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Patterns of the relationship between the Secchi disk depth and the optical characteristics of water in the Neva Estuary (Baltic Sea): the influence of environmental variables

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The Secchi disc depth (D_{sd}) measurement is widely used to monitor eutrophication and the quality of the aquatic environment. This study aimed to investigate the relationship between D_{sd} and various factors, including the coefficient of attenuation of photosynthetically active radiation [K_d (PAR)], the depth of the euphotic zone (D_{eu}), PAR at the Secchi disk depth (E_{sd}) and the absorption coefficient of PAR (F) in the Neva Estuary, one of largest estuaries of the Baltic Sea. Environmental variables impacting these indices were identified using data collected from midsummer 2012 to 2020. The D_{sd} values in the estuary ranged from 0.3 to 4.0 m, with an average value of 1.8 m, while the D_{eu}/D_{sd} ratio ranged from 1.5 to 6.0 with an average value of 2.8. These values were significantly lower than those observed in the open waters of the Baltic Sea. The highest D_{eu}/D_{sd} ratio was observed in turbid waters characterized by high K_d (PAR) and low D_{sd} . Contrary to expectations, D_{sd} did not exhibit a significant relationship with the concentration of chlorophyll *a*, raising doubts about the utility of historical D_{sd} data for reconstructing phytoplankton development in the estuary. Principal component analysis did not identify the primary environmental variables strongly affecting the optical characteristics of water. However, recursive partitioning of the dataset using analysis of variance (CART approach) revealed that the concentration of suspended mineral matter (SMM) was the primary predictor of D_{eu}/D_{sd} , K_d (PAR), and F . This SMM was associated with the frequent resuspension of bottom sediments during windy weather and construction activities in the estuary. Concentrations of suspended organic matter and the depth of the water area were found to be less significant as environmental variables. Furthermore, the CART approach demonstrated that different combinations of environmental variables in estuarine waters could result in similar optical indicator values. To reliably interpret the data and determine the optical characteristics of water in estuaries from D_{sd} , more complex models incorporating machine learning and neural connections are required. Additionally, reference determinations of E_{sd} in various regions with specific sets of environmental variables would be valuable for comparative analyses and better understanding of estuarine systems.

KEYWORDS

attenuation coefficient, euphotic zone depth, water transparency, suspended matter, chlorophyll, CDOM, regression trees, Baltic Sea

1 Introduction

Measurement of water transparency is widely accepted and cost-effective method for assessing the quality of aquatic environments in the fields of hydrobiology, limnology and marine ecology (Effler et al., 2017; Kahru et al., 2022; Guo et al., 2022a). This method holds significance due to the crucial role of light necessary for autotrophic organisms to carry out photosynthesis and create primary production by plankton and benthos (Kirk, 2011; Hall et al., 2015; Golubkov and Golubkov, 2022). Furthermore, changes in water transparency have profound effects on trophic interactions within aquatic ecosystems, influencing the dynamics of species relationships (Lunt and Smee, 2014; Reustle and Smee, 2020). By measuring water transparency and the concentration of algal or photosynthetic pigments, it is possible to estimate the primary production of a reservoir and conduct comparative analyses across diverse aquatic environments.

The simplest method for assessing water transparency involves visually determining the depth at which the white Secchi disk becomes invisible to the observer (D_{sd}). This method, originally proposed in the mid-19th century by Secchi (1864), is included in various reference indicators for characterizing the environmental quality of natural waters (Aas et al., 2014) and is widely employed to detect and monitor eutrophication (Fleming-Lehtinen and Laamanen, 2012; Jacobs et al., 2020; Lim and Jeong, 2022; Roy and Das, 2022; Guo et al., 2022b). It has also been included in the widely used metric Trophic State Index (TSI), which can be calculated from both chlorophyll *a* concentration and Secchi disk depth (Carlson, 1977). The Secchi disk use for determining water transparency is facilitated by the fact that the visible light spectrum and the spectrum of photosynthetically active radiation (PAR) (400–700 nm) largely overlap (Kirk, 2011; Neumann et al., 2015). Therefore, D_{sd} is included in the list of indicators of the degree of eutrophication of the Baltic Sea (HELCOM, 2018).

Water transparency is influenced by the physical properties of light dispersion, scattering by water molecules, the presence of dissolved and suspended substances that scatter and absorb light, and the light absorption by autotrophic organisms (Kirk, 2011; Lee et al., 2018; Bowers et al., 2020). As a result, the D_{sd} depends on multiple factors, including the presence of colored organic matter, mineral and organic suspended particles, algal concentration, as well as the angle of sunlight incidence at different latitudes and times of the day (Lee et al., 2015; Bowers et al., 2020; Martin et al., 2021; Guo et al., 2022a).

Preisendorfer (1986) formulated the “10 laws of the Secchi disk”, which encompass the conditions affecting the depth at which the disk disappears from the researcher’s view. However, interpreting Secchi disk measurements can still be challenging and yield contradictory results across different water areas (Harvey et al., 2019). For example, D_{sd} ceased to be a useful indicator of chlorophyll *a* concentration in a eutrophic reservoir when it exceeded 100 mg/m³, but was a good predictor of chlorophyll concentration below this threshold (Golubkov and Golubkov, 2024). To address this issue, various algorithms are being developed to derive D_{sd} from satellite data, aiming to interpret empirical data from diverse conditions in large-scale

areas (Idris et al., 2022; Msusa et al., 2022; Roy and Das, 2022; Guo et al., 2022a). Nonetheless, field studies directly conducted on water bodies are still necessary to obtain empirical data from a variety of locations for further refinement of algorithms (Zhang et al., 2022).

Calculation methods based on the physical principles of light dispersion and the optical properties of water suggest that D_{sd} should correspond to approximately 10% of the PAR reaching the water surface (Preisendorfer, 1986; Kirk, 2011). The degree of PAR attenuation with depth is estimated using the diffuse attenuation coefficient [Kd(PAR)]. The depth at which only 1% of the incident PAR remains is known as the euphotic zone depth (D_{eu}), representing the zone where active photosynthesis can occur. Having determined the Kd(PAR) value, it is easy to calculate the amount of PAR at any depth, as well as the depth of the euphotic zone (Kirk, 2011). This coefficient is also widely used in remote sensing of water areas, since its value can be determined from satellite measurements of radiation at a wavelength of 490 nm [Kd(490)] in the upper meter layer of water (Pierson et al., 2008).

According to Poole and Atkins (1929) Kd(PAR) can be calculated from D_{sd} using the formula:

$$Kd(PAR) = \frac{F}{D_{sd}} \quad (1)$$

where Kd(PAR) is the PAR diffuse attenuation coefficient, F is the PAR absorption coefficient, and D_{sd} is the depth at which Secchi disk disappears from the observer’s field of view.

Poole and Atkins (1929) initially suggested a value of F=1.7, but Holmes (1970) demonstrated that F=1.4 is more suitable for turbid estuaries. Preisendorfer (1986) calculated that the depth at which 10% of the surface PAR remains ($D_{10\%}$) and D_{sd} are the same at F=2.29. The range of published values of F in marine waters typically falls within 1.3–2 (Lee et al., 2018). In estuaries and coastal lagoons with turbid water, the F value is approximately 3 (Bowers et al., 2020). In comparison, colored lakes with high concentrations of colored dissolved organic matter (CDOM) have an average F value of 2.7, clear lakes 1.99, and turbid lakes have 1.05 (Koenings and Edmundson, 1991). In the Baltic Sea, F ranges from 1.7 to 2.3 depending on the region (HELCOM, 2002). Additionally, some studies suggest that F increases as salinity decreases, indicating a relationship between F and continental runoff of CDOM (Kratzer et al., 2003).

According to published data, the depth of the euphotic zone can range from 1 to 10 times the Secchi disk depth (Koenings and Edmundson, 1991; Kirk, 2011; Luhtala and Tolvanen, 2013). The D_{eu}/D_{sd} ratio is approximately 1.7 in colored lakes with high CDOM concentrations, 2.4 in clear lakes, and 4.8 in turbid ones (Koenings and Edmundson, 1991). Generally, D_{eu}/D_{sd} ratio of 2.4 is considered the average for all world waters (Lee et al., 2018). However, recent studies indicate that for clear sea waters with depths exceeding 10 meters, a D_{eu}/D_{sd} ratio of 3.5 should be used (Lee et al., 2018), which roughly aligns with the ratio observed in turbid lake waters (Koenings and Edmundson, 1991). This example highlights the inconsistencies in available data and conclusions, as noted by Harvey et al. (2019).

