

Department of Mechanical Engineering

Synthetic High Resolution Solar Data

Author: David Alexander McCracken

Supervisor: Dr Nick Kelly

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Abstract

In the modern world the collation of data relating to a field of study is the key to improving and advancing that field of study. This statement rings true for the study of how the variability of solar radiation impacts on dependent systems such as photovoltaic's and systems which are intrinsically linked to solar radiation levels such as lighting and heat gains and thus energy usage, within the built environment.

However, for years the analysis of energy generation using photovoltaic's and the modelling of energy usage within the built environment have been studied using hourly averages of solar radiation data. This for a long time was deemed adequate but as the drive to reduce global warming has intensified, so too has research in these areas and it has become abundantly clear that to truly improve and understand the effect solar variability has on these dependents, higher temporal solar radiation data is needed. Only now have institutions and universities started to measure solar data to such a resolution that it can be used in detailed investigations, and currently only a small number of places are doing so limiting the locations of these studies. As such a method of turning the abundantly available hourly averages of solar data into high resolution data sets would be extremely useful to the science and engineering community. Consequently the aim of this project was to develop a simple method which could be utilised to generate 5 minute resolution data using hourly means as a basis.

In order to achieve this aim the model constructed used an approach that utilised a number of transition probability matrixes in conjunction with a first order Markov chain process, which would allow for the stochastic generation of synthetic high resolution data.

It is shown within this report that this method achieves good correlation with measured data when a frequency distribution is considered over the period of a year. This is achieved not only for the location of Reading where the original data to construct the transition probability matrixes was sourced but also for locations outside of Reading.

The model described within this report is provided freely and is available for download from the Energy systems research unit (ESRU), University of Strathclyde website.

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Table of Contents

1.0 - Introduction	10
1.1. Increase in renewable energy and Photovoltaic's	12
1.2 - Reducing energy usage in the built environment	14
1.3 – The need for high temporal resolution in solar data	14
<i>1.3.1 - Supply and demand</i>	<i>15</i>
<i>1.3.2 - Solar inverter and variable solar irradiance</i>	<i>19</i>
2.0 - Background information.....	22
2.1 - Solar Energy	22
2.2 - Solar radiation and its components.....	22
2.3 - Generation of electricity using Photovoltaic's	23
2.4 - Solar measurement	25
2.5 - Cloud cover.....	26
2.5 - Aim s & Objectives	30
3.0 -Literature review	31
3.1 - Solar models and studies of low resolution	31
3.2 - Existing computational models	32
3.3 - Current High resolution models	34
4.0 - Model Construction.....	37
4.1 - User input	38
4.2 - Generation of Clear sky index	39
<i>4.2.1 - Clear Sky Radiation(R_{so}).....</i>	<i>39</i>
<i>4.2.2 - Clear Sky model equations and operation</i>	<i>41</i>

4.2.3 - <i>Clear sky index calculations</i>	49
4.3 - Generation of 5-minute resolution Global horizontal radiation data.....	50
4.3.1 - <i>Transition probability matrix (TPM) construction</i>	50
4.3.2 - <i>First order Markov Chain process</i>	52
4.3.3 - <i>The generation of 5-minute resolution global horizontal radiation data</i> .	53
4.4 – Splitting Global horizontal radiation into direct and diffuse components.....	56
5.0 - Model Validation	61
5.1 – Method used and frequency distribution graphs	62
5.2 – Input data from different location	64
6.0 – Conclusion and Further work	66
6.1 – Conclusion	66
6.2 – Further work.....	67
7.0- References	69
8.0 - Bibliography	74
9.0 - Appendixes	75

List of Figures

Figure 1: CO2 emissions by source	11
Figure 2: UK energy production by fuel type.....	11
Figure 3: Installed capacity Photovoltaic's.....	13
Figure 4: Summer and winter demand profiles	16
Figure 5: Supply/demand profile	18
Figure 6: Supply/demand profile	18
Figure 7: Characteristic I-V curves of PV arrays.....	20
Figure 8: String inverter	21
Figure 9 : Micro inverter.....	21
Figure 10: Solar spectrum.....	22
Figure 11: Solar spectrum.....	24
Figure 12: Shaded Pyranometer.....	26
Figure 13: Pyranometer.....	26
Figure 14: Types of cloud.....	28
Figure 15: Overview of generation procedure	37
Figure 16: User interface	38
Figure 17: Comparison of measured and modelled data	40
Figure 18: Comparison of measured and modelled data (ASCE) method only	41
Figure 19: Solar altitude angle (β)	44
Figure 20: Solar Azimuth (γ_s)	44
Figure 21: Solar Zenith angle (θ_z).....	45
Figure 22: Clear sky Index (5 minute intervals).....	50
Figure 23: Transition probability matrix example.....	52

Figure 24: Operational process of model.....	54
Figure 25: Global horizontal radiation calculation procedure	56
Figure 26: Comparison of calculated and measured direct components	60
Figure 27: Comparison of calculated and measured diffuse components	60
Figure 28: Stochastic generation difference (Good fit to measured data)	61
Figure 29: Stochastic generation difference	62
Figure 30: Frequency distribution for Reading (2008).....	63
Figure 31: Frequency distribution for Reading (2010).....	64
Figure 32: Frequency distribution for Lewis (2010)	65

List of Tables

Table 1: Generation tariffs for solar PV	14
Table 2: Transmittance of different types of cloud.....	29

1.0 - Introduction

Since the 1980's when scientists and engineers started to show that instead of the expected global cooling the world was in fact entering a period of global warming a vast amount of resources have been invested into firstly determining a reason for such a rise and secondly investigating ways of combating the rise in annual mean temperature [1].

