Simplified Method for Analyzing the Availability of Rooftop Photovoltaic Potential

Primož Mavsar 1, Klemen Sredenšek 2, Bojan Štumberger 2,3, Miralem Hadžiselimović 2,3 and Sebastijan Seme 2,3,*

1 Regional Surveying and Mapping Authority Novo mesto, Ljubljanska cesta 26, 8000 Novo mesto, Slovenia; primozmusa@hotmail.com
2 Faculty of Energy Technology, University of Maribor, Hočevarjev trg 1, 8270 Krško, Slovenia; klemen.sredensek@um.si (K.S.); bojan.stumberger@um.si (B.Š.); miralem.h@um.si (M.H.)
3 Faculty of Electrical Engineering and Computer Science, University of Maribor, Koroška cesta 46, 2000 Maribor, Slovenia
* Correspondence: sebastijan.seme@um.si

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Abstract: This paper presents a new simplified method for analyzing the availability of photovoltaic potential on roofs. Photovoltaic systems on roofs are widespread as they represent a sustainable and safe investment and, therefore, a means of energy self-sufficiency. With the growth of photovoltaic systems, it is also crucial to correctly evaluate their global efficiency. Thus, this paper presents a comparison between known methods for estimating the photovoltaic potential (as physical, geographic and technical contributions) on a roof and proposes a new simplified method, that takes into account the economic potential of a building that already has installed a photovoltaic system. The measured values of generated electricity of the photovoltaic system were compared with calculated photovoltaic potential. In general, the annual physical, geographic, technical and economic potentials were 1273.7, 1253.8, 14.2 MWh, and 279.1 Wh, respectively. The analysis of all four potentials is essential for further understanding of the sustainable and safe investment in photovoltaic systems.

Keywords: photovoltaic system; rooftop photovoltaic potential; economic potential; light detection and ranging (LiDAR)

1. Introduction

Population growth and, consequently, per capita consumption are increasing worldwide in developed countries. The solution to this problem is to integrate renewable energy in urban areas [1–5]. Renewable energy integration is suitable for all economies in urban areas, whether developed, emerging or underdeveloped [6]. To successfully continue the integration of the photovoltaic (PV) systems into the grid and to improve current policies and directives, it is essential to correctly estimate rooftop PV potential [7–16]. In [17], the advantages and disadvantages of various methods for estimating rooftop suitability for PV installations are described. The analysis was carried out with the constant-value method, manual selection method and Geographic Information System (GIS)-based method [18], which are (according to [17]) the three most important methods for estimating rooftop suitability for PV installations. The rooftop PV potential can be hierarchically divided into four steps: physical, geographic, technical [19,20], and economic potential [21–23]. All four steps are graphically presented in Figure 1.
A similar estimation of PV potential was performed by Hong et al. [24]. The first steps were performed using Hillshade analysis, while in [25], a new economic potential was added, taking into account spatial and temporal diversity. In places where buildings are of higher dimensions, the installation of PV systems may not always be suitable due to the shading of surrounding objects. Therefore, in order to correctly evaluate the geographic potential, it is imperative to take into account the shading caused by the surrounding objects. Accordingly, several studies analyzed the geographic potential of building rooftops using various methods to determine the shadow effect [26–30]. Lukač et al. [31,32], using graphics processing units (GPU), and light detection and ranging (LiDAR) data, showed how the determination of shadows can drastically reduce the estimated PV potential. Rodriguez et al. [4] presented a new methodology approach for determining the PV potential at an urban and regional scale using a description of CityGML geometry. The proposed method determines the PV potential for each building separately using building-by-building analysis, irradiance simulation, and reduction factors for energy yield. Furthermore, PV potential greatly depends on terrain configuration and on the inclination and orientation of the PV systems. Seme et al. [33] compared energy production between differently inclined and oriented PV systems under outdoor conditions. The basis for determining the geographic potential in this paper is the use of LiDAR data that show a 3D representation of the terrain surface, with high accuracy for large geometrically simple roofs [34].