In the context of climate change and the productivity of aquatic ecosystems, understanding past photosynthetic conditions is crucial

for improving forecasting accuracy (Boyce and Worm, 2015). Given the extensive historical data on Secchi disk depth accumulated over the past two centuries, there is potential to reconstruct eutrophication levels and changes in phytoplankton productivity using historical D_{sd} data (Fleming-Lehtinen and Laamanen, 2012; Lee et al., 2018). For instance, in the Neva Estuary, the first measurements of water transparency using the Secchi disk were conducted as early as 1911 (Zalessky and Wulf, 1913). One may also agree with Angradi et al. (2018) that Secchi disk depth units (meters) are easier to understand by non-scientists and managers who routinely monitor and make decisions about the aesthetic value of an aquatic area, its public health and economic importance, benefits associated with the development of tourism. However, reconstructing the eutrophication process in aquatic ecosystems based on D_{sd} data requires a more detailed knowledge and models to characterize the relationship of D_{sd} with the optical properties of water, determined by the presence of impurities and suspended substances, including phytoplankton algae. Previous studies have demonstrated that even slight impurities in the water, unrelated to algae, can result in inaccurate assessments of water eutrophication and chlorophyll *a* concentrations derived from Secchi disk depth measurements (Lind, 1986; Harvey et al., 2019).

The aim of this study was to establish the relationships between D_{eu} , PAR, $K_d(\text{PAR})$ and D_{sd} values, and to determine the most appropriate *F* coefficient for estuarine waters. We also examined changes in D_{sd} , D_{eu}/D_{sd} and *F* across different parts of the estuary and identified the environmental factors that exert the greatest influence on these variables. The resulting data and relationships were employed to discuss the feasibility of reconstructing the optical characteristics of water using historical D_{sd} data in the Neva Estuary.

2 Materials and methods

2.1 Study site

The Neva Estuary (Figure 1) is located in the northeastern part of the Baltic Sea. It serves as the outlet for the Neva River, which originated from Lake Ladoga and flows into the Gulf of Finland. The Neva River has an average discharge of $2,490 \text{ m}^3 \text{ s}^{-1}$ ($78.6 \text{ km}^3 \text{ yr}^{-1}$), and its catchment area spans over $280,000 \text{ km}^2$ (Golubkov and Golubkov, 2020). Geographically, the estuary is located at the northern boundary of the temperate zone and at the southern boundary of the subpolar zone (Meteoblue, 2023). Climate type in the region is classified as Dfc — Snowy climate, fully humid, with cool summers according to Köppen-Geiger climate classification (Kottek et al., 2006).

The Neva Estuary is a shallow and non-tidal system, exhibiting a gradual transition from freshwater in the upper part to slightly saline water in the lower part (Golubkov and Golubkov, 2020). Due to its proximity to the 5-million-strong city St. Petersburg, the estuary is subject to significant anthropogenic impacts (Golubkov et al., 2019). To mitigate the risk of flooding, the Flood Protective Facility, consisting of 11 dams, was constructed in the late 1980s, effectively separating the upper part (UP) of the estuary from its

middle part (MP) (Figure 1). The UP of the estuary lacks temperature stratification, and the presence of low water transparency and strong currents hinders the development of benthic aquatic vegetation. The MP of the estuary, located between Kotlin Island and the arbitrary boundary of $29^\circ 10' \text{ E}$, experiences temperature stratification during summer season. For a comprehensive understanding of ecological characteristics of the Neva Estuary and a detailed description of both its parts, refer to Golubkov and Golubkov (2021).

2.2 Data and methods

In this study, we used data from long-term scientific monitoring of the Neva Estuary ecosystem. The data used for analysis collected in the upper and middle parts of the estuary (Figure 1) during late July-early August 2012–2020. Salinity (*S*) and temperature (*T*) were conducted using a CTD90M probe from Sea&Sun Tech (Germany). The concentration of colored dissolved organic matter (CDOM) and chlorophyll *a* (CHL), as well as turbidity (Turb), were determined using Cyclop-7 sensors connected to a submersible C-6 multi-sensor platform (TurnersDesigns, United States). Prior to measurements, the Cyclop-7 sensors were checked and calibrated using TurnersDesigns solid standard. In the case of CDOM, the Cyclop-7 sensor was additionally calibrated based on the corrections proposed by Downing et al. (2012). Test material of humic substances (HS) derived from typical soils in the Neva Estuary watershed was obtained from the Department of Soil Science and Soil Ecology of St. Petersburg State University. HS solutions were prepared by dissolving 1 g of test material in 1 L of organic-free deionized water, after which it was diluted to several solutions with HS concentration of 0.0005, 0.001, 0.002, 0.003, 0.004, 0.005, 0.007, 0.01 mass percent. Fluorimeter measurements were carried out following the method described by Downing et al. (2012), and the carbon concentration in these solutions was measured using high-temperature catalytic oxidation with Shimadzu TOC-L_{CPN/CSN} instrument (Shimadzu Scientific Instrument, Japan) based on the method by Bird et al. (2003).

The PAR values were determined using a LiCOR-193SA scalar quantum sensor connected with a tripod to a CTD90M probe and controlled using Sea&Sun Tech software. The LiCOR-193SA sensor was calibrated every 2 years (according to recommendations of manufacture) using spectrophotometer and reference silicon photodiodes traceable to the National Institute of Standards and Technology (NIST).

All probes were programmed to take readings at 1-second intervals. To minimize errors related to waves, the readings were averaged over standard intervals in 10-centimeter increments. The PAR value on the water surface (PAR_0) was taken as 100%, and the decrease in PAR with depth was expressed as a percentage of PAR_0 .

The depth at which the PAR value reached 10% of the PAR_0 value was designated as the depth of 10% PAR_0 ($D_{10\%}$). The depth at which the PAR value reached 1% of PAR_0 was accepted as the depth of the euphotic zone (D_{eu}). The number of photons reaching this depth ($\text{PAR}_{D_{eu}}$) was also measured. The $K_d(\text{PAR})$ was calculated from measurements of the amount of incident radiation near the

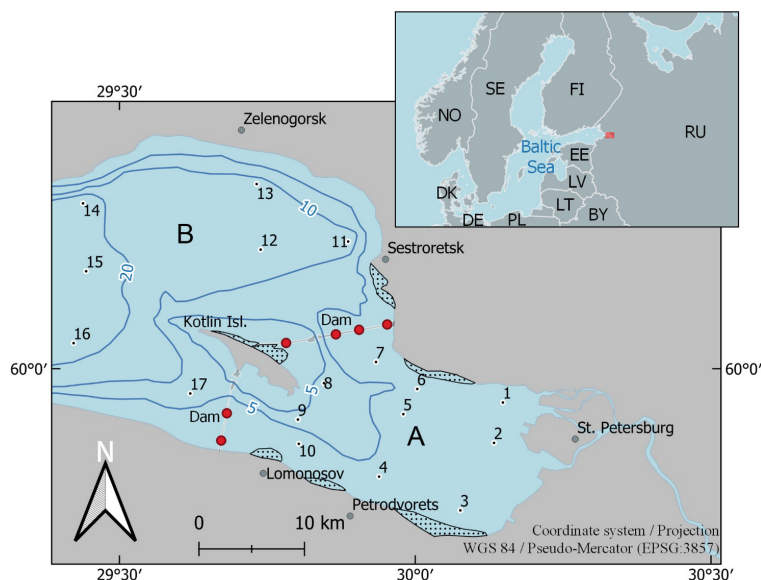


FIGURE 1

The upper (A) and middle (B) parts of the Neva Estuary with indication of sampling stations in midsummer 2012–2020. Blue lines: isobaths of 5, 10, and 20 m. Areas with dots indicate dense reeds. Dam — the St. Petersburg Flood Protection Facility. Red circles - waters gates in the Dam. Red rectangle in the top block of the map — the location of the Neva Estuary. Two-letter country codes are given according to ISO 3166-1 alpha-2 (International Organization for Standardization (ISO), 2023).

water surface and of the euphotic zone (Kirk, 2011):

$$Kd(PAR) = -\frac{1}{z} \ln \frac{PAR_{D_{eu}}}{PAR_0} \quad (\text{Equation 2})$$

where PAR_0 and $PAR_{D_{eu}}$ are the PAR just below the surface and at D_{eu} , respectively.

Data on $Kd(PAR)$ and D_{eu} were published and discussed earlier (Golubkov and Golubkov, 2023).

The Secchi disk depth (D_{sd}) was measured using a 30 cm matt white disk from the shaded side of the boat. Since the depth of the Secchi disk and the vertical profile of PAR were measured simultaneously, we were able to determine the proportion of PAR from the water surface reached D_{sd} (E_{sd}). The coefficient F was calculated by the transformed Equation 1:

$$F = Kd(PAR) \times D_{sd} \quad (3)$$

The concentrations of total suspended matter (SM) and suspended organic matter (SOM) were also determined, as these variables affect the optical properties of the water column. The concentration of SM in water was obtained using a standard gravimetric method (Grasshoff et al., 1999). The SOM concentration was determined using the dichromate oxidation method of a suspension pre-deposited on preheated (1 h, 105°C) and weighed GF/F filters (Grasshoff et al., 1999). Suspended mineral matter (SMM) concentrations were determined by subtracting SOM values from SM values. Water samples for determining SOM and SM were collected using a 2-liter bathometer from the surface, half a meter above the bottom and from three equidistant horizons between them in the UP of the estuary. In the MP of the estuary, samples for SOM and SM were taken from a horizon above the thermocline, which was determined using CTD90M probe. The

samples were obtained from the surface, a depth of 0.5 m above the thermocline, and three equidistant horizons. This allowed us to obtain composite samples of 10 l, from which 0.5 l of water was taken for laboratory determination of SOM and SM.