The first issue as to what in fact causes global warming has been a contentious subject since the issue was first reported, there have been many theories over the last 30 years but the most compelling and convincing of these is the fact that humans are at least partly to blame through a process known as the greenhouse effect[2].

The greenhouse effect is the process by which greenhouse gases absorb and re-emit thermal radiation. Greenhouse gases are gaseous constituents of the atmosphere that effectively absorb and re-emit radiation at specific wavelengths, as the radiation is emitted on all sides part of this radiation is therefore emitted back towards the earth's surface thus giving the effect that greenhouse gases trap heat within the surface troposphere system [3]. The primary greenhouse gases are water vapour (H₂O), carbon dioxide (CO₂), nitrous oxide (N₂O), methane (CH₄) and ozone (O₃) however there are a number of entirely man made gases which also contribute to the greenhouse effect these include halocarbons(CFCs) and other chlorine and bromine containing substances , hydrofluorocarbons (HFCs) and perfluorocarbons (PFCs) [3].

The reason humans are to blame for this increase in annual mean temperature is that since the industrial revolution the amount of greenhouse gases being put into the atmosphere has increased significantly. This is evident for all greenhouse gases but carbon dioxide is often focused on as it has the biggest percentage contribution to the greenhouse effect due to the concentration of it within the atmosphere (CO₂ makes up approximately 29% of all greenhouse gases [4]) this increase is also something that has been caused mainly due to the activities of humans.

Since the industrial revolution the CO₂ concentration has increased by approximately 39% from 280 to 390 parts per million (ppm)[5]. In the UK carbon dioxide accounted for approximately 84% of the UK's total greenhouse gas emissions in 2009

(approximately 473.7 metric tonnes)[6].As would be expected the CO₂ emission discussed previously are emitted from a variety of different sources ranging from energy to transport, figure 1 however shows that the biggest single contributor to the CO₂ output of the UK is the energy sector with 38.82%, this is because currently 73.5% of the UK's energy comes from either the burning of coal or natural gas[7], both of which produce large quantities CO₂ when burned, the energy sector is followed closely by the CO₂ emissions emitted from the built environment (i.e. residential & business) with a combined total of 33.13% of the UK's Co2 emissions, a breakdown of these figures can be seen in figure 1 & 2.