The European Union (EU) directive on renewable energy sources (2009/28/ES) in self-sufficient energy communities at the local and national levels gives producers and consumers a new way of trading electricity in the retail market. Thus, the producer can sell electricity to consumers without intermediaries. In the near future, each producer will have the opportunity to sell electricity directly to the electricity market. Therefore, it makes sense to introduce the economic potential that describes the economic aspect of eligibility of the PV system. The economic potential takes into account the technical generation of electricity, the hourly price of electricity on the market, the investment cost, and the installed power of the PV system. Maciejowska et al. [35], presents forecasts of the price spread between the intraday/balancing and day-ahead markets in the decision process. Therefore, economic potential shows that the economic and technical potential are not proportional, but it all depends on the price of electricity at a given moment. It is shown that the price of electricity has significantly higher meanings than the projected technical potential. The contribution of this research is the development of a combined procedure for determining photovoltaic potential on rooftops that are suitable for PV installation. Figure 2 shows the location of the Krško-Sevnica School Center, on which the estimation of the rooftop PV potential was performed.
2. Methods

The methodology for estimating the rooftop PV potential is divided into four parts (see Figure 3): (a) calculation of the physical potential; (b) calculation of the geographic potential; (c) calculation of the technical potential; and (d) calculation of the economic potential.

2.1. Physical Potential

The physical potential $PP$ (Wh) is basically the amount of energy that the total area receives over a certain period of time, and can be calculated by Equation (1):

$$PP = A_T \cdot \int_{\text{sunrise}}^{\text{sunset}} G_h(t) \cdot dt$$  

Figure 2. Location of the photovoltaic system Krško-Sevnica School Center, Slovenia.

Figure 3. Visualization of the method for analyzing the availability of rooftop PV potential: (a) Physical potential; (b) Geographic potential; (c) Technical potential; (d) Economic potential.
where $A_T \text{ (m}^2\text{)}$ stands for total rooftop area of the selected location and $G_h(t) \text{ (W/m}^2\text{)}$ stands for the global solar radiation on a horizontal surface and can be calculated by Equation (2):

$$G_h(t) = G_b(t) + G_d(t) \tag{2}$$

where $G_b(t) \text{ (W/m}^2\text{)}$ stands for the beam solar radiation and $G_d(t) \text{ (W/m}^2\text{)}$ stands for the diffuse solar radiation on a horizontal surface. Solar radiation data for global and diffuse components are provided in a half-hourly intervals basis by the Slovenian Environment Agency [36].

2.2. Geographic Potential

The geographic potential $GP \text{ (Wh)}$, is basically the amount of energy that the available area receives over a certain period of time, and can be calculated by Equation (3):

$$GP = A_{eff} \cdot \int_{sunrise}^{sunset} G_t(t) \cdot dt \tag{3}$$

where $A_{eff}$ stands for effective rooftop area of the selected location, and $G_t(t) \text{ (W/m}^2\text{)}$ stands for the total solar radiation on an inclined surface and can be calculated by Equation (4) [37]:

$$G_t(t) = \left(G_b(t) \cdot \frac{\cos i}{\sin \alpha}\right) + G_d(t) \cdot \left(\frac{1 + \cos \beta}{2}\right) + G_h(t) \cdot \left(\frac{\rho \cdot (1 - \cos \beta)}{2}\right) \tag{4}$$

where $i \text{ (°)}$ stands for the incidence angle of the sun, $\alpha \text{ (°)}$ stands for solar altitude angle, $\beta \text{ (°)}$ stands for inclination angle of the surface above the horizontal, and $\rho$ stands for a surface albedo. To calculate the above-presented parameters, the following parameters must also be considered: declination angle $\delta \text{ (°)}$, solar hour angle $\omega \text{ (°)}$, zenith angle $z \text{ (°)}$, and solar azimuth angle $\gamma$.

Accurate prediction of the geographic potential must also consider the shadowing effect of terrain and surrounding objects, since shadow areas do not receive a direct component of the solar radiation. To determine the shadows of a specific area, the characteristics of individual elements in the area must be described (height, slope, dimensions) and projected into the environment, depending on the current position of the Sun. The determination of the area of shadows at a given moment was carried out from LiDAR data in the following steps: (1) calculation of the Sun’s coordinates at a given moment; (2) determination of cells on the beam path–line rasterization; and (3) determination of the shadow factor.

