

Energy Income Estimation for Solar Cell Powered Wireless Sensor Nodes [†]

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Abstract: Solar cells are one common choice to power energy-autonomous wireless sensor nodes (WSNs). There are different approaches to improve their service and reliability via solar energy prediction algorithms, to allow the WSN to “know” about the future energy income. All these algorithms require information on the energy income of the sensor node obtained e.g., via separate measurements of light intensity or via monitoring the current flow to the WSNs energy storage. Here, we present a method to determine the energy income via temporarily switching the PV cell’s input current to a capacitor for determining the energy income by the accumulated charge. The data are compared to measurements with a pyranometer. This system provides advantages with respect to the consideration of the multiple different losses in the WSNs power management.

Keywords: wireless sensor node; energy management; energy harvesting; solar cell; energy income estimation

1. Introduction

Wireless sensor nodes (WSNs) are miniaturized electronic systems commonly used to measure, collect and transmit environmental data. Usually they operate in remote areas with difficult access, i.e., energy has to be provided by means of a battery, by energy harvesting or by a combination of both. To power a WSN via energy harvesting, photovoltaic (PV) cells with specialized DC/DC converters are commonly used in applications that give access to solar or artificial radiation, with the basic setup of Figure 1a. This concept is also used within this study, however, it should be mentioned that all principles described here can also be applied for WSNs powered from thermal or mechanical energy harvesting. As the gathered data have to be sent to a receiver via a wireless connection, with a significant power drain from the WSNs limited energy budget, severe problems will occur, when the WSN is “unexpectedly” running out of energy. Therefore, intelligent control algorithms are used in a WSN, to adjust internal energy consumption and usage to the energy income to achieve, overall, an energy-neutral operation [1]. These algorithms can be based on pure averaging of the energy income or, in the next step, on the prediction of energy income. To achieve good prediction results, the knowledge of local energy income of a WSN is indispensable.

There are different possibilities to measure energy income from solar radiation. As one example a photodiode with a spectral response similar to the PV cell in use can be used [2]. Furthermore, the current flow to a storage unit can be measured and the energy income can be calculated from the gathered data. The disadvantage of using a separate sensor for energy income is the negligence of the electrical losses in the system. The second option, i.e., measuring the input current [3], also causes some problems. First, to measure the current a serial resistor has to be included, which is generating losses. Second, if the storage is fully charged, the input current decreases, especially when energy is

available. Thus, information about the currently available energy is lost for prediction purpose or even can falsify prediction.

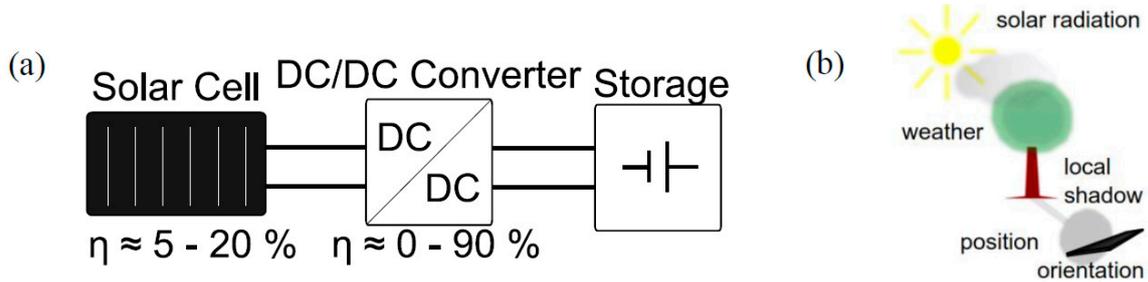


Figure 1. Influences on the harvestable energy of a sensor node: (a) Basic set up of a sensor node with a solar cell, DC-DC-Converter and an energy storage and their influence to the energy income. (b) Outdoor wireless sensor node and some general influences on the energy income.

Furthermore, there are lots of factors effecting the power income of a sensor node, as shown in Figure 1b for photovoltaic energy harvesting. The effect of weather, local shadows of buildings or trees and the orientation of the PV cell are important and variable from application to application.

In this study, a third option for energy measurement is presented, where the transformed energy from a DC/DC converter is stored on an extra capacitor and the voltage difference is used to determine the actual income of solar energy via a charge measurement. Comparison with standard radiation profile delivers the basis for a self-learning prediction algorithm with low computing effort and reasonable accuracy in different application scenarios, as described in the following chapters.