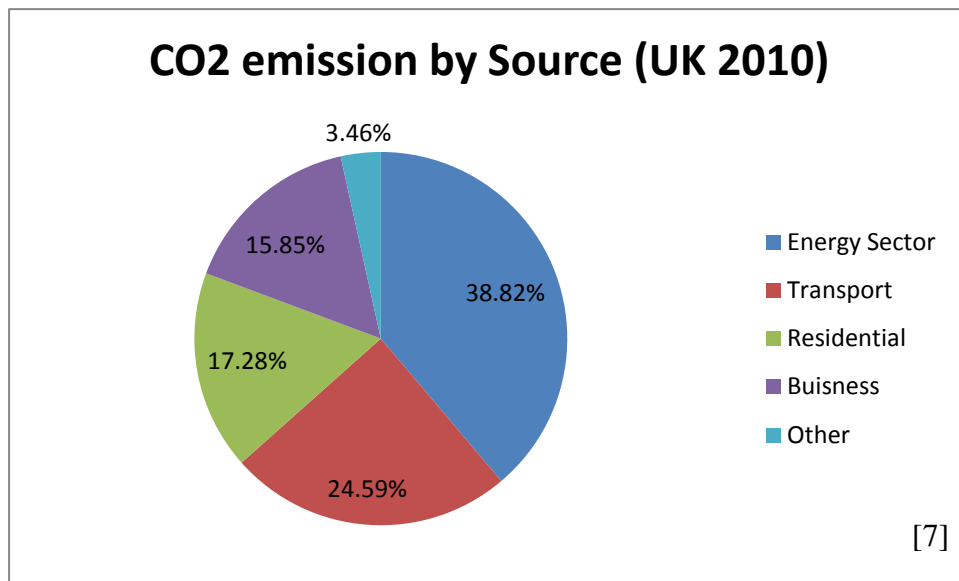


Figure 1: CO2 emissions by source

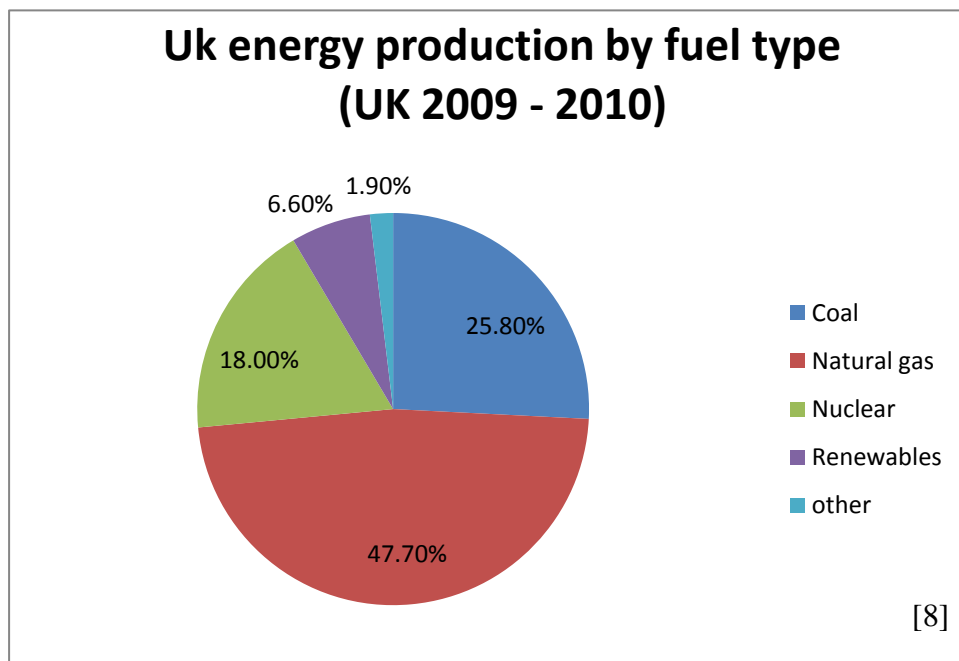


Figure 2: UK energy production by fuel type

As such it stands to reason that a reduction in these sectors would help the UK in reducing its greenhouse gas emissions and tackle global warming. In an aim to achieve this, a variety of government legislation has been implemented with the specific objective of reducing the CO₂ output of the country. The first major piece of legislation was the “Kyoto protocol” this aimed to reduce the greenhouse gas emissions of the world’s major countries (referred to as Annex 1 countries) which the UK is a constituent of, by 5.2% from 1990 levels [9]. However this has since been superseded in most cases by greenhouse gas emission targets set by the individual constituent countries which make up Annex 1. In the UK a number of different targets exist as emission targets can be set by the devolved governments therefore Scotland has a slightly different target from the rest of the UK. The current target set by the UK government is outlined within the “climate change act 2008”, this document proposes a target of reducing the greenhouse gas emission of the UK by at least “80% by 2050, compared to 1990 levels” it also implements an interim target of reducing greenhouse gas emission by 34% by 2020, again this target is against a 1990 baseline [10]. Scotland like the rest of the UK also plans on reducing CO₂ emission by 80% from 1990 levels by 2050 however the interim target for 2020 has been increased from 34% to 42% [11].

1.1. Increase in renewable energy and Photovoltaic’s

Both the greenhouse gas emissions targets set out in the UK climate change act and the fact that the largest share of these emissions stems from the production of energy has resulted in renewable energy becoming increasingly popular within the UK as this is seen as a way of reducing our reliance on fossil fuels and therefore reducing overall CO₂ emissions. Since 2005 the generation capacity of electricity from renewable sources has increased by 49% (from 16936 GWh to 25222 GWh [12]). This generation capacity has been made up of wind (both onshore and offshore), hydro (both small scale and large scale), biomass, wave, tidal and solar photovoltaics. The majority of this generation to date is produced by older technologies such as wind, hydro and Biomass. However other technologies such as photovoltaics are gaining popularity and generation capacity in this area is increasing. Over the last 5 years the total generation capacity of photovoltaic’s in the UK has increased fivefold this is shown in figure 3 below.

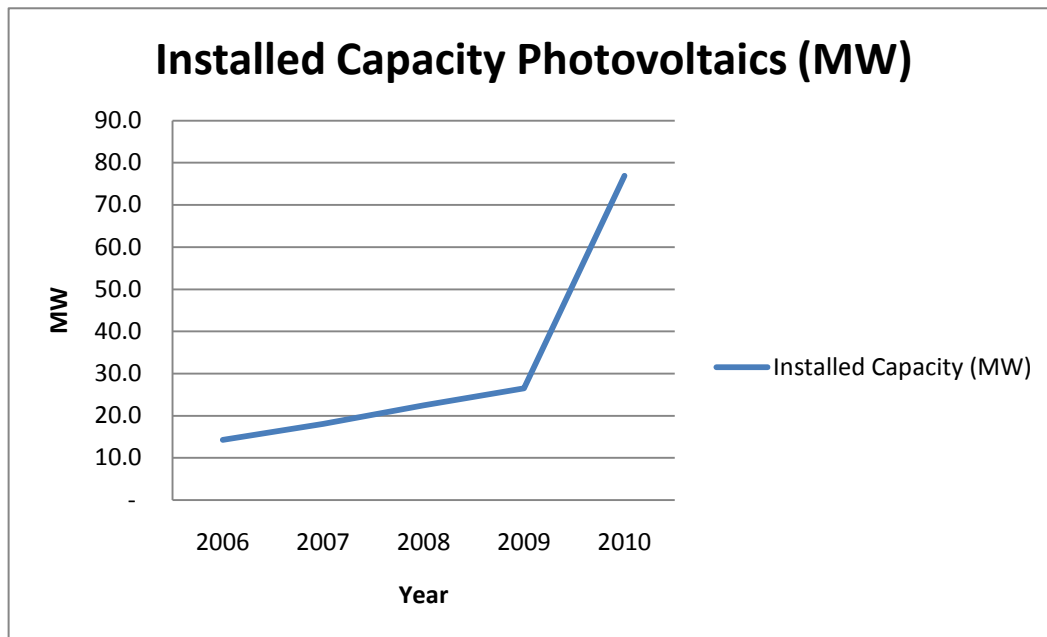


Figure 3: *Installed capacity Photovoltaic's*[12]