2.2.1. Calculation of the Sun’s Coordinates

Coordinates of the points in LiDAR data are given in a rectangular coordinate system. To determine the position of the Sun, the spherical coordinates of the Sun must be transformed into Cartesian coordinates. After the transformation, a topo-centric coordinate system is defined as an orthogonal reference system with the origin at the observed position on the surface of the Earth (see Figure 4). The coordinate axes are defined as follows:

1. The X-axis is tangential to the Earth’s surface in the East-West direction and positive eastwards.
2. The Y-axis is tangential in the North-South direction and positive southwards.
3. The Z-axis lies along the Earth’s radius and is positive upwards.

The vector in the direction of the Sun can be calculated using spherical geometry (solar azimuth and zenith angles). When the Sun lies on the ZY-plane (vertical plane), the coordinates of the Sun can be written as Equation (5):

$$s_0 = (0, \sin(\varphi - \delta), \cos(\varphi - \delta)) \tag{5}$$
where $s_0$ stands for vector, $\varphi$ stands for the latitude, and $\delta$ stands for the declination angle. In general, the solar vector $s$ in the rectangular coordinate system at any time is determined by Equation (6):

$$s = r_x(\varphi)r_z(\omega)r_x(-\gamma) \cdot s_0$$

where $r$ stands for rotational matrix around the axis in subscript and angle in brackets, $\gamma$ (°) stands for the angle between Earth's axis and the topo-centric coordinate system Z-axis, and $\omega$ (°) stands for the hour angles. The final equation of the solar vector $s$ for every coordinate $(x, y, z)$ is defined by Equation (7):

$$s = \begin{pmatrix}
-\sin \omega \cos \delta \\
\sin \varphi \cos \omega \cos \delta - \cos \varphi \sin \delta \\
\cos \varphi \cos \omega \cos \delta + \sin \varphi \sin \delta
\end{pmatrix}$$

Figure 4. Rotation of the topo-centric coordinate system XYZ at angle $\omega$, from time $t_0$ to time $t_1$ [37].

2.2.2. Determination of Cells on the Beam Path–Line Rasterization

Based on solar vector $s$ and known algorithm [38], the cells on the beam path can be determined. The algorithm works according to the principle of line rasterization. Figure 5 presents the determination of cells on the beam path.
2.2.3. Determination of Shadow Factor

The shadow factor can be determined on the basis of two known methods. The first approach (or vector method) presented in [38] determines the projection of individual cells on the plane $S_p$, which is perpendicular to the sun’s rays. The vector from the origin to any point $P_{ij}$ at column $i$, row $j$, can be calculated by Equation (8):

$$ P_{ij} = \left( l_i, l_j, z_{ij} \right) $$

(8)

where $l_i$ and $l_j$ stand for the cell sizes, and $z_{ij}$ stands for the elevation of a cell at $i, j$. The plane, perpendicular to the Sun’s rays, is calculated as a combination of two cross products in Equation (9):

$$ s_p = s \times s \times s_{xy0} $$

$$ \left| \frac{s \times s_{xy0}}{s} \right| $$

(9)

where $s$ stands for solar vector and $s_{xy0}$ stands for (XY) plane. The projection $P'_{ij}$ of $P_{ij}$ is calculated by Equation (10):

$$ P'_{ij} = P_{ij} \cdot s $$

(10)

The second approach (or geometrical method) presented in [38] determines whether the examined cell casts a shadow on the surrounding cells if the Sun’s rays are projected onto the model. The unit change of height $\mu$ for the solar vector $s = [s_x, s_y, s_z]$ is defined by Equation (11):

$$ \mu = \frac{s_y}{\sqrt{s_x^2 + s_z^2}} $$

(11)

Using the change of height $\mu$, when the solar vector $s$ passes through the horizontal distance $D$ between two given cells, $a = [a_x, a_y, a_z]$ and $b = [b_x, b_y, b_z]$, the height $h$ can be determined by Equation (12):

$$ h = D \cdot \mu = \left( \sqrt{(a_x - b_x)^2 + (a_z - b_z)^2} \right) \cdot \mu $$

(12)

The minimum height required to keep the cell out of the shade is $h_{\text{min}} = a_y - h$. Otherwise, if the $b_y < h_{\text{min}}$, then cell $b$ is shadowed by cell $a$ [38]. The algorithm scans the entire model by the method described in Section 2.2.2, and determines for each cell a shadow factor $s_i$, which is 0 for shaded cells and 1 for illuminated cells.