2.3 Statistical analysis and data modeling

The D_{sd} , $D_{10\%}$, subtraction value between D_{sd} and $D_{10\%}$, and the ratio between D_{eu} and D_{sd} were averaged for each station over 2012–2020 and visualized with SURFER 8.0 using the kriging data extrapolation model. A total of 153 data points were used for this visualization, including 18 stations over 9 years.

Untransformed data was used for statistical analysis. Data from stations where the PAR penetration depth was limited by the bottom were excluded from regression analysis, this resulted in a matrix of 89 data points (Supplementary Tables 1S, 2S). Regression analysis was performed in Microsoft Excel. We used the same data set for this analysis as in Golubkov and Golubkov (2023).

Part of the SOM data was missing in 2020; in this regard, we analyzed in principal component analysis, recursive splitting and building of regression trees only data for 2012–2019. Some stations also did not have all indicators. As a result, for principal component analysis, recursive splitting and building of regression trees, we got a final matrix of 67 data lines for each indicator (Supplementary Tables 1S, 3S).

Principal component analysis (PCA), as well as classification and regression tree analysis (CART) were performed using R software (version 4.2.2) (R Core Team, 2023). PCA analysis was applied to elucidate the relationship between the optical characteristics of water and environmental variables, which was performed using the

“prcomp” command of the “Stat” package (R Core Team, 2023), and the results were visualized using the “fviz_pca_biplot” command of the package “factoextra” (Kassambara and Mundt, 2020). The script and the full output of principal component analysis are shown in Supplementary Table 4S.

Regression trees were constructed using recursive partitioning of data based on analysis of variance (Breiman, 1984). CART is a prediction algorithm used in machine learning to predict outcomes based on predictor variables. The decision tree construction algorithm was performed from top to bottom by choosing a variable at each step that best splits the set of elements, which is the most commonly used method (Rokach and Maimon, 2005). It explains how the values of the target variable can be predicted based on other values. The goal of CART is to develop a sequence of partitioning rules to split response variable observations (i.e., Secchi disk depths) into relatively small, homogeneous groups of observations using related explanatory variable values. This is a decision tree where each fork is split into a predictor variable and each node at the end has a prediction for the target variable.

In this work, the decrease in the variance of the dependent variable was chosen as a measure of the best solution, for which the analysis of variance was used. This process, which is repeated recursively on each resulting subset, is called recursive partitioning. The recursion is terminated when a subset in a node has the same target variable value, or when the split adds no value to the predictions. This top-down induction of decision trees (TDIDT) process is the most commonly used strategy for learning decision trees from data (Quinlan, 1986).

The tree models were built using the “rpart” command of the “rpart” package (Therneau et al., 2022) using the “anova” partitioning method, i.e. search for a solution with the smallest variance of the dependent variable at the node. The “rpart.plot” command from the “rpart.plot” package (Milborrow, 2022) was used to visualize the regression trees. The script and the full output of CART are shown in Table 5S.

3 Results

3.1 Secchi disk depth

The Secchi disk depth changed seven times in the midsummer, averaging 1.4 m (Table 1). The lowest D_{sd} values were observed in the coastal area of the UP of the estuary, while the highest values were found at the confluence of the Neva River into the UP of the estuary and in the seaward western region of the MP of the estuary (Figure 2A). The amount of PAR at the Secchi disk depth (E_{sd}) averaged about 15% of the PAR at the water surface, but the range of E_{sd} values was from 2 to 45%.

The spatial distribution of D_{sd} and $D_{10\%}$ practically coincide (Figures 2A, B), and regression analysis confirmed a significant relationship between these indicators (Figure 3A). However, D_{sd} was generally less than $D_{10\%}$ (Table 1, Figures 2C, 3A). The largest differences were observed in the seaward part of the studied water area and in the northwestern UP of the estuary, where D_{sd} was more than 0.5 m less than $D_{10\%}$ (Figure 2C).

Although D_{sd} and $D_{10\%}$ are related, their relationship in the Neva Estuary deviates from the theoretical linear trend (Figure 3A).

Instead, a logarithmic dependence of D_{sd} on $D_{10\%}$ demonstrated the highest coefficient of determination. The theoretical and empirical trend lines closely matched for $D_{10\%}$ values ranging from 1 to approximately 2.5 m, but noticeable divergence occurred for $D_{10\%}$ values below 1 and above 3 m (Figure 3A). In the Neva Estuary, the median E_{sd} value was 12.9% (Table 1), which is proximity to the theoretical $E_{sd} = 10\%$. However, E_{sd} in the Neva Estuary was not constant. We obtained a statistically significant linear relationship between D_{sd} and E_{sd} (Figure 3B). This relationship revealed that at low D_{sd} values, E_{sd} could reach up to 40% (Figure 3B). Conversely, at the highest D_{sd} values, E_{sd} in some cases was less than 5% (Figure 3B).

3.2 Relationship between the Secchi disk depth and the depth of the euphotic zone

The depth of the Secchi disk in the Neva Estuary was, on average, 2.8 times less than the depth of the euphotic zone, while the minimum and maximum values differed four times (Table 1). The diffuse PAR attenuation coefficient was related by a power function to D_{sd} (Figure 4A). D_{sd} decreased most strongly at $Kd(PAR)$ values up to 2 m^{-1} , at which D_{sd} was approximately 0.5 m, and a further increase in $Kd(PAR)$ led to a slight decrease in D_{sd} (Figure 4A). Although D_{sd} was significantly linearly related to D_{eu} , the scatter of D_{eu} values increased with increasing Secchi disk depth (Figure 4B). For example, with D_{sd} about 2 m, D_{eu} could be both 3.5 and 8 m. There was also a significant linear relationship between D_{eu}/D_{sd} and E_{sd} (Figure 4C). At stations where 30-40% of the PAR reached the Secchi disk depth, the D_{eu} was three to six times greater than the D_{sd} (Figure 4C).

The coefficient F showed an inverse proportional relationship with E_{sd} (Figure 4D). High E_{sd} values of around 40% corresponded to a F value of approximately 1 m^{-1} , while E_{sd} values close to 5% resulted in a F value about 4 m^{-1} (Figure 4D). The spatial distribution of the D_{eu}/D_{sd} ratio revealed that, on average, the highest values of the ratio were observed in the southeastern parts of the UP and the MP of the estuary (Figure 2D). This indicates that there was no consistent increasing or decreasing the D_{eu}/D_{sd} ratio with distance from the river mouth or from the coast towards the central part of the estuary (Figure 2D). The D_{eu}/D_{sd} ratio was related to the F coefficient by a power function (Figure 5). The relationship showed that the D_{eu}/D_{sd} ratio reached its maximum when the F coefficient values were at their minimum, which occurred in areas with high $Kd(PAR)$ and low D_{sd} . Conversely, when $Kd(PAR)$ values were low and D_{sd} values were high, the D_{eu}/D_{sd} ratio was low (Figure 5). In other words, in more transparent waters of the Neva Estuary, the difference between the depth of the euphotic zone and the Secchi disk depth decreased.

3.3 Relationships of environmental variables with the values of the D_{eu}/D_{sd} ratio and the coefficient F

The minimum, maximum, mean and median values of the environmental variables used in the statistical analysis are shown in Table 2.

TABLE 1 The Secchi disk depth (D_{sd}), the depth at which 10% of the PAR incident the water surface remains ($D_{10\%}$), the proportion of PAR at the Secchi disk depth (E_{sd}) of the PAR on the water surface, the ratio of the depth of the euphotic zone to the Secchi disk depth (D_{eu}/D_{sd}) and the value of the absorption coefficient of PAR (F).

Index	Number	Minimum	Maximum	Average	Median	SE
D_{sd} [m]	89	0.3	2.1	1.4	1.4	0.01
$D_{10\%}$ [m]	89	0.5	4.4	1.8	1.7	0.01
E_{sd} [% from surface]	89	2.5	43.4	14.8	12.9	0.20
D_{eu}/D_{sd}	89	1.5	6.4	2.8	2.6	0.02
F [m^{-1}]	89	0.8	3.7	2.1	2.1	0.01

m, meter; Number, the number of values; SE, standard error.

Principal component analysis demonstrated that the D_{eu}/D_{sd} ratio could exceed the median value of 2.6 for any combination of the studied factors (Figure 6). According to PCA, vectors of turbidity change, SMM concentration, and percent PAR at the Secchi disc depth were positively related to each other and negatively related to CDOM concentration and the values of F coefficient. At the same time, Turb, SMM, and E_{sd} did not depend on the water area depth (D) and the salinity of the waters, which in this case could be taken as a measure of the distance from the mouth of the Neva River. This suggests that very different values of these indicators can be observed in both the seaward and deeper part of the estuary and in the shallow part near the mouth of the Neva River (Figure 6).