2. Measurement Set Up

Within the study two PV cells in parallel configuration are used. The area of a single cell is $39 \times 35 \text{ mm} \pm 0.2 \text{ mm}$, its open cell voltage under AM 1.5 conditions is $V_{OC} \approx 4.2 \text{ V}$ and its short circuit current $I_{SC} \approx 37 \text{ mA}$ [4]. A specially tailored low-voltage DC-DC-converter [5] with up to 70% power conversion efficiency is connected in series to the PV panel. The converter’s output voltage is fixed to 3.3 V, in order not to exceed the input voltage range voltage of the microcontroller’s AD converter. In this study, an Arduino Due is used with an internal 12-bit AD converter. Using two MOSFET switches, the output of the DC-DC converter is switched to a 1000 μF or 330 μF capacitor in regular intervals and for a defined measurement interval (Figure 2).

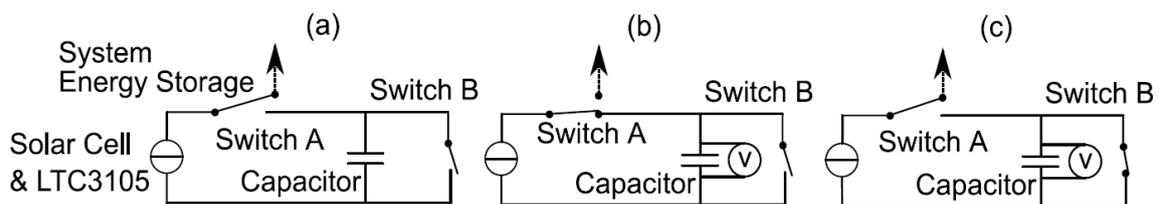


Figure 2. Schematic of the measurement setup with two switches and the capacitor powered by two solar cells connected to a LTC3105, (a) empty capacitor, (b) charging the capacitor and measure voltage (c) empty the capacitor again before going to state (a).

For switch A, a p-channel MOSFET (IRLML6402), whereas a n-channel MOSFET (IRLML250) is connected as switch B together with a 10 Ω discharge resistor. Both MOSFETs are operated from the Arduino Due.

As soon as switch A turns on (Figure 2b), the empty capacitor is charged. After a waiting period of 120 μs , a pre-charged voltage U_0 is determined. Then, after a following period of 30 ms, the charging process is stopped and the end voltage U_1 is measured. Finally, the capacitor is discharged again again (Figure 2c). With the basic equation

$$W = \frac{1}{2} C(U_1^2 - U_0^2) \tag{1}$$

the energy on the capacitor can be calculated with reasonable accuracy, with neglecting all inrush or leakage currents present in the capacitor.

Figure 3 shows, as an example, the trend of the capacitor voltage when recorded in the laboratory. For illumination a natural light bulb (EXO Terra, “Natural Light”, 25 W, HAGEN Deutschland GmbH & Co. KG, Holm, Germany) with daylight spectrum is used. The voltages U_0 and U_1 are highlighted.

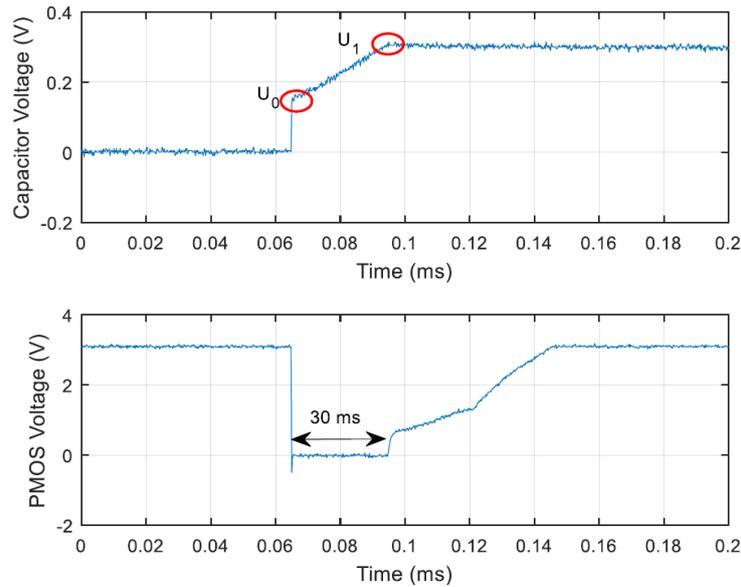


Figure 3. Example measurement of the capacitor voltage measurement and the voltage over switch A during a 30 ms measurement cycle. Illumination is happening with an incident power of 130 W/m² from a full spectrum daylight bulb.

Figure 4 shows the principal set up used for outdoor measurements. It consists again of two PV cells parallel [4] and the aforementioned DC-DC converter [5]. In parallel, solar irradiance is measured with a pyranometer (CMP3 and Ampbox, Kipp&Zonen, Delft, the Netherlands) [6].



Figure 4. Picture of the setup placed onto a sunny balcony.

3. Results

Measurement results for a full day are shown in Figure 5, with the ideal energy income is calculated based on the actual sun position. A value is recorded every minute. The pyranometer

values (left y -axis) are starting to increase later than expected due to shadow effects of the building close to the measurement site. Also, reflections from the building wall and windows are leading to a higher peak radiation up to 990 W/m^2 . With Equation (1) the resulting energy on the capacitor is calculated. The values are following the pyranometer measurement. At the beginning and the end of the day (box), a larger difference is visible between pyranometer values and the capacitor measurements. This is due to the fact that the DC-DC converter requires a certain start-up voltage for resuming operation. Taking into account by a calculation, that the efficiency of DC-DC conversion ($\eta_{\text{DC}} \approx 60\%$) and PV cell ($\eta_{\text{solar}} \approx 8\%$), the actual energy income is only about 10–15% of the expected energy available from solar radiation at the effective cell area. Further influences arise from the realistic sun angle, calculated by the ideal course of the sun, from the resulting effective area of the PV cells and more effects. Especially the delayed start-up and earlier drop-out of DC-DC conversion is an important influence parameter reflected in the measurement results. This effect does also show the importance of measuring energy income at the appropriate power transfer point, here the output of the DC-DC converter, to gain realistic data.

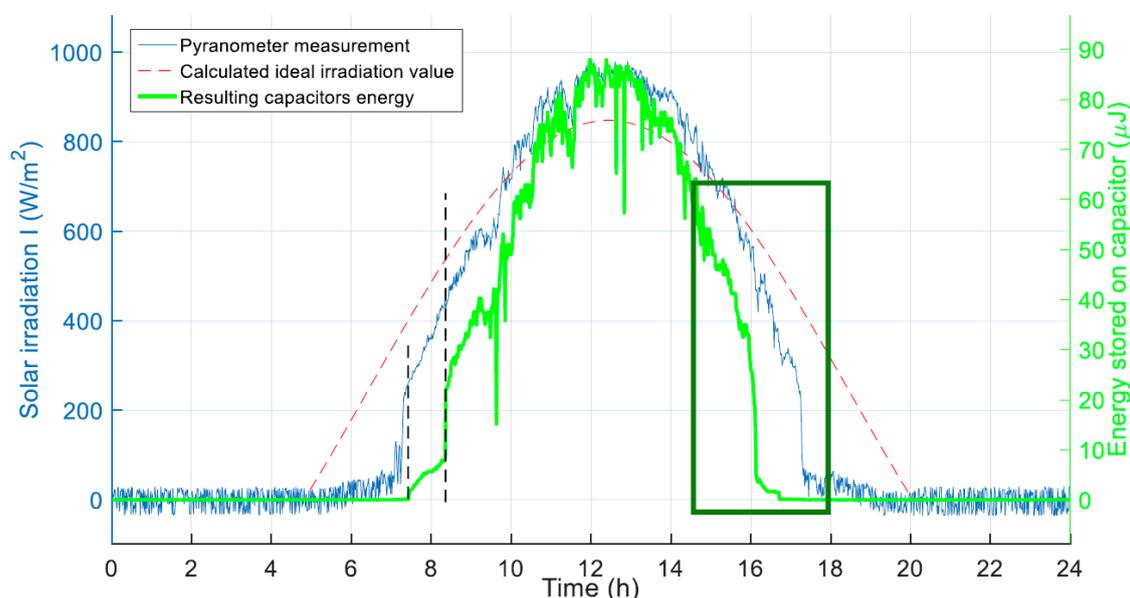


Figure 5. Measurement of the energy income estimated with the stated system and compared to the measured global radiation, at the 22 April 2018 in Freiburg, Germany.

4. Discussion

The results of the measurements are representing the desired properties. Not only the shadowing of the PV cells is viewable in its effect, also the start-up voltage of the DC-DC converter, which allows, in this set up, effective energy harvesting only above an irradiation of 400 W/m^2 takes severe influence on the actually harvested energy. Both is important information for a realistic estimation of the energy income. The difference between the expected energy income when calculated from datasheet values of the components and calculated from the capacitive energy measurement has to be further examined, since here are still deviations detected. Nevertheless, the idea of sensing the energy income works and is adaptable to different types of energy harvesting systems.

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