2.2.4. Determination of Geographic Potential Based on LiDAR Data

LiDAR data issued by the Slovenian Environment Agency [36] are in the ASCII format of cloud points located in the correct square network, where each point is defined by a coordinate in the 3-dimensional Cartesian coordinate system. To generate a surface model from LiDAR data, the plane enclosed by four data points $z_{ij}, z_{i+1,j}, z_{ij+1}, z_{i+1,j+1}$ must be determined, where $z_{ij}$ is the elevation of points at row $i$, column $j$.

To calculate the geographic potential according to Equation (4), the algorithm must calculate the vector normal to the grid cell $n$, inclination $\beta$ and azimuth angle $\gamma$ of the cell. In a regular square grid of cell size $l$, the $x, y, z$ components of the vectors along the side of the grid cell are defined by Equation (13) [36] and shown in Figure 6:

$$
\begin{align*}
a &= (l, 0, \Delta z_a) \quad \text{with} \quad \Delta z_a &= z_{i+1,j} - z_{ij} \\
b &= (0, l, \Delta z_b) \quad \text{with} \quad \Delta z_b &= z_{i,j+1} - z_{ij+1} \\
c &= (-l, 0, \Delta z_c) \quad \text{with} \quad \Delta z_c &= z_{i,j+1} + z_{i+1,j+1} \\
d &= (0, -l, \Delta z_d) \quad \text{with} \quad \Delta z_d &= z_{i+1,j} - z_{i+1,j+1}
\end{align*}
$$

(13)
The vector normal to the surface is defined by half the sum of cross products of vectors along the sides of the grid cell in Equation (14):

\[
n = \frac{(a \times b)}{2} + \frac{(c \times d)}{2} = \frac{1}{2} \begin{vmatrix} i & j & k \\ l & 0 & \Delta z_a \\ 0 & l & \Delta z_b \end{vmatrix} + \frac{1}{2} \begin{vmatrix} i & j & k \\ -l & 0 & \Delta z_c \\ 0 & -l & \Delta z_d \end{vmatrix}
\]  

By simplifying Equation (14), the components of the vector normal to the surface \(n\) can be written with the grid elevation points \(z\) and cell spacing \(l\) as Equation (15):

\[
n = \begin{pmatrix} \frac{1}{2} (z_{i,j} - z_{i+1,j} + z_{i,j+1} - z_{i,j+1} + z_{i+1,j}) \\ \frac{1}{2} (z_{i,j} + z_{i+1,j} - z_{i,j+1} - z_{i,j+1}) \end{pmatrix} \quad \beta
\]  

The inclination angle \(\beta\) and azimuth angle \(\gamma\) of the surface (cell) can be calculated from the component of the unit vector \(n_u\), as follows in Equations (16) and (17):

\[
\beta = \cos^{-1} n_{uz}
\]

\[
\gamma = \frac{\pi}{2} + \tan^{-1} \left( \frac{n_{uy}}{n_{ux}} \right)
\]

To determine the geographic potential that considers the shadowing effects of surroundings, the component of the solar radiation beam \(G_b(t)\) (described in Equation (4)) is multiplied by the shadow factor \(s_f\) (described in Section 2.2.2), which determines whether or not the cells in the model are shaded. Therefore, the geographic potential can be calculated by Equation (18):

\[
G(t) = (G_b(t) \cdot \frac{\cos i}{\sin \alpha}) \cdot s_f + G_d(t) \cdot \left( \frac{1 + \cos \beta}{2} \right) + G_b(t) \cdot \left( \frac{\rho \cdot (1 - \cos \beta)}{2} \right)
\]  

Figure 7 shows the example of geographic potential on a real surface model, with and without shadowing effects of surroundings for October 2018. Shadowing is most visible in the area with a dense distribution of objects (see left part of the Figure 7b).
2.3. Technical Potential

The technical potential (TP) (Wh) presents the amount of energy that an available area with PV modules receives over a certain period of time, including the efficiency of the PV system. Technical potential can be calculated by Equation (19):

\[ TP = \eta_{PV} \cdot GP \]  

(19)

where \( \eta_{PV} \) stands for the efficiency of the PV module. To optimize the technical potential, the efficiency of the PV module can be calculated as a function of solar radiation \( G \) and temperature of the PV module \( T \) in three different ways, as shown in Equations (20)–(22) [39].

\[ \eta_{PV}(G) = x_1 \left[ x_2 \frac{G}{G_0} + \left( \frac{G}{G_0} \right)^x_3 \right] \left[ 1 + x_4 + x_5 + 1 \right] \]  