The distribution of D_{eu}/D_{sd} values showed more localized changes within the boundaries of $D_{eu}/D_{sd} < 2.6$, which was

contained within the distribution ellipse of $D_{eu}/D_{sd} > 2.6$ (Figure 6). This means that the D_{eu}/D_{sd} ratio can be high and low for the same environment variables. It is noteworthy that the vectors of depth and salinity changes in the estuary were elongated along the ellipse $D_{eu}/D_{sd} < 2.6$, which indicates that the D_{eu}/D_{sd} ratio can be low both in the seaward and saline part of the estuary and in its shallow freshwater part (Figure 6).

Recursive partitioning based on analysis of variance provided more interpretable results. It revealed that the concentration of mineral suspended matter was the main factor contributing to the variance in D_{sd} (Figure 7A). When the SMM concentration exceeded 3.9 g m^{-3} , the average D_{sd} was about 0.9 m, whereas it was 1.6 m when the concentration was below 3.9 g m^{-3} (Figure 7A). CDOM concentration and estuary depth were the next significant predictors of D_{sd} variance. When the SMM concentration was below 3.9 g m^{-3} ,

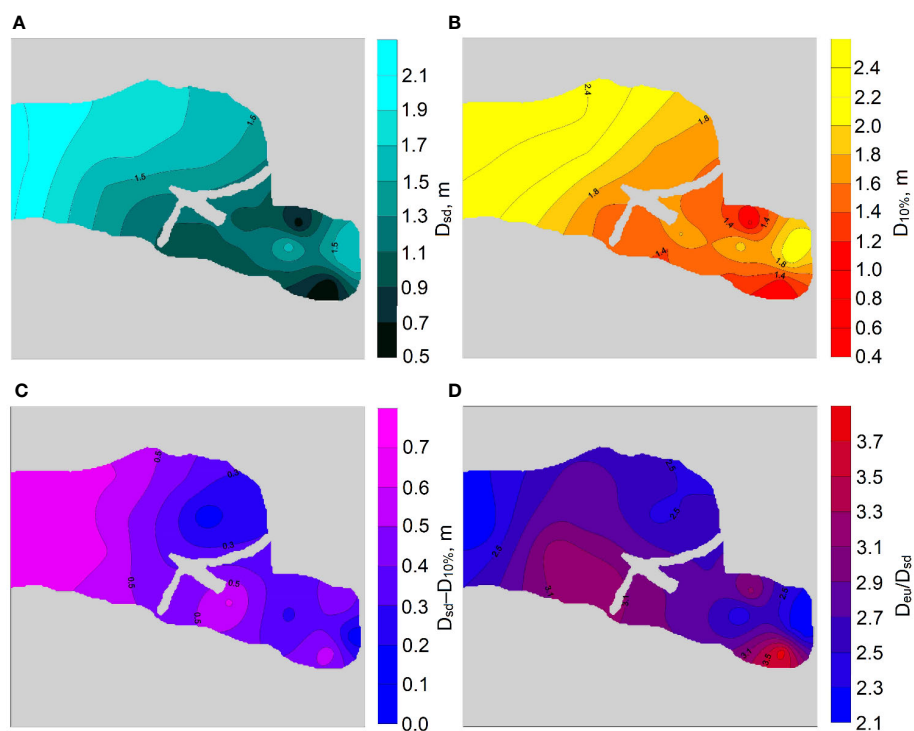


FIGURE 2

Average values of the Secchi disk depth (D_{sd}) (A) and the depth at which 10% of the surface PAR remains ($D_{10\%}$) (B), as well as the difference between $D_{10\%}$ and D_{sd} (C), and the ratio of the depth of the euphotic zone to the Secchi disk depth (D_{eu}/D_{sd}) (D) in midsummer 2012–2020.

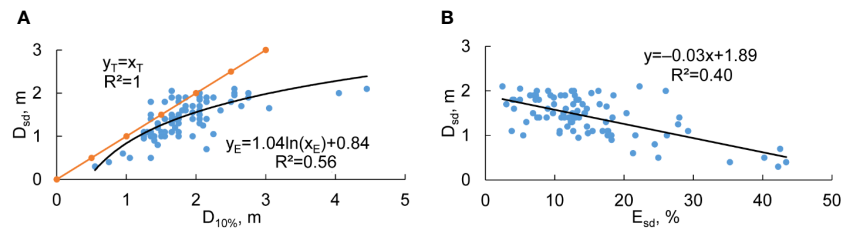


FIGURE 3

Relationship between the Secchi disk depth (D_{sd}) and the depth at which 10% of the surface PAR remains ($D_{10\%}$) (A), and the percentage of the surface PAR at the Secchi disk depth (E_{sd}) (B). The black line is the trend line of the empirical data, the orange line is the theoretical dependence, where it is assumed that $D_{10\%} = D_{sd}$. subscript "T" specifies theoretical data, subscript "E" indicates empirical ones. R^2 – coefficient of determination.

D became the second most influential factor, while CDOM played this role when the SMM concentration exceeded 3.9 g m^{-3} . Additionally, for depths greater than 11 m and SMM concentration below 3.9 g m^{-3} , the dispersion of D_{sd} was significantly determined by chlorophyll a concentration (Figure 7A).

In contrast to D_{sd} , the variance in D_{eu} values was primarily explained by two predictors: estuary depth and SMM concentration (Figure 7B). Estuary depth was the main environmental factor influencing the dispersion of D_{eu} . For stations with depths exceeding 11 m and SMM concentration below 1.5 g m^{-3} , the average D_{eu} was around 5 m. In the shallow parts of the estuary, where the depth was less than 11 m and SMM concentration exceeded 3.9 g m^{-3} , the average D_{eu} value was about 2.6 m (Figure 7B). Therefore, although D_{sd} and D_{eu} were linearly related (Figure 4B), the set of environmental factors influencing their variance was slightly different (Figure 8), which likely explains the high variability observed in D_{eu} values at the same D_{sd} during linear regression (Figure 4B).

The D_{eu}/D_{sd} ratio was mainly determined by changes in D_{sd} rather than D_{eu} (Figure 8A). The regression tree constructed for the D_{eu}/D_{sd} ratio, using environmental factors as predictors, showed

that SMM concentration was the primary predictor determining the value of this ratio (Figure 8B). When SMM concentration exceeded 7.5 g m^{-3} and D_{sd} was below 0.85 m, the D_{eu}/D_{sd} ratio was around 4 (Figures 8A, B). A more complex relationship was observed for lower SMM concentrations and higher D_{sd} values where different combinations of mineral and organic suspended matter concentrations, estuary depth and differences between D_{sd} and $D_{10\%}$ or D_{eu} led to similar D_{eu}/D_{sd} values (Figure 8).

For example, the average D_{eu}/D_{sd} ratio was about 2.7 when:

1. the SMM was below 7.5 g m^{-3} , the SOM concentration was greater than or equal to 2.3 g m^{-3} , and the depth was above 13 m;
2. the SMM was below 7.5 g m^{-3} and SOM concentration was less than 1.8 g m^{-3} , with estuary depth not affecting the variance of D_{eu}/D_{sd} ratio (Figure 8B).

The lowest D_{eu}/D_{sd} values around 2.0 were observed at concentrations of $\text{SMM} < 1.4 \text{ g m}^{-3}$ and $\text{SOM} > 2.3 \text{ g m}^{-3}$, while the $D < 13 \text{ m}$, $D_{sd} > 0.85 \text{ m}$, $D_{eu} < 3.5 \text{ m}$, and D_{sd} differed from the theoretical $D_{10\%}$ less than 0.28 m (Figures 8A, B). These findings