As can be seen from figure 3, there has been significant increase in recent years of generation capacity from Photovoltaic's this has been brought about by increased support from government incentives which has seen non-commercial scale photovoltaic installations increase dramatically. By generating electricity using photovoltaic's the owner of the building/land will be eligible for the government incentive scheme called "feed in tariffs" this scheme, which was introduced on the 1st of April 2010 pays the owner a fixed rate of income called a "Generation tariff" for every kWh of electricity that is generated. The amount paid varies dependent on the scale of the installed device, table 1 below shows the various generation tariffs currently available for solar PV. If this generated power is then exported to the grid a further "export tariff" of 3.1 p/kWh is paid to the owner. This type of incentive scheme has been utilised throughout Europe and like the UK has seen dramatic increase in the amount of PV installation installed across the board [13][14].

Energy Source	Scale	Tariff (P/KWh)	Duration (Years)
Solar PV	≤4 kW new	37.8	25
Solar PV	≤4 kW retrofit	43.3	25
Solar PV	>4-10kW	37.8	25
Solar PV	>10 - 100kW	32.9	25
Solar PV	>100kW - 5MW	30.7	25

Table 1: Generation tariffs for Solar PV [15]

1.2 - Reducing energy usage in the built environment

The solar radiation used in the production of energy via photovoltaic's also plays its part in the reduction of energy usage within the built environment which as shown in section 1 is also a key contributor to the UKs overall greenhouse gas emissions, this is being achieved by understanding how solar radiation and natural light can be best utilised within a building, and how buildings and structures can be improved to further utilise this resource and reduce energy. These sorts of tasks are being investigated more and more during the design phase of buildings and structures via modelling, where integrated energy modelling tools such as ESP-r are used.

1.3 – The need for high temporal resolution in solar data

Thorough investigations into photovoltaic's and building energy usage are generally carried out in conjunction with measured solar radiation data which is used to stipulate the intensity of the sun throughout the simulations. Until fairly recently it has been felt it is adequate to use hourly averages of solar radiation data for these activities. As the drive for reduction of greenhouse gas emissions continues and the amount and detail of the modelling of PV output and energy usage simulations increase it has been widely realised that hourly averages of solar radiation are no longer sufficient as they mitigate the majority of the variability of the sun brought about by cloud coverage which is experienced at higher temporal resolutions.

As such there has been a greater call from the science and engineering community for higher resolution solar measurements to become the norm. A technical report realised in early 2008 by the National Renewable Energy Laboratory (NREL) titled “*Solar resource assessment*” [16] aims to explore the availability of solar data and review models which can generate accurate solar data which are pertinent to the implementation of distributed solar PV, although this report mainly focuses on the United States of America and hones in on PV its findings are indicative to the state of affairs seen throughout the world for solar measurement data. It states that the resolution of current data is “*inadequate*” for analysing systems which are dependent of solar radiation and that “*high quality ground measurement stations that can provide adequate time series data are very limited in number*”, to this end it concludes and recommends, among other things, that there is a need for high-resolution (15 minutes or less) resource data that has “*either been derived from hourly measured or modelled results and/or for additional high quality measurement stations*”[16]. Many other reports for example Vijaykumar et al [17], Skartveit and Olseth [18], Janark[19], Suehrcke and McCormick [20,21,22], Walkenhorst, O [23], Gansler et al. [24], Jurado et al. [25] Olmo et al [26] and Tovar et al. [27,], have also shown important and critical differences between low resolution (hourly) and high resolution solar data (15 minute or less) and its effects on a variety of systems which are dependent on solar irradiance.

The subsequent section of the report aims to explore the issues relating to the inherent variability of solar energy and how this in turn impacts on a variety of issues such as supply and demand, thus justifying that there is a major need for high resolution solar data for use in a variety of industries.

1.3.1 - Supply and demand

All national electricity supply structures have variations in their power demand over the course of a day, these demand characteristics can be fairly predictable over a 24 hour period with a minimum usually occurring during the night and peaks usually occurring in the morning and evening when most of the population is preparing for or arriving home from work. These peaks and troughs also vary seasonally with the winter periods requiring more energy as a result of heating loads which are obviously

not experienced to the same extent during the summer months; national grid demand profiles are shown below in figure 4 to give an example of the variations which are experienced.