(20)

\[ \eta_{PV}(T) = x_1 \left[ x_2 + 1 \right] \left[ 1 + x_4 \frac{T}{T_0} + x_5 + 1 \right] \]  

(21)

\[ \eta_{PV}(G, T) = x_1 \left[ x_2 \frac{G}{G_0} + \left( \frac{G}{G_0} \right)^x_3 \right] \left[ 1 + x_4 \frac{T}{T_0} + x_5 + 1 \right] \]  

(22)

where \( G_0 \) (1000 W/m\(^2\)) and \( T_0 \) (25 °C) stands for the solar radiation and temperature of the PV module under standard test conditions (STC). The parameters \( x_1, x_2, x_3, x_4, \) and \( x_5 \) are determined by minimizing the root mean square error (RMSE) differences between measured and simulated values of the efficiency of the PV module.

2.4. Economic Potential

Economic potential \( EP \) (Wh) presents the amount of energy, taking into account technical potential, the price of electricity, the investment price, and the installed power of the PV system. The economic potential can be calculated by Equation (23):

\[ EP = \frac{PR_{el} \cdot TP^2}{PR_{PV} \cdot P_{MPP}} \]  

(23)

where \( PR_{el} \) stands for the price of electricity (€/Wh), \( PR_{PV} \) stands for the price of the PV system (€/W), \( TP \) stands for technical potential (Wh), and \( P_{MPP} \) stands for the installed power of the PV system. Other studies that also consider economic potential are extremely complicated and difficult to calculate,
considering many different parameters that are often difficult to access, such as: building information (building characteristics, energy consumption, physical characteristics with spatial information), system size (maximum installation system size, eligible system size for policy implementation and 100% self-sufficiency), business model (self-consumption, power purchase agreement), etc. The new approach described in the paper offers a simplified model that still delivers accurate results and yet, makes it easier to calculate with easily accessible parameters such as: price of electricity, price of the PV system, installed power of the PV system and technical potential.

3. Results

In this section, the calculation of all potentials will be made on the rooftop of the Krško-Sevnica School Center ($\lambda = 15^\circ 29' 11'', \varphi = 45^\circ 56' 42''$), which is divided into 8 planes (see Figure 8). The required parameters for the calculation of different potentials are the total area of rooftop $A_T$, the inclination angle $\beta$ and the azimuth angle $\gamma$, which are described for each plane in Table 1.

![Figure 8. Visualization of the rooftop divided into 8 planes (colors in image present relative altitude).](image)

| Parameters | Plane 1 | Plane 2 | Plane 3 | Plane 4 | Plane 5 | Plane 6 | Plane 7 | Plane 8 |
|------------|---------|---------|---------|---------|---------|---------|---------|---------|
| $A_T$ [m$^2$] | 224 | 91 | 215 | 80 | 177 | 58 | 51 | 112 |
| $\beta$ [°] | 15 | 15 | 15 | 15 | 15 | 15 | 15 | 15 |
| $\gamma$ [°] | 0 | 90 | $-180$ | $-90$ | 90 | $-180$ | 0 | $-90$ |

3.1. Physical and Geographic Potential

The physical and geographic potentials were calculated for the period of one year, and Figure 9 shows the physical and geographic potentials (or hourly solar radiation on a horizontal surface) for an average day of the month. In Figure 9, the physical and geographic potentials for each month are presented in colored lines, according to the seasons, as (a) spring (green), (b) summer (orange), (c) autumn (yellow), and (d) winter (blue). First, the monthly values of physical and geographic potentials were the highest from 12 to 1 p.m., ranging from 185.5 to 826.6 kWh (for all months). Meanwhile, the monthly values of physical and geographic potentials were the lowest from 5 to 6 a.m., ranging from 6.552 to 73.584 kWh (from April to August). Second, in terms of seasons, physical and geographic potentials were the highest in July, May, September and February with monthly values of 417.9, 281.1, 220.9 and 104.9 MWh. The results from Figure 8 show that the deviation between physical and geographic potential is 1.56%, with annual physical potential of 1273.7 MWh and annual
geographic potential of 1253.8 MWh. Figures 10 and 11 show the geographic potential in graphical form for each month of the year, which also takes into account the shading effect. Figures 10 and 11 shows which rooftops receive the highest amount of solar radiation and thus represent the optimal location for the installation of the PV system.