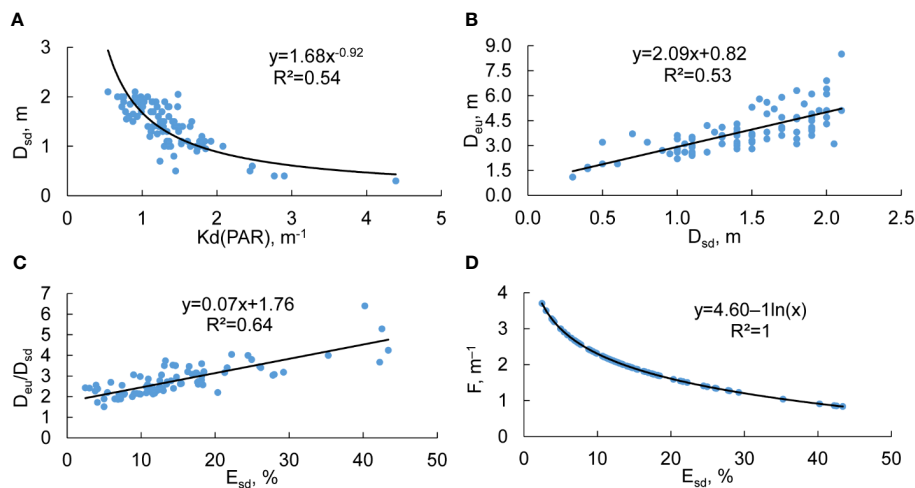
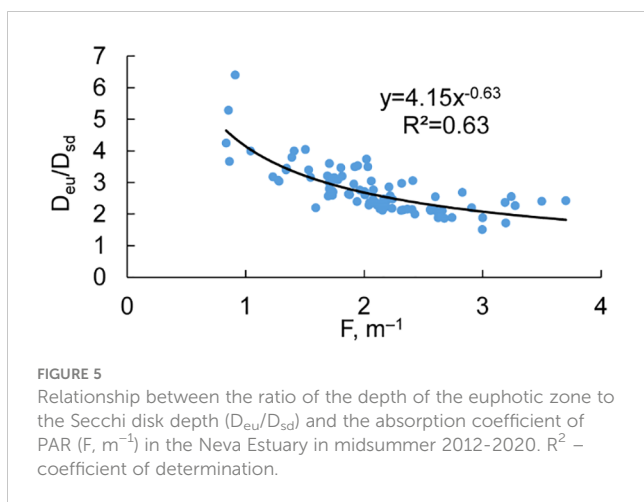


FIGURE 4

Relationship between the Secchi disk depth (D_{sd}) and the diffuse PAR attenuation coefficient [$Kd(\text{PAR})$] in the euphotic layer (A) and the depth of this layer (D_{eu}) (B); the dependence of D_{eu}/D_{sd} ratio (C) and absorption coefficient of PAR (F) (D) on the percentage of PAR at the Secchi disk depth (E_{sd}). R^2 – coefficient of determination.



indicate that the calculation of D_{eu} from D_{sd} values in estuaries using simple linear regressions is hardly justified, as real data can deviate significantly from theoretical predictions.

The dispersion of the F coefficient, which relates with the optical characteristics of water, was explained by the amount of PAR remaining at the Secchi disk depth. When E_{sd} was less than or equal to 10% of surface PAR, the averaged F was $2.8 m^{-1}$, while it was $1.2 m^{-1}$ when E_{sd} exceeded 20% (Figure 9A).

The main predictor associated with low F coefficient and a high D_{eu}/D_{sd} ratio was an SMM concentration greater than $7.5 g m^{-3}$ (Figures 8B, 9B). Conversely, a high F coefficient was observed when the SMM concentration was less than $0.8 g m^{-3}$, but the SOM concentration was greater than $2.4 g m^{-3}$; indicating that the organic fraction constituted about 75% of the total suspended matter. It was challenging to predict precise F values based on the studied environmental factors, as different combinations of the same factors led to overlapping F values (Figure 9B). For instance, the mean F value was $1.8 m^{-1}$ when the SMM concentration was below $7.5 g m^{-3}$, SOM concentration was below $2.3 g m^{-3}$, and depth was less than 12 m. In the deeper parts of the estuary with depths greater than 12 m, this coefficient

averaged $2.0 m^{-1}$. Similarly, the average F value was $1.9 m^{-1}$ when the SMM concentration was less than $7.5 g m^{-3}$, but above $0.8 g m^{-3}$, and the SOM concentration ranged above $3.0 g m^{-3}$. Under these conditions, the depth did not significantly influence the variance of the F coefficient. In general, the value of the F coefficient in different parts of the Neva Estuary was determined by a combination of E_{sd} , SMM, SOM values and, in some cases, the depth of the water area (Figure 9).

Using the CART approach, it was found that the concentration of SMM was the main predictor of E_{sd} in different parts the estuary. When the SMM concentration exceeded $7.5 g m^{-3}$, an average of 28% PAR from the surface remained at the Secchi disk depth, significantly higher than the theoretical value of 10% PAR. At SMM concentrations below $7.5 g m^{-3}$, the variance of the E_{sd} indicator was determined by various combinations of three variables: depth, SMM concentration, and SOM concentration. The lowest percentage of PAR from the surface at the Secchi disk depth was observed at SOM concentrations greater than $2.3 g m^{-3}$ and SMM concentrations less than $0.8 g m^{-3}$ (Figure 10A). Under these conditions, E_{sd} averaged 7.3%. This suggests that the Secchi disk disappeared from view at a later depth than predicted by the theoretical relationship, where 10% of the PAR value on the water surface should remain at the Secchi disk depth. This case corresponds to the regression points of the $D_{sd}-D_{10\%}$ relationship located above the orange line in Figure 3A when $D_{sd} > D_{10\%}$. Overall, the difference between $D_{10\%}$ and D_{sd} was mainly determined by water salinity and SMM concentration. At water salinity above 0.73 ‰ and SMM concentrations above $1.8 g m^{-3}$, the depth at which 10% of the surface PAR remained was 0.94 m greater than the Secchi disk depth (Figure 10B). The smallest difference between these indicators was observed at salinity from 0.34 to 0.73 ‰, while the largest difference occurred in more saline waters (Figure 10B).

Based on the data, the Secchi disk depth did not show a statistically significant relationship with chlorophyll a concentration in the upper and middle parts of the Neva Estuary (Figure 11A). Therefore, it is difficult to reconstruct historical phytoplankton

TABLE 2 Morphometric, physio-chemical, and biological indicators at sampling stations in the Neva Estuary in late July – early August 2012–2020.

Index	Minimum	Maximum	Average	Median
D [m]	2.5	25.5	11.6	10.4
D_{11-17} [m]	6.5	25.5	14.9	13.0
DT_{11-17} [m]	2.1	19.2	8.2	7.2
S [PSU]	0.06	2.24	0.51	0.38
Turb [NTU]	2.28	65.68	9.29	6.16
CDOM [$g m^{-3}$]	7.06	21.05	13.75	13.56
SOM [$g m^{-3}$]	1.10	6.66	2.53	2.45
SMM [$g m^{-3}$]	0.21	20.15	3.40	1.92
CHL [$mg m^{-3}$]	8.82	85.47	20.42	15.81

D is water depth (m); D_{11-17} , is water depth at stations 11–17; DT_{11-17} , thermocline depth at stations 11–17; S, salinity; Turb, turbidity; CDOM, colored dissolved organic matter; SM, suspended particulate matter; SOM, suspended particulate organic matter; SMM, suspended particulate mineral matter; CHL, chlorophyll a .

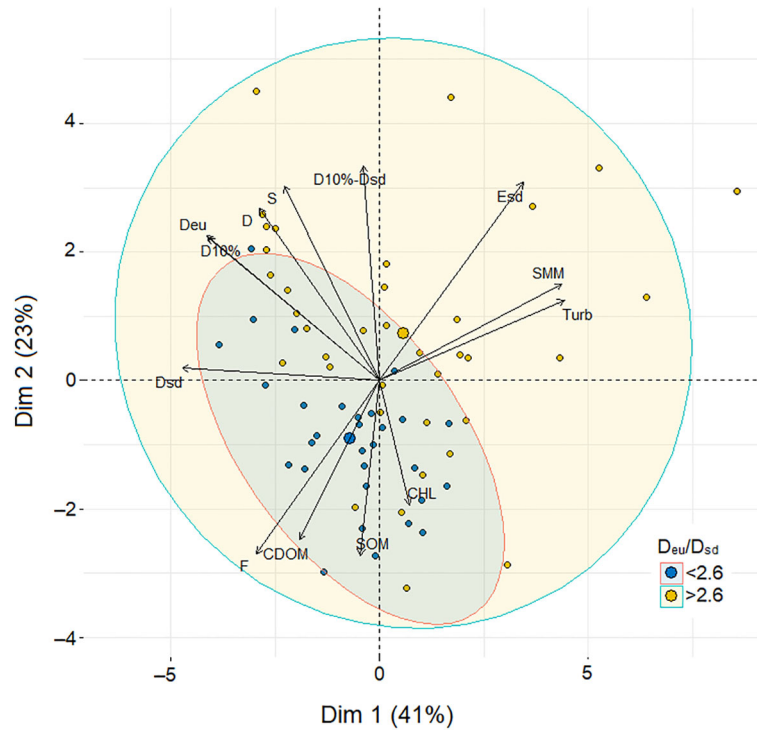


FIGURE 6

Results of principal component analysis of relationships between water transparency indices and environmental variables. All observation sites are ranked according to the D_{eu}/D_{sd} ratio. Yellow dots are observation sites where D_{eu}/D_{sd} was less than the median of the entire dataset (< 2.6). Blue dots are observation sites where D_{eu}/D_{sd} was greater than the median of the entire dataset (> 2.6). The contours show the individuals (data points) distribution for each data set in which the D_{eu}/D_{sd} ratio was greater than or less than 2.6. The large blue and yellow dots are the centers of the dispersion ellipses for each data set. The arrows show the vectors of changes in the indicators of the optical characteristics of water and environmental variables.