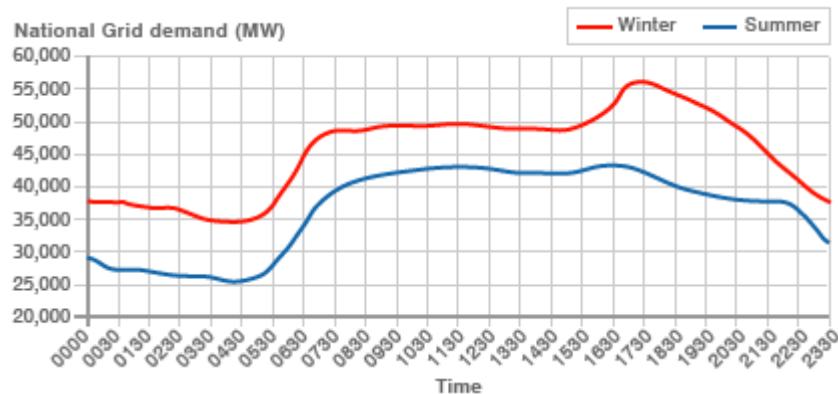


Figure 4: Summer and winter demand profiles [28]

The area under the curve in figure 4 represents the daily energy demand of the UK it shows that the base load in the UK is of the order of 25GW during the summer, rising to just under 35GW during the winter. However this rises to a peak of approximately 44GW and 56GW in summer and winter respectively due to the daily variations which were discussed previously.

It is the National Grids duty to ensure that the supply of electricity matches the demand shown in figure 4 at all times, failure to do so can result in voltage dips, surges, frequency oscillation, harmonics and even blackouts. To ensure this does not occur a mixture of generation plant is utilized [29]. Some of these generation plant operate continually, these are generally used to provide the “base load” (i.e. the load which always need to be met). Other plant is kept on a “spinning standby” this is where it is connected to the grid and operating at part load so as to ensure stability of connection , this is most often done with steam plant such as coal and gas and allows for rapid increase of generation so as to match demand within an adequate time scale. Additional power stations are available for “instant start-up” this is when the plant is started from cold to match demand although the ability to do this is heavily dependent on fuel type so power stations such as hydro, gas and diesel plants are normally used for this type of generation [29].

Most renewable energy systems however cannot be controlled in the same way as the power plants that are discussed above, as they only operate when the natural resource

they exploit allows them to (i.e. when the sun is shining or the wind is blowing etc), this variability therefore increases the need for further spinning standby as backup.

The solar energy which is discussed throughout this project falls within this bracket as it can be highly variable dependent on the meteorological conditions observed, this would therefore affect the supply of any generation from Photovoltaic devices both embedded and on a commercial scale. A recent report by the North American Electric Reliability Corporation (NERC) [30] suggested that the output of some of their larger multi- MW photovoltaic plants currently operation in the United States can change by “*more than 70% in five to ten minutes on partly cloudy days*” [31] therefore if a installation of this size was to be modelled it can be clearly seen that it would be highly inappropriate to use hourly averages of solar radiation data.

Currently however no single generation site within the UK is to the same scale as the plants discussed within the NERC report, although this is something which is rapidly changing, for instance planning approval has recently been given for a 5MW installation at Bodmin, near Cornwall [32]. On the whole however the UK’s photovoltaic generation is mostly decentralised unlike these commercial projects, although this decentralised approach has been shown to have a smoothing effect on the variability of power output as the PV panels are located over a wide geographic area and thus exposed to a wide variety of meteorological conditions[33], these localized effects should not be discounted, the graphs below show the net resultant effect of an average demand profile (Taken from annex 42[34]) and a simple domestic PV installation of 5 m² on the horizontal plain, as can be seen the variability of the high resolution data produces a vastly different profile from the graph were hourly averages are used(Positive is supply to the grid / negative demand from the grid). This would inevitably have an impact on the grid and if considered over the lifespan of the installation the output and therefore savings/profit would be very different.