**Figure 9.** Results of the physical and geographic potential for the rooftops of the Krško-Sevnica School Center.

![Results of the physical and geographic potential for the rooftops of the Krško-Sevnica School Center.](image)

**Figure 10.** Visualization of the geographic potential for January, February, March, April, May and June.

![Visualization of the geographic potential for January, February, March, April, May and June.](image)
Figure 10. Visualization of the geographic potential for January, February, March, April, May and June.

Figure 11. Visualization of the geographic potential for July, August, September, October, November and December.

3.2. Technical Potential

The technical potential was calculated for the rooftop (Plane 1) for a period of one year (see Figure 8), on which a PV system with an installed power ($P_{MPP}$) of 12,600 Wp is located. Measurement data was obtained from the owner of the PV system and included electrical quantities on the DC and AC sides of the inverter, air temperature, and solar radiation from a pyranometer mounted on the same inclination angle as the PV system modules [33]. To calculate the technical potential, the efficiency of the PV system was taken into account as a function of solar radiation $G$ and temperature of the PV module $T$. The RMSE values of efficiency of the PV system (Table 2) show that the smallest RMSE is achieved by considering both solar radiation $G$ and temperature of the PV module $T$ in the calculation.

### Table 2. RMSE values of efficiency of the PV system.

| Equation       | Parameters | $x_1$ | $x_2$  | $x_3$  | $x_4$  | $x_5$  | RMSE   |
|----------------|------------|-------|--------|--------|--------|--------|--------|
| Equation (20)  | $\eta(G)$  | 0.0568| -0.4678| 0.2368 | -2.7297| 4.1496 | 2.0727 |
| Equation (21)  | $\eta(T)$  | -0.0732| -1.7987| /      | -0.0925| 0.1063 | 3.0412 |
| Equation (22)  | $\eta(G,T)$| 0.1235| -0.4544| 0.2420 | -0.0309| -0.3854| 2.0217 |

Therefore, Figure 12 presents the comparison between measured and modelled values of efficiency as a function of (a) solar radiation $G$, (b) temperature of PV module $T$, and (c) solar radiation $G$ and temperature of PV module $T$. It can be seen that the efficiency of the PV system yields more accurate results if both solar radiation $G$ and the temperature of PV module $T$ are used in the calculation, as can
be seen in Table 2. In Figure 13, the values of technical potential or generated electricity are presented for an average day of the month. First, the monthly values of technical potential were the highest from 12 to 1 p.m., ranging from 98.371 to 253.607 kWh (for all months). Meanwhile, the monthly values of technical potential were the lowest from 5 to 6 a.m., ranging from 0.0175 to 11.635 kWh (from April to August). Similar to the physical potential, the technical potential (in terms of seasons) was the highest in July, May, September and February with monthly values of 2197.8, 1511.1, 1280.5 and 588.3 kWh.

![Figure 12](image1.png)  
![Figure 12](image2.png)  
![Figure 12](image3.png)

**Figure 12.** Comparison between measured and modelled values of efficiency as a function of (a) solar radiation $G$, (b) temperature of PV module $T$, and (c) solar radiation $G$ and temperature of PV module $T$.

![Figure 13](image4.png)

**Figure 13.** Results of the technical potential for the rooftops of the Krško-Sevnica School Center.

Figure 14 shows the comparison between technical potential and measurements of the PV system with an installed power of 12.6 kW. The results show that the deviation between measured values of generated electricity and technical potential is 3.45%, with annual measured generated electricity of 13.7 MWh and annual technical potential of 14.2 MWh.
3.3. Economic Potential

The economic potential was calculated based on the technical potential of the rooftop (Plane 1) and the variable price of electricity. Figure 15 presents the variable price of electricity throughout the day and year, showing that the price of electricity varies throughout every day of the year. Therefore, it is important to highlight in which part of the day electricity is the cheapest and most expensive. The lowest price for electricity is mainly at night (between midnight and 6 a.m.) and during working hours (between 11 a.m. and 4 p.m.), while the most expensive is in the morning (between 7 a.m. and 10 a.m.) and evening (between 5 p.m. and 10 p.m.). Furthermore, the price of electricity is the cheapest in January, February and April, while the most expensive is in September, October and November.

First, the monthly values of economic potential were the highest from 12 to 1 p.m., ranging from 1.22 to 8.41 Wh (for all months). However, in terms of seasons, the economic potential was slightly different from the technical potential, due to the impact of the price of electricity. In terms of physical, geographical and technical potential, the highest values first appeared in summer, spring, autumn and winter, while in economic potential, the highest values first appeared in summer (higher prices of...
electricity), autumn, spring and then winter. The economic potential was the highest in August, May, September and February with monthly values of 60.3, 29.87, 22.79 and 5.72 Wh.

The Slovenian electricity market is located at the intersection of three major European markets, the German-Austrian, Italian and Southeast European markets. There are several reasons for raising the average prices of base-load and peak energy in the market for the day ahead. Most important is the poor hydrology of the rivers and therefore, the relatively low production of electricity from hydropower plants. Economic growth and the growth of industrial production in EU countries also contributed to the rise in prices, which increased energy demand and is a consequence of the change in electricity prices shown in Figure 15. The price of the PV system \( P_{RPV} \) was determined as the average value of all costs of the PV system in Slovenia (installation costs), which included: solar modules, installation, inverters, and documentation. Figure 16 shows the reflection of technical potential if specific economic parameters are taken into account. It can be seen that the most suitable months for the generation of electricity, regardless of the technical parameters, are the summer months. Indeed, during the summer months, the most substantial yield of solar radiation occurs, but this is also influenced by the medium-high cost of electricity.

![Figure 16. Results of the economic potential for the rooftops of the Krško-Sevnica School Center.](image)

3.4. Comparison of Physical, Geographic, Technical, and Economic Potentials

In this subsection, the comparison between physical, geographic, technical, and economic potentials is presented. Figure 16 shows the average monthly potentials of the rooftop in normalized values ranging from 0 to 1. As shown in Figures 9, 13 and 16, and especially in Figure 17, the monthly physical, geographic, and technical potentials had the increase of potential in spring and summer and decrease in fall and winter. It is evident that the most important role in determining the economic potential is played by the price of electricity, since in the summer and spring months the economic potential ranges between 0.7 and 1, and in the other months between 0 and 0.5. Overall, the annual physical, geographic, technical and economic potentials are 1273.7, 1253.8, 14.2 MWh and 279.1 Wh, respectively. As can be seen from Figure 17, the geographic and physical potentials are almost identical. In other studies, there is a major discrepancy between these two potentials. In our case, similarities occur because of the smaller surrounding objects and thus, the smaller influence of the shadow on the observed area. In other studies, [24], the difference between them is much larger, since the calculation is made on a much larger area, where the layout of the objects is much more diverse. The most important part of this study was the calculation of economic potential, as it will play a massive role in the near future.
Figure 17. Results of the comparison between physical, geographic, technical and economic potentials for the rooftops of the Krško-Sevnica School Center.

4. Conclusions

This paper presents a simplified approach for estimating the rooftop solar potential. Apart from other studies presented in this paper, the estimation of rooftop solar potential is divided into the following four parts: (a) physical potential, (b) geographic potential, (c) technical potential, and (d) economic potential. In summary, based on the annual values of physical, geographic and technical potential, an analysis of the proportion by season was performed, which accounts for about 28%, 45%, 17% and 10% for spring, summer, fall and winter. The effects of different economic indicators and the dynamics of the stock market have a slightly different effect on the economic potential, since the share for the fall is approximately the same, while the share for the summer is 14% higher and 5% lower for the winter. The results presented in the paper show that the months from May to September are the most suitable months for Slovenia in terms of economic potential, which represents as much as 78% (3/4) of the annual economic potential, while the technical potential (and other potentials) have a more evenly distributed potential throughout the year. The message of this paper is that in the near future it will be necessary to take greater emphasis on the economic potential, than on the technical potential, when installing the PV systems, or in other words, less important will be the cumulative amount of annually produced electricity, and more significant will be the financial outflow. This may have a strategic impact on decision-making policies for the deployment of the PV systems in urban environments.

The new simplified approach differs from other studies by a simple but accurate calculation, since other studies are more complex and take into account more difficult-to-access information that is impossible to obtain for a larger urban area. The calculation takes into account the essential characteristics of the PV system and the dynamics of the electricity market, while other studies, in addition, place greater emphasis on the energy consumption by the consumer of the building on which the PV system is installed.

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