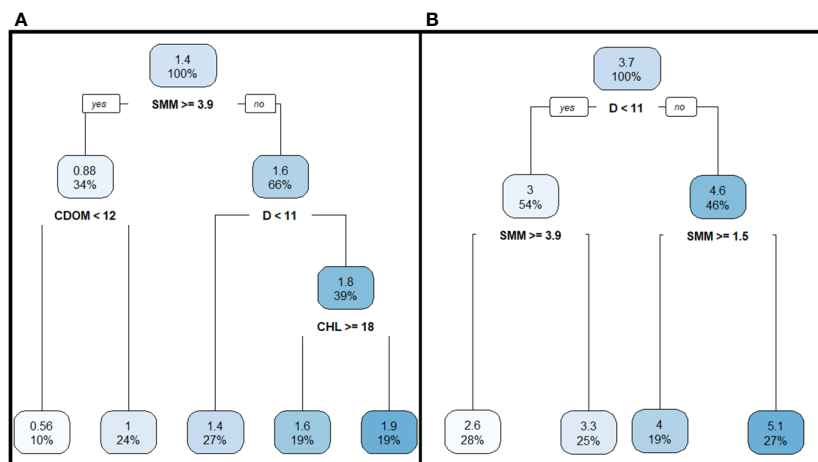


FIGURE 7

An inverted regression trees for Secchi disk depth (A) and depth of euphotic zone (B) obtained as a result of recursive partitioning of the data set based on the results of the analysis of variance. The numbers inside the block are the average value of the indicator (in meter) shown along with the percentage of cases of the total number of observations. At low values of the indicator, the color of the block is pale; the more saturated the color, the higher the values. The legend below the block is the abbreviated name of the main environmental factor in its units that determines the variance of the array. The numbers show the average values of the main environmental factors influencing D_{sd} and D_{eu} ; the signs $>$ and $<$ are symbols more and less. Yes and No are indicators of compliance or non-compliance with the necessary condition (the value of the influencing factor). The factors that determine the dispersion are arranged in a hierarchical pyramid. The factor at the top of the pyramid is the determining factor for the entire data set and a necessary condition when considering the factors below that determine the variance in certain parts of the entire data set.

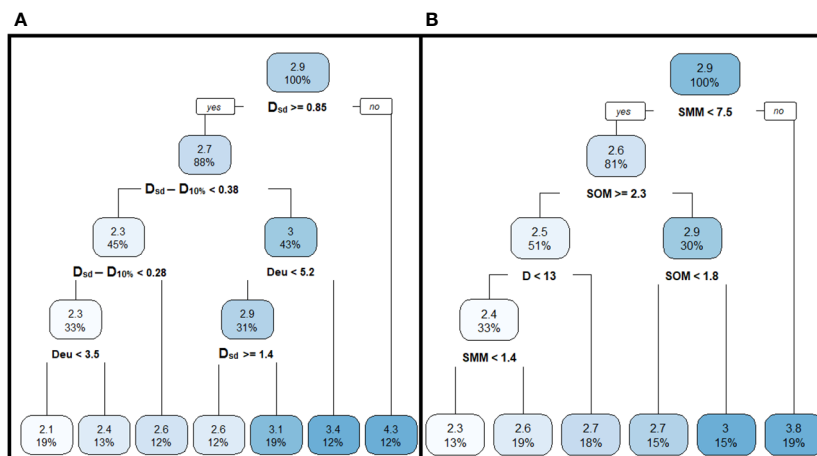


FIGURE 8
An inverted regression trees for the values of the D_{eu}/D_{sd} ratio depending on the optical characteristics of water (A) and environmental factors (B) as a result of recursive partitioning of the data array based on the results of the analysis of variance. The designations are the same as in Figure 7.

development solely based on D_{sd} in the Neva Estuary, as D_{sd} depends more on water turbidity than phytoplankton biomass (Figure 11B).

4 Discussion

Published data on the Secchi disk depth in the Baltic Sea reveal variations across different regions. The Gulf of Riga shows a lower limit of D_{sd} at 0.5 m, while the Sea of Bothnia reaches up to 15.3 m (Fleming-Lehtinen and Laamanen, 2012). Harvey et al. (2019) reported similar findings, with minimum D_{sd} values of 0.5 m near the Swedish coast of the Bothnian Sea and a maximum value of 12.8 m in the Baltic Proper during midsummer. Neumann et al. (2015) found that in the open waters of the Baltic Sea, D_{sd} ranged from 2.5 to 22 m along the transect from the Bay of Bothnia to the entrance

to the North Sea. In the shallow fjords of the Danish Straits, which connect the Baltic and North Seas, D_{sd} has fluctuated between 2 and 6 m over the past 30 years (Pedersen et al., 2014). Comparing our data with these published findings, it is evident that the upper and middle parts of the Neva Estuary are characterized lower D_{sd} values (Table 1) compared to the open waters of the Baltic Sea. However, in the Estonian coastal waters of the Gulf of Finland, D_{sd} periodically reached as low as 0.2 m, likely due to dredging activities in the area (Liblik and Lips, 2011). Similarly, in the Neva Estuary, hydrotechnical work was also periodically carried out, and the minimum values of D_{sd} (Table 1) were timed to coincide with these works, when a lot of suspended particles entered the water column, and negatively affected the productivity and species richness of phytoplankton in the estuary (Golubkov and Golubkov, 2022; Golubkov et al., 2023).

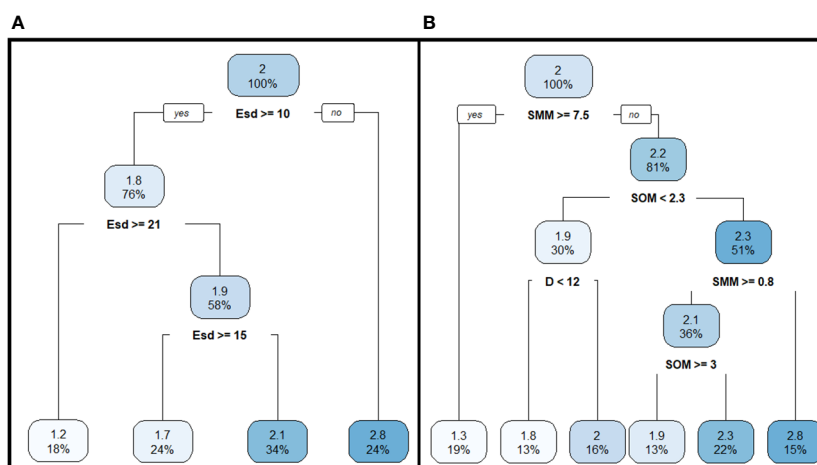


FIGURE 9
An inverted regression trees for the coefficient $F (m^{-1})$ depending on the optical characteristic of water (A) and environmental factors (B) obtained by recursive partitioning of the data array based on the results of the analysis of variance. The designations are the same as in Figure 7.

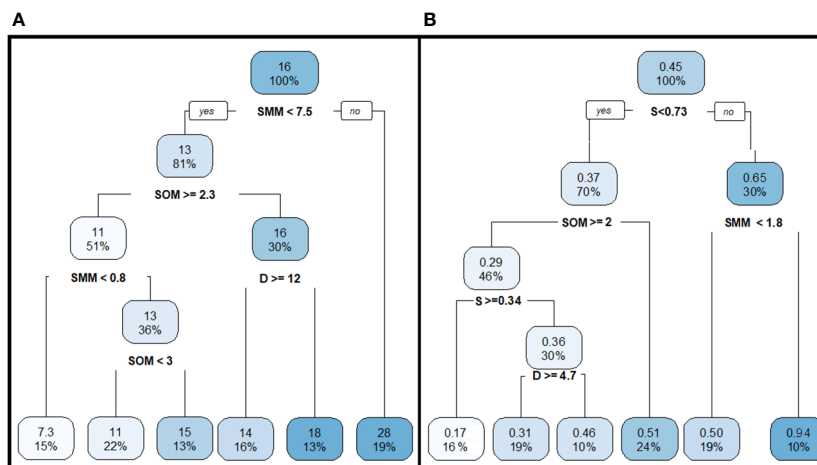


FIGURE 10
An inverted regression trees for E_{sd} (%) (A) and the difference between $D_{10\%}$ and D_{sd} (m) (B) depending on environmental factors, obtained as a result of recursive partitioning of the data array based on analysis of variance. The designations are the same as in Figure 7.

The theoretical understanding is that the Secchi disk disappears from view at a depth where 10% of the surface PAR remains (Preisendorfer, 1986; Kirk, 2011). This has been confirmed by studies in the Atlantic coastal waters at depths greater than 10 m (Lee et al., 2018). However, according to numerous empirical data, the Secchi disk becomes invisible to the human eye at depths where 5 to 40% of the surface PAR reaches, with an average of about 18% (Koenings and Edmundson, 1991). In the Neva Estuary, E_{sd} fluctuated even more widely, from 2.5 to 43.5%, with an average value of about 15% (Table 1), which exceeds the theoretical calculated 10% value. The largest difference between $D_{10\%}$ and D_{sd} was observed in the zones of hydrotechnical works, where suspended particles concentration was exceptionally high (Figure 2C, Golubkov and Golubkov, 2022; Golubkov et al., 2023). In these areas with SMM concentrations above 1.8 g m^{-3} , the Secchi disk became invisible almost a meter before reaching the $D_{10\%}$ (Figure 10B).

Contrary to these observations, Bowers et al. (2020) reported that at suspended matter concentrations greater than 20 g m^{-3} , the classical theory (Holmes, 1970; Preisendorfer, 1986) underestimates the depth of the Secchi disk because it suggests that diffuse sunlight falling on the disk is reflected back to the observer's eye along the most direct route, as a beam. However, actually, in turbid waters,

some of the light reflected by the disk returns to the eye as diffuse light and increases its brightness to the human eye. As a result, the Secchi disk is visible at greater depths (by a factor of up to 4) than predicted by this theory (Bowers et al., 2020). Our data indicate that under certain conditions, this effect may also occur in the Neva Estuary, in the case when the Secchi disk disappeared later than the $D_{10\%}$, when an average of 7.3% of the PAR remained at the D_{sd} from its value on the water surface (Figure 10A). In such cases, D_{eu}/D_{sd} ratio was 2.3 (Figure 8B), that is very closer to classical average 2.4 for all world waters (Lee et al., 2018). This was observed at relatively low SMM concentrations but close to average SOM concentrations (Figure 10A, Table 2). However, in most cases, the Secchi disk disappeared much earlier than reaching the $D_{10\%}$ (Table 1, Figures 2C, 10). This phenomenon is likely due to the high content of CDOM in the waters of the Neva Estuary (Table 2, Golubkov and Golubkov, 2023) compared to open ocean waters (Kratzer and Moore, 2018). According to Bowers et al. (2020), photons reflected from the Secchi disk can be absorbed strongly by the CDOM on their way back to the water surface. Under such conditions, the eye of the researcher does not distinguish the disk well and, as a result, $D_{10\%}$ greatly exceeds D_{sd} .

In the Chesapeake Bay (USA), the Secchi disk depth increased as the water salinity increased (Testa et al., 2019). A similar spatial

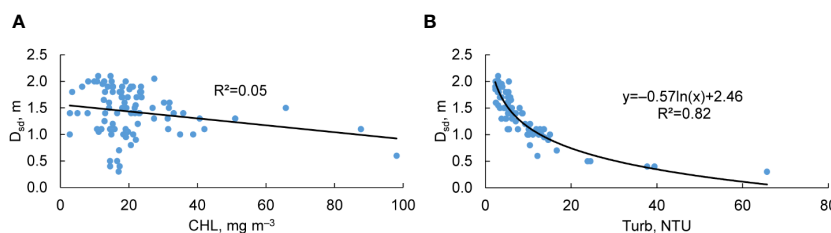


FIGURE 11
Relationship of Secchi disk depth (D_{sd}) with chlorophyll a concentration (CHL) (A) and total water turbidity (Turb) (B). R^2 – coefficient of determination.

distribution of D_{sd} and $D_{10\%}$ was also observed in the Neva Estuary (Figures 2A, B). However, the difference between these indicators sometimes approached one meter (Figure 10A), and there was no consistent confinement of this difference to specific parts of the part of the estuary (Figure 2C). Although, in general, this difference increased with distance from the mouth of the Neva River, increasing salinity and depth (Figures 2C, 6, 10B), a large difference between these indicators was also observed in some parts of the upper part of the estuary (Figure 2C). In the Neva Estuary, salinity did not affect D_{sd} (Figure 7A), but was a significant predictor of the difference between $D_{10\%}$ and D_{sd} . When the water salinity was above 0.73 ‰ this difference was greater, and when it was less, the difference was smaller (Figure 10A). However, this is explained not so much by the salinity of the water or the remoteness from the river mouth, but rather by the construction of port infrastructure that increase the SMM concentration in the middle part of the estuary, where the salinity was higher than in its upper part (Golubkov and Golubkov, 2022; Golubkov et al., 2023). This demonstrates that anthropogenic factors can also influence the difference between $D_{10\%}$ and D_{sd} .

Similar to other regions in the Baltic Sea (Neumann et al., 2015), our data show that prediction and reconstruction of $K_d(\text{PAR})$ and D_{eu} in the Neva Estuary based on D_{sd} data is possible. This is supported by the statistically significant power-law relationship between $K_d(\text{PAR})$ and D_{sd} (Figure 4A), as well as the linear relationship between D_{sd} and D_{eu} (Figure 4B). The significant correlation coefficients indicate that changes in $K_d(\text{PAR})$, D_{eu} , and D_{sd} occurred concurrently in the estuary, and these indicators changed synchronously by approximately 50%. This finding aligns with the long-standing use of D_{sd} to determine the penetration depth of PAR (Fleming-Lehtinen and Laamanen, 2012; Aas et al., 2014). However, the 50% uncertainty of the obtained regressions (Figures 4A, B) imposes certain restrictions on their use when accurate and relatively small-scale estimates are needed. Modeling the relationship between the Secchi disk depth and $K_d(\text{PAR})$ in various oceanic regions has shown inconsistent results, indicating that the relationship between these indicators is not straightforward and can be influenced by multiple factors (Castillo-Ramírez et al., 2021). Caution should be exercised when using D_{sd} to calculate $K_d(\text{PAR})$ due to the difficulty in obtaining stable and interpretable results. For large-scale comparisons, the spread of values, as observed in Fleming-Lehtinen and Laamanen (2012) for different zones of the Baltic Sea, may not be critical. Nevertheless, when examining a specific depth range, such as up to 20 meters, the scatter of points can be substantial, and a reliable relationship may not be clearly visible, as seen in the study by Nishijima et al. (2016) in the Seto Inland Sea, Japan.

In the open waters of the Baltic Sea and the Neva Estuary, the Secchi disk depth was related by a power law to $K_d(\text{PAR})$ with close coefficient values (Figure 4A, Neumann et al., 2015). The coefficient of proportionality in the equation for the Baltic waters was 1.61, against 1.68 in the Neva Estuary, and the exponent was -0.94 and -0.92 , respectively. However, the range of $K_d(\text{PAR})$ and D_{sd} values for which these equations were obtained differed significantly. According to Neumann et al., 2015, in the open waters of the Baltic Sea from the Bothnian Bay to the entrance of

the North Sea, $K_d(\text{PAR})$ varied from about 0.04 to 0.94 m^{-1} , and D_{sd} from 3 to 24 m. In the Neva Estuary $K_d(\text{PAR})$ ranged from 0.54 to 9.21 m^{-1} (Golubkov and Golubkov, 2023) and D_{sd} from 0.3 to 2 m (Table 1). The similarity of the coefficients of these equations to each other means that both dependencies are valid for both coastal and open waters of the Baltic Sea. The power equations of dependencies $K_d(\text{PAR}) - D_{sd}$ with completely different coefficients were obtained for lakes located in the catchment area of the Baltic Sea (Paavel et al., 2008; Ficek and Zapadka, 2010).

Although the equations obtained for the Baltic well describe the average dependences of $K_d(\text{PAR})$ on D_{sd} , the spread of the measured values of D_{sd} used to calculate the equations can reach 3-4 times (Figure 4A, Neumann et al., 2015). This variation arises because D_{sd} can be influenced by different environmental factors, and the same D_{sd} values in different regions of the Baltic Sea may not always correspond to the same environmental conditions (Harvey et al., 2019). Our results confirm this opinion. Principal component analysis, in which the entire data set was analyzed as a single vector, did not allow us to separate the data by the D_{eu}/D_{sd} ratio in the Neva Estuary depending on environmental variables (Figure 6). Modeling by building regression trees based on recursive partitioning of the data arrays made it possible to identify combinations of factors that affect the D_{eu}/D_{sd} ratio, but showed that combinations of different factors can lead to the same values of this ratio (Figure 8).

The same result was obtained for the PAR absorption coefficient F , used in the Equation 1 for calculating $K_d(\text{PAR})$ from D_{sd} measurements (Figure 10). Recursive partitioning of the data array showed that, on average, F was 2.8 for $E_{sd} < 10\%$, 2.1 for E_{sd} from 10 to 14%, 1.7 for E_{sd} between 15 to 20% and 1.2 for $E_{sd} \geq 21\%$ of the PAR on the water surface (Figure 9A). These results roughly align with theoretical calculations that associate 7.6% of the PAR on the surface with an F value of 2.57 and the D_{eu}/D_{sd} ratio of 1.79 (Preisendorfer, 1986) and are consistent with the finding of Poole and Atkins (1929), who reported an average F value of 1.7 for an average E_{sd} of 15%. In our study, the D_{eu}/D_{sd} ratio calculated according equation in Figure 5 was 2.2 if the F value was 2.8 m^{-1} and the E_{sd} value was less than 10% of the surface PAR (Figures 6, 9A). This was observed when the SMM concentration was less than 0.8 g m^{-3} , the SOM concentration was greater than 2.3 g m^{-3} , and the depth of estuary was less than 13 m (Figures 8B, 9A, 10A). No other environmental variables emerged as predictors of these indicators in the Neva Estuary.