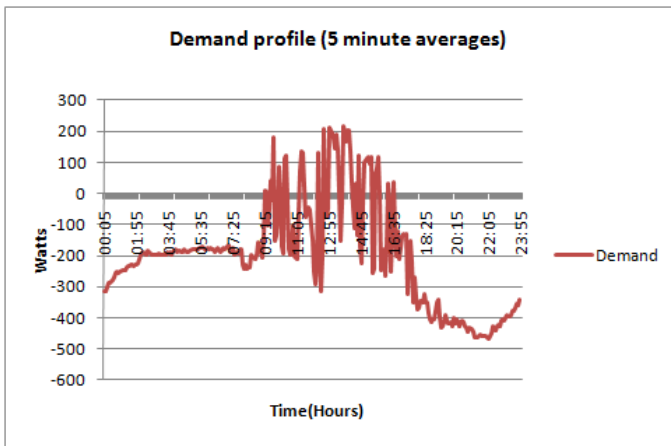


Figure 6: Supply/demand profile

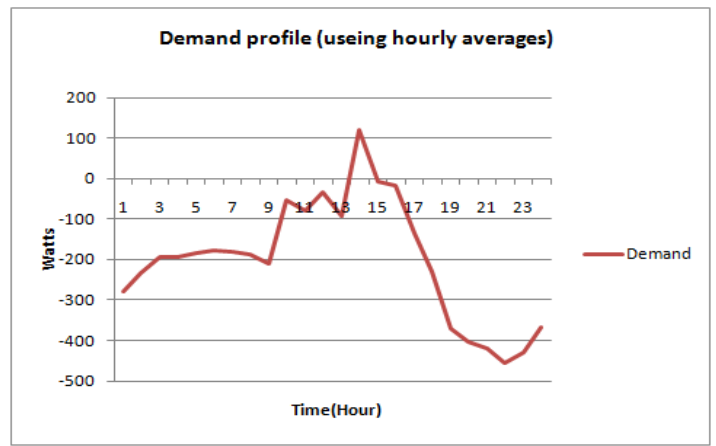


Figure 5: Supply/demand profile

As well as the impact on supply of electricity due to the variability of solar irradiance the demand for electricity also varies, this is mainly due to the increased use of devices which are intrinsically linked to the outdoor irradiance levels, such as lighting and heating. As passing clouds reduce interior light levels and solar gains humans will react by switching on artificial lighting and heating to compensate, this can be highly dependent on the task they are carrying out at the time, as humans also have the ability differentiate between passing clouds and constant dull skies. Quantifying to what degree this will affect the buildings energy usage can be extremely difficult, in an effort to calculate these affects it has become increasingly popular for structures to undergo a period of simulation during their design phase. This is where integrated energy modelling tools such as ESP-r and RADIANCE are used so as to establish thermal, visual and acoustic performance. Simulation programs like this however tend to use hourly data or have limited high resolution data for specific sites thus underestimating the amount of energy and light available to the occupants the author envisages that the model developed throughout this report could be used to generate high resolution data in such cases giving better estimates of the variable light levels and solar gains during a typical day and thus there impact on energy usage of the structure. [35][36]

As supply and demand are increasingly effected by the variability of renewables the national grid can be exposed to problems in terms of predicting demand and security of supply, however National grid fully encourages the increase in renewable

generation and has developed ways of combating the variability brought about by this increase, besides increasing spare generation capacity. Currently the national grid intends to implement a combination of schemes to tackle not only the variability of renewables but also the expected and unexpected plant shutdowns which are experienced throughout the national electricity supply structure. These include “better forecasting, interconnectors to the European network, electricity storage, smart grids and use of flexible generation”. For this project the emphasis on better forecasting is important as this is another area the model being developed by the author may be useful. Currently most of the forecasting work done by the national grid has been in the area of wind power this is entirely understandable as this is currently the biggest variable source of power within the UK, but as has been shown in previous sections embedded generation using photovoltaic’s is increasing and if current trends continue this could rise to approximately 3500MW of generation capacity within the UK [37] this is also been recognised in the national grids consultation paper “*Operating of electricity transmission network in 2020*” which aims to explore ways of dealing with the changing supply and demand market [37].

With greater availability to high resolution solar data there will be greater ability to model and predict the likely hood of energy generation from devices such as Photovoltaic’s. Conversely the ability to determine the likelihood and usage characteristics of demand side devices such as lighting will increase thus inevitably resulting in a decrease in the amount of “spinning standby” which needs to be put in place to ensure stability to the grid.

1.3.2 - Solar inverter and variable solar irradiance

As has been explored in the previous section photovoltaic installation outputs are obviously majorly affected by solar variability. At the heart of these grid-connected photovoltaic devices is the “Inverter”. The inverter is the photovoltaic’s arrays connection to the grid and performs a number of functions. Firstly the main function of the inverter is to covert direct current (DC) into alternating current (AC) which can be delivered to the grid , secondly the inverter also provides a maximum power tracking function (MPPT) , this is the method used to get the maximum possible power from the PV array, known as the Maximum power point (MPP). As solar irradiance and PV temperature change, so too does the output efficiency known as an

I-V curve, the maximum power point is located on the “knee” of the I-V curve (see figure 7) therefore in order to get the most from the PV array as the conditions experienced by the array change maximum power point tracking is used. This is done by complex algorithms which sample the output and apply a resistance so as to always remain at the maximum power point.

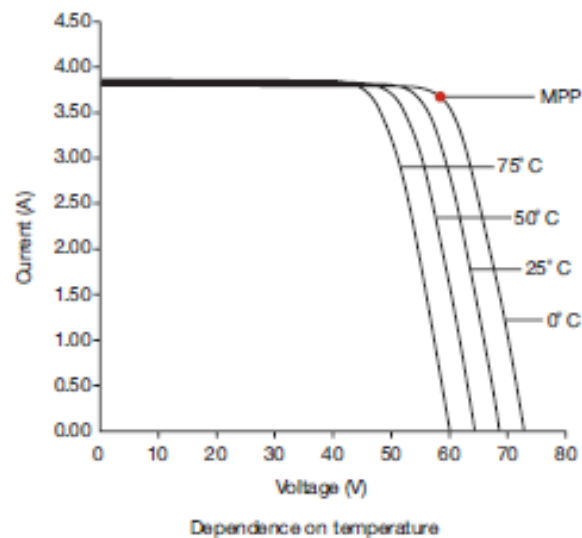


Figure 7: Characteristic I-V curves of PV arrays [38]

Thirdly, for safety and grid stability the inverter is responsible for grid disconnection in the case of the frequency or voltage being different from pre-determined standards, or in the case of an “island” meaning the grid has suffered an outage [38].

Currently in the UK there are two architectures used when it comes to inverters these are “String inverters” and “Micro inverters”. By far the most popular of these two is the string inverter approach mainly due to it being the cheaper option, unfortunately this approach can be effected by the variability of cloud cover when considering a photovoltaic array consisting of a number of solar panels. To understand how this occurs the difference between a “string inverter” setup and “micro inverters” setup must first be explored.

String Inverters receives DC power from an number of different solar panels or a “string” of panels this DC power then flows into a single inverter which converts the direct current in to alternating current (see figure 8), “Micro inverters” on the other

hand make use of smaller inverters which are attached to individual panels or small groups of panels (see figure 9).

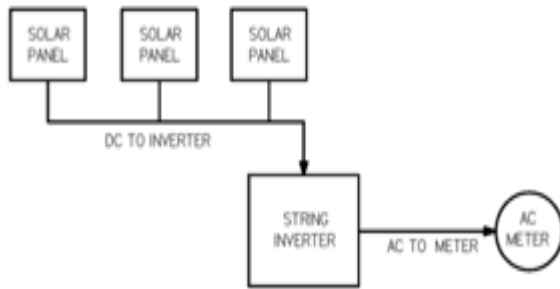


Figure 8: String inverter [39]

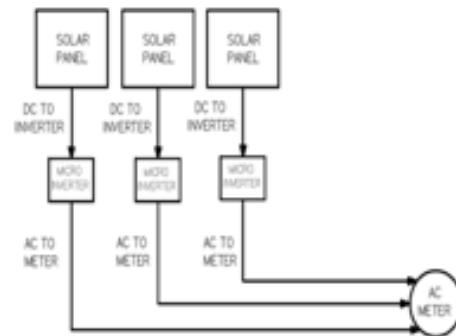


Figure 9 : Micro inverter [39]

Utilizing a string inverter, the panels are effectively connected in series therefore whatever happens to one of the panels will inevitably have an impact on all of the panels, when passing clouds partially shades just one of the panels connected in series the overall output of all the panels in the array is effected, this is not the case with the micro inverter.

The solar variability caused by passing clouds also has an effect on the predicted output of any PV installation, if the solar variability causes the output to drop below a certain level the inverter shuts down and enters a standby mode, upon the cloud clearing the inverter will enter a start up procedure during which it monitors AC voltage and frequency before eventually going online, during this time there is no power being supplied, furthermore the inverter actually draws power from the grid. The use of hourly averages of solar radiation data during any analysis of a proposed system will not take into account this variability and thus any points at which the installation enters the shutdown/start-up procedure will not be fully accounted for, thus inevitably leading to incorrect predictions of output. Therefore a method of generating high resolution data from these hourly means would be beneficial for manufacturers and installers of PV devices who wish increase the accuracy of their predictions.

2.0- Background information

2.1 - Solar Energy

The amount of solar energy which reaches the earth's outer atmosphere is 1367 W/m^2 this number is referred to as the "Solar constant", however as the solar energy passes through the atmosphere it is absorbed and reflected by the atmosphere itself, clouds and particles in the sky thus meaning when the energy arrives at the surface of earth it has a maximum intensity of approximately 1000 W/m^2 . The wavelength of this solar radiation ranges from $0.3 \mu\text{m}$ to $2.5 \mu\text{m}$ this incorporates ultraviolet, visible and infrared light, figure 10 shows the solar spectrum of solar radiation of both extraterrestrial and terrestrial radiation [40].

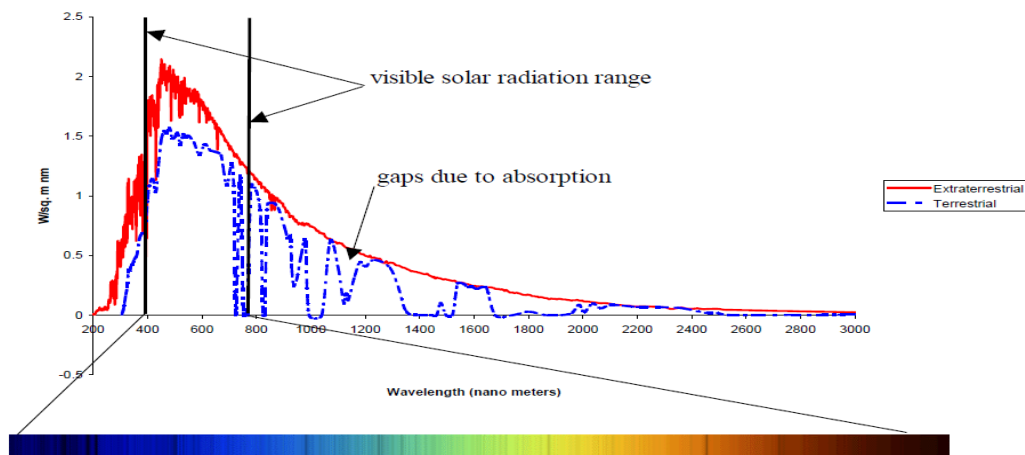


Figure 10: Solar spectrum [40]

2.2 - Solar radiation and its components

The above section makes many references to solar radiation which is incident on the earth's surface by this the author is typically referring to global radiation (I_t). Global radiation however is made up of two distinct components which it is important to distinguish between at this point, these are the direct beam radiation and the diffuse radiation.

Direct beam radiation (I_b) is the portion of global radiation which travels in a straight line and reaches a specific geographic location direct from the sun. Whereas diffuse radiation (I_d) is the portion of global radiation which is scattered in the atmosphere by

clouds and particulates and therefore does not arrive direct from the sun. They are related such that the combination of direct and diffuse is equal to the global radiation reaching a point on the earth's surface as shown in equation 1 [40].

$$I_t = I_b + I_d \left(\frac{W}{m^2} \right) \quad (\text{Equation 1})$$

2.3 - Generation of electricity using Photovoltaic's

Photovoltaic devices respond to both direct and diffuse radiation, however by their very method of construction they have a fundamental limit as to the amount of solar radiation which can produce energy, this limit is imposed due to the physics within the PV cell itself.

To produce energy a PV cell is constructed in a certain way so as to encourage the flow of electrons. A complete cell consists of two pieces of a semiconductor material placed one above the other. One of the sections will be doped (*the process of treating a semiconductor material with chemicals to acquire special properties*) using a substance with 3 outer electrons (such as Boron) this will create a shortage of electrons leaving behind "holes" thus creating a positive charge; this section is referred to as the "P-type layer".

The other section will be doped with a substance with 5 outer electrons (such as Phosphorous), this creates an excess of electrons and gives the layer a negative charge, This section is referred to as the "N-type layer" [41][42]. The area between the P-type layer and the N-type layer is called the P-N junction, the P-N junction has a strong reverse electric field that acts to maintain the separation between electrons and holes.

When a light photon from incoming solar radiation with sufficient energy hits the PV panel it excites an electron which then has the ability to break free and move into a "conduction band". Once free electrons are swept to the N-type region, conversely the holes left by the absence of the electrons are swept into the P-type region. If metal contacts are connected to both the positive and negative sides an electrical current can be taken and used to run a load.

However as mentioned previously there is a fundamental limit which relates to the energy required to free an electron from the atom so as it can enter the conduction state. This relates to the effect the absorption of a photon has on an electron. Electrons tend to form valence bonds with other atoms to create an element, when light photons excite an electron it can move from its stable energy level to a higher energy level, at this higher energy level the electron is capable of conduction hence why it is generally called the conduction band. The difference in energy between the valence band (Stable energy level) and the conduction band is called the “Bandgap” energy therefore it stands to reason that only photons with energy level greater than this “Bandgap” energy can excite electrons enough so as to move them from the valence band to the conduction band. This provides a fundamental limit to the efficiency of the solar panel as it means that only a fraction of the solar radiation incident on the earth’s surface has enough energy to promote an electron to the conduction band, figure 11 shows the solar spectrum shown in figure 10 with the “Bandgap” energy limit taken into account [41].

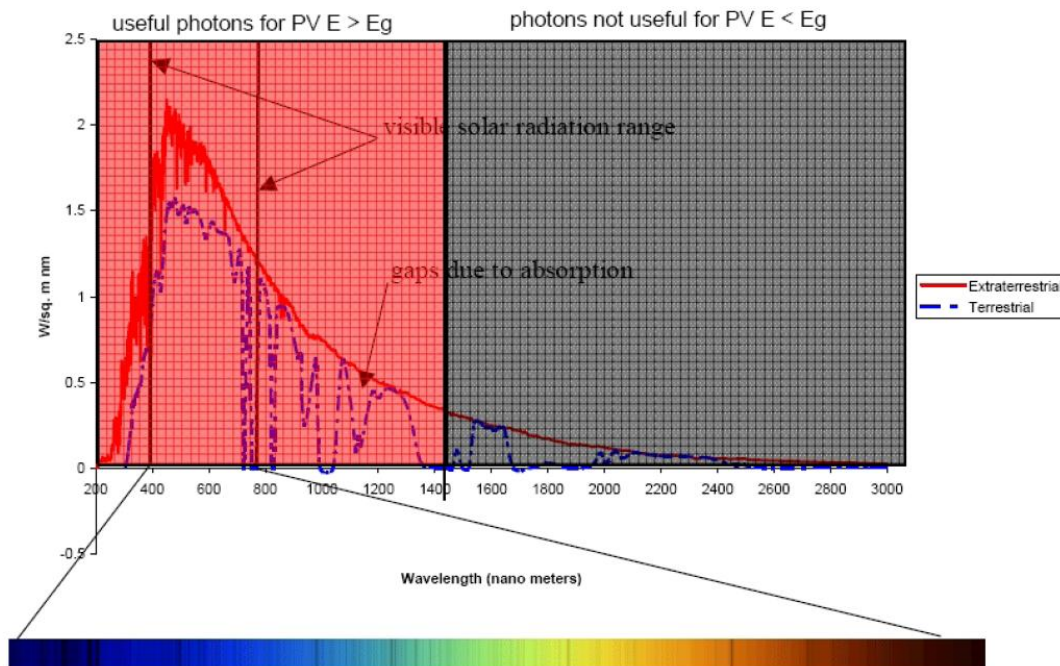


Figure 11: Solar spectrum [40]

2.4 - Solar measurement

When a photovoltaic array is proposed the energy which can be generated is usually calculated using modelling tools this requires the use of solar data, the most common type of this data is termed “Global horizontal radiation” which as discussed in section 2.2 consists of both a direct and diffuse component, when global horizontal radiation is measured it is done so by use of a device called a Pyranometer.

Pyranometer

As discussed previously the device used to measure total global radiation (I_t) on a planar surface is known as a Pyranometer. Pyranometer's consