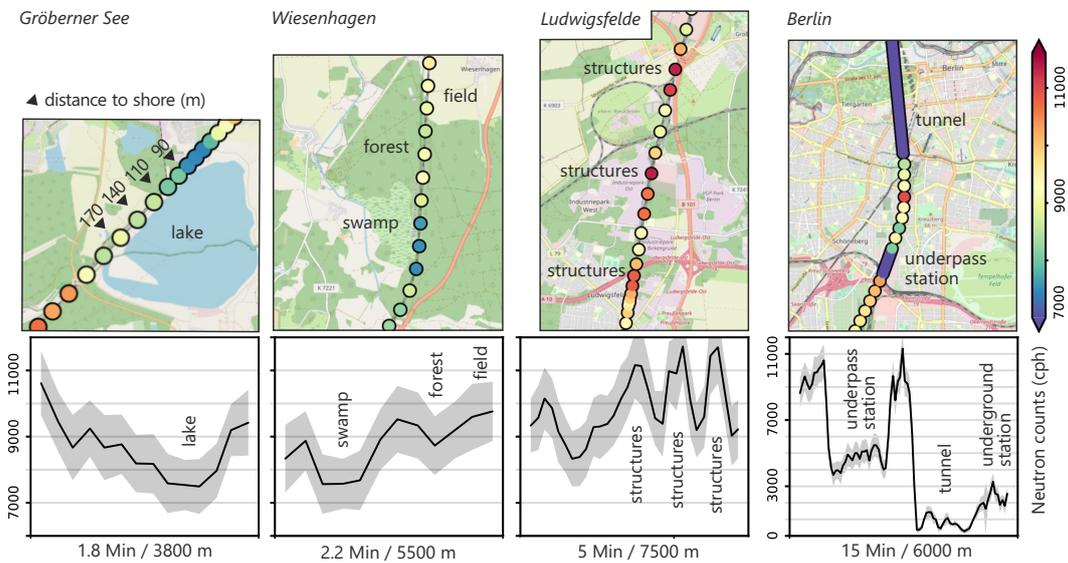


Figure 4. Exemplary sections from the train ride with a regional train from Leipzig to Berlin. Measurement points are visualized along the path at the end of each record period (10 sec). Neutron count rates drop on passing lake *Gröberner See* and on passing a swamp near *Wiesenhagen*. High count rates can be observed near extensive motorway or railway structures, large concrete areas or buildings. The cosmogenic neutron radiation drops substantially at a station underpassing a massive railway bridge, and it almost vanishes in the 14 m deep tunnel below the groundwater level in Berlin. Map background provided by OpenStreetMap.

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3.2 Train-based regional soil moisture measurements and groundtruthing

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Groundtruthing of CRNS measurements on rails with conventional point measurements is a difficult challenge due to the mismatch of scales. To make groundtruthing possible, we used car-borne CRNS roving, which has been extensively validated by various independent measurements already in previous studies (e.g. Schrön, Rosolem, et al., 2018). Between the German cities Dessau and Zerbst, the train route is often crossed by regular roads which in some parts also run in the close vicinity to the track. Measurements were conducted on the train several times a day at ~ 60 m a.s.l. and 60–120 km/h. Afterwards, the sensor was loaded into a car in horizontal mode and without bottom shield, and the railway route was followed on nearby roads. Occasionally, additional soil moisture measurements were taken using a hand-held TDR device at 0–10 cm depth along the track. The sparse data (not shown) ranged between 1 and 7 %, confirming the very dry conditions on that day as well as the overall performance of the mobile CRNS technique in this region.

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Figure 5 presents the CRNS railway measurements (blue) and CRNS rover measurements within < 150 m distance around the railway. The neutron data has been filtered with a 1-minute moving average window and converted to soil moisture using $N_{\max} = 223$ cpm. Car-borne rover processing and parameters were similar to Schrön, Rosolem, et al. (2018) with corrections for 5 m-wide concrete roads. The measurements on rails and on roads show a good agreement along the track despite the only partial spatial overlap, the different type of vehicles used, and potential influence of track beds and passenger fluctuations. The influence of those and other local environmental features to both methods might be substantial (Schrön et al., 2017; Schrön, Rosolem, et al., 2018; Fersch et al., 2018) and may pose some challenge for future signal processing.

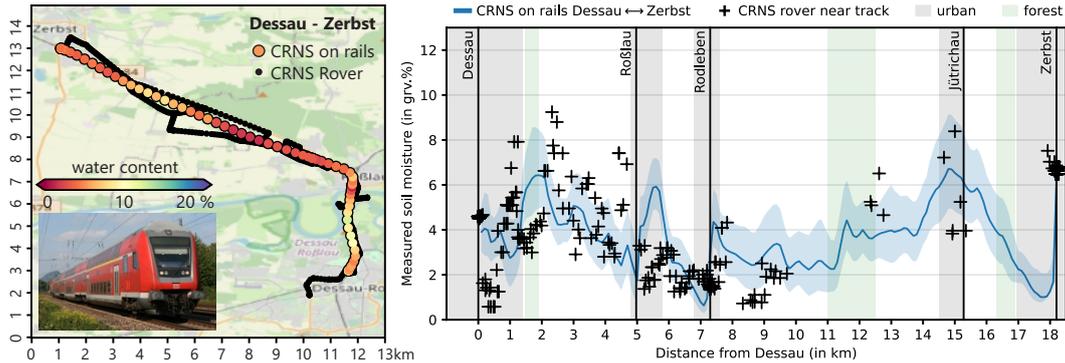


Figure 5. Validation experiment at the rail track between *Dessau* to *Zerbst* with a CRNS system in a local train (blue line and shaded stochastic error) and the same CRNS system in a car (CRNS rover) at accessible roads within < 150 m near the railway (black cross) under particularly dry summer conditions (RMSE=1.9%). The track passed a number of urban areas, nearby forests, and served two railway stations, *Roßlau* and *Rodleben*, which may have led to fluctuations of passengers. Map background provided by OpenStreetMap.

245 3.3 Train-based long-distance measurement of snow water equivalent

246 To examine the capabilities of mobile CRNS on trans-regional railways, we con-
 247 ducted a long-distance experiment in February 2019 using a regional train from Garmisch-
 248 Partenkirchen (pre-alpine, southern Germany) to Munich, and a high-speed train from
 249 Munich to Leipzig (lowland, central Germany). The train journey covered altitudes from
 250 735 to 200 m a.s.l. and ran at 50–250 km/h. Data on air temperature and air humidity
 251 were taken from the German weather services. Incidental gaps in the GPS signal (due
 252 to the electrostatic shielding of train windows) were interpolated linearly.

253 Figure 6 presents the measured neutron counts versus the 5 km product of Snow
 254 Water Equivalent (SWE) based on space-borne microwave radiometry and weather sta-
 255 tion data (Pulliainen, 2006; Takala et al., 2011). Since the separation of snow and soil
 256 water from the signal is still an open research topic, we do not offer a final SWE prod-
 257 uct from CRNS at this stage of the analysis. The data, however, already show a clear
 258 correlation between SWE obtained from remote sensing and neutron counts on rails, up
 259 to the observed maximum of 80 mm SWE, which is in accordance with the theory and
 260 observations made with stationary CRNS sensors (Schattan et al., 2017, 2019; Bogena
 261 et al., 2020). Larger values of snow pack could be expected near Garmisch-Partenkirchen
 262 following the topographic trend, however, these space-borne products are not available
 263 in complex alpine terrain (grey shaded area in the figure). The variability of the neu-
 264 tron counts and SWE measurements can be attributed to the larger CRNS uncertainty
 265 under wet conditions as well as the high heterogeneity of snow pack which is represented
 266 differently by both methods (Schattan et al., 2017).

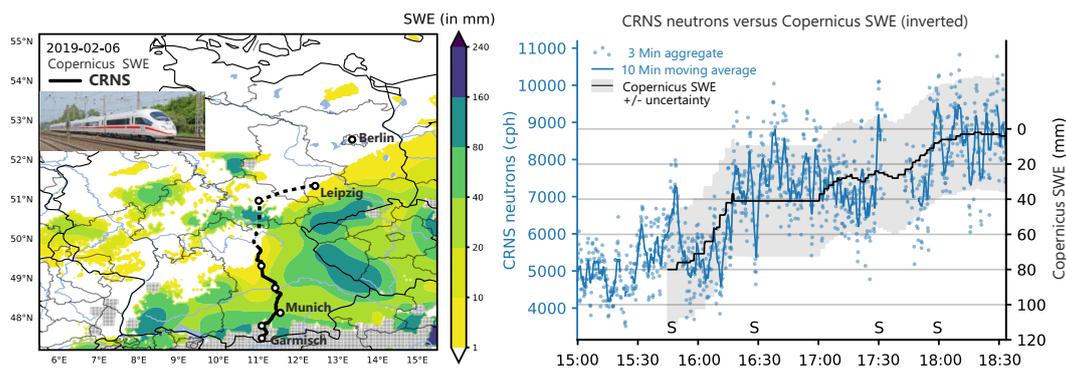


Figure 6. Left: Railway experiment on a long-distance train journey between *Garmisch-Partenkirchen* and *Leipzig* along a clear snow gradient (color scale). Note that mountainous regions cannot be resolved by this remote sensing product (hashed areas). Right: Measured neutrons (blue) show an inverse dependence on surrounding snow water equivalent (SWE) as indicated by the 5 km satellite product (black, inverted). Frequent variability indicates spatial heterogeneity along the railway track which is differently represented by the two methods. The letter "S" indicates railway stations.

267 **4 Future challenges and potential for large-scale geophysics**

268 This study demonstrates a proof of concept for trans-regional monitoring of soil
269 water content and snow with cosmic-ray neutron sensors on rails. Theoretical consid-
270 erations showed that the stochastic precision of derived soil moisture lies in the range
271 of 0.5–3 % or 5–25 % for dry or wet soil, respectively. It largely depends on the detec-
272 tor efficiency, altitude, and averaging period. The latter also influences the spatial res-
273 olution, which could span several kilometers depending on vehicle velocity. Repeated tran-
274 sits along the same route would further contribute to higher measurement precision. Sim-
275 ulations revealed that the sensitivity to soil water changes remains almost unchanged
276 for sensors in trains, suggesting that the same processing approaches are applicable as
277 for the established method of car-borne CRNS roving. Several experiments in different
278 trains provided clear evidence for significant neutron response to surrounding changes
279 of water content and confirmed the theoretical considerations.

280 The footprint of mobile neutron measurements could cover up to half of a km² pixel.
281 While currently there is a lack of alternative geophysical methods which could observe
282 root-zone water content at comparable scales (Binley et al., 2015), further research is needed
283 to resolve a potential spatial scale mismatch to remote-sensing or model resolutions. Fu-
284 ture studies are also required to further test this method under variable conditions. Cal-
285 ibration and validation along the railway will remain a difficult challenge in the future
286 but could be addressed by car-borne roving or mobile soil moisture sensor networks. Fu-
287 ture research should particularly investigate ways to correct the signal for non-hydrological
288 features, e.g. track beds, urban structures, or forest biomass. Artificial intelligence or
289 the application of elementary orthogonal functions could be promising approaches to iden-
290 tify those constant spatial patterns (Finkenbiner et al., 2019). Dynamic train load, such
291 as cargo or human passengers, may have little impact as long as the sensor system is lo-
292 cated in a separate wagon.

293 The method indicates huge potential for large-scale mapping of soil water in the
294 root zone and snow. Due to regular train traffic in public or cargo transport at national
295 and international scales, CRNS on rails has the potential to become a highly distributed
296 technique with minimal infrastructural investment. The method could be particularly
297 useful on slow-moving trains in mountainous regions, where hydrological research is cur-
298 rently very limited by low coverage of observations, while models and satellite products
299 still have issues with complex terrain and weather conditions. Due to the hectare-scale
300 footprint and the potentially large spatial extent, CRNS on rails could make an impor-
301 tant contribution to filling the critical gaps between in-situ data, hydrologic modeling,
302 and remote sensing products. Smart combinations of CRNS on rails, remote sensing, mod-
303 els, and regionalization approaches could have the potential to support near-realtime drought
304 monitors and other forecasting systems beyond the regional scale.

Acknowledgments

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