Our average F value of 2.1 m^{-1} for the Neva Estuary (Table 1) coincides with the middle of its range from 1.7 to 2.3 m^{-1} reported for the Baltic Sea (HELCOM, 2002). However, the range of fluctuations in the F value in the Neva Estuary is wider than according to the HELCOM data for other areas of the Baltic Sea. The lowest F value of 0.8 m^{-1} is significantly lower (Table 1) than the proposed F value for turbid water estuaries (1.4 m^{-1}) (Holmes, 1970) and the average F value of 1.4 m^{-1} obtained from statistical analysis of data from 1419 Chinese lakes (Zhang et al., 2020). On the other hand, the upper limit of the F range in the Neva Estuary is 3.7 m^{-1} (Table 1), which is close to the maximum F values observed in lakes with colored water (Koenings and Edmundson, 1991). This shows that in estuaries the range of fluctuations in the optical characteristics of water and the

depth of penetration of light into the water column are higher than in open sea waters and lakes, which is apparently a common feature of estuaries as water bodies with a complex dynamic water structure (Emelyanov, 2005; Bianchi, 2007).

It is believed that F increases with decreasing salinity in the Baltic Sea, and based on this opinion, it is concluded that the value of the coefficient F is a function of continental CDOM runoff (Kratzer et al., 2003; Pierson et al., 2008). In the Neva Estuary, changes in the F coefficient were positively related with CDOM concentration (Figure 6), but the values could differ or overlap depending on the combination of mineral and organic suspended matter concentrations (Figure 9B), whose changes can be caused by both natural and anthropogenic factors (Golubkov and Golubkov, 2021, 2022). Therefore, the Secchi disk depth in the Neva Estuary was a function of many indicators affecting the water turbidity, but not just one, such as chlorophyll a concentration, which did not show a statistically significant relationship with D_{sd} (Figure 11A). This was quite expected, since in the rather shallow Neva Estuary, where bottom sediments are easily stirred up in the water column in windy weather, suspended mineral matter was the main predictor of both the Secchi disk depth (Figure 7) and the depth of the euphotic zone (Figure 7, Golubkov and Golubkov, 2023).

The values of the F coefficient and the D_{eu}/D_{sd} ratio in the Neva Estuary were closely related (Figure 5). The D_{eu}/D_{sd} ratio decreases at high F values (Figure 5) and SMM concentrations above 7.5 g m^{-3} (Figure 8B, 9B). At the same time, in the middle of the range of their values, F and D_{eu}/D_{sd} showed a more complex dependence on environmental factors and optical characteristics of water, since different combinations of environmental variables led to the same or intersecting values of F and D_{eu}/D_{sd} (Figure 8, 9). In this regard, it cannot be stated without alternative that with an increase, for example, in the concentration of SMM, the coefficient F will consistently decrease, and the D_{eu}/D_{sd} ratio will increase. The reason for this may be that estuaries are zones where continental waters mix with marine waters, and as a result, countercurrents are formed and changes in biogeochemical cycles are observed. They include changes in the behavior of both suspended and dissolved substances, sorption and desorption processes, flocculation and sedimentation (Emelyanov, 2005; Bianchi, 2007). All this leads to various combinations of numerous environmental factors.

The D_{eu}/D_{sd} ratio and the F coefficient correlate with the PAR remaining at the depth of the disappearance of the white Secchi disk from the observer's view (Figures 4C, D). In addition, although the Secchi disk depth and the euphotic zone depth are linearly related, there is a significant scatter of points from the trend line (Figure 4B). This suggests that the D_{eu}/D_{sd} ratio and F values are also depend on how the human eye perceives the contrast of the Secchi disk. Regression trees conducted separately for D_{sd} and D_{eu} indicate that the predictor sets for these indicators do not exactly match (Figure 7). Although for both indicators, the concentration of mineral suspended matter and the depth of the estuary were significant predictors (Figure 7, Golubkov and Golubkov, 2023), for Secchi disk depth, as in Testa et al., 2019, CDOM concentration and chlorophyll a concentration were also significant (Figure 7A). In other words, various admixtures in the water have a greater impact on the visibility of the white disk and, to a lesser extent, on

the actual depth of the euphotic zone as measured by PAR sensors. The disk disappears from the field of view when the contrast drops to a critical value, after which the eye of the researcher can no longer distinguish the disk from the background. CDOM and suspensions in water change the contrast and affect the perception of the disc by the human eye (Bowers et al., 2020). For example, studies conducted in a small eutrophic reservoir located on the northern coast of the Neva Estuary showed that at turbidity below 20 NTU, the Secchi disk clearly distinguished differences in water transparency, but when Turb exceeded 40 NTU, D_{sd} remained virtually unchanged with further increases in chlorophyll a (Golubkov and Golubkov, 2024).

Lee et al. (2018) suggest using a D_{eu}/D_{sd} ratio of 3.5 versus the classic 2.4. Our data show that this ratio is reasonable for calculations in the Neva Estuary, but only at high SMM concentrations (Figure 8B). This is the difference between our data and the data of Lee et al. (2018), where the D_{eu}/D_{sd} ratio was mainly influenced by the concentration of chlorophyll a . However, in lakes, the high D_{eu}/D_{sd} value of 3.5 is characteristic of turbid glacial lakes with high amounts of suspended mineral particles (Koenings and Edmundson, 1991). This also suggests that the same D_{eu}/D_{sd} ratios can be caused by different environmental factors, and it is not so easy to choose which ratio to take in estuaries for the transition from D_{sd} to D_{eu} (Figure 8), for example, to calculate the plankton primary production. A study of the lakes of Alaska showed that using the Secchi disk, one can reliably determine the depth of light penetration in large clear lakes with low color and suspended matter concentration (LaPerriere and Edmundson, 2000). However, in lakes with the most complex optical characteristics of water, the scatter of values in the dependence of D_{sd} on $K_d(\text{PAR})$ increases significantly. Similarly, predicting the depth of seagrass distribution in the turbid Roskilde Fjord Estuary using the Secchi disk or $K_d(\text{PAR})$ gave a difference in the $D_{10\%}$ of 0.5 to 1.5 m (Pedersen et al., 2014).

The Secchi disk depth is included in the list of key indicators for assessing eutrophication and nutrient pollution in the Baltic Sea, since chlorophyll a concentration is generally inversely proportional to the value of D_{sd} (Fleming-Lehtinen and Laamanen, 2012; HELCOM, 2018). However, in Fleming-Lehtinen and Laamanen (2012), at D_{sd} values less than 5 m, the scatter in the chlorophyll a concentration is very large. For example, at D_{sd} about 3 m, chlorophyll a concentration varied by an order of magnitude from 3 to 30 mg m^{-3} . Our data were obtained in the D_{sd} range from 0.3 to 2 m (Table 1) in which this dependence does not seem to work. In the UP and MP of the Neva Estuary, we did not obtain a reliable relationship between chlorophyll a and Secchi disk depth (Figure 11A).

In summary, given the complexity and variability of the factors that influence the relationship between the Secchi disk depth and the optical characteristics of water, this indicator should be used with caution for ecological studies and assessments of the degree of water eutrophication. Studies conducted in the Neva Estuary showed that to ensure the reliability of using the Secchi disk to determine the optical characteristics of water, more detailed knowledge is needed about the effect of the amount and combination of various impurities and forms of suspended matter on the Secchi disk depth. Obviously, in order to reliably interpret the data and determine from D_{sd} , for example, the depth of the euphotic zone, instead of linear regressions, it is

necessary to apply more complex models, using machine learning and neural connections, which are now beginning to be actively developed, e.g., Khanna et al. (2022), Zhang et al. (2022), Lin et al. (2022). In order to establish region-specific relationships between D_{sd} , $Kd(PAR)$ and D_{eu}/D_{sd} ratio, it would be beneficial to conduct benchmark determinations of PAR at the Secchi disk disappearance depth in different regions with specific combinations of environmental variables. This would contribute to improve the accuracy of eutrophication assessments and the reconstruction of historical phytoplankton productivity based on D_{sd} data.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary Material. Further inquiries can be directed to the corresponding author.

Author contributions

MG: Conceptualization, Data curation, Formal Analysis, Investigation, Methodology, Resources, Software, Supervision, Validation, Visualization, Writing – original draft. SG: Conceptualization, Data curation, Formal Analysis, Investigation, Methodology, Resources, Software, Supervision, Validation, Visualization, Writing – original draft.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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Supplementary material

The Supplementary Material for this article can be found online at: <https://www.frontiersin.org/articles/10.3389/fmars.2024.1265382/full#supplementary-material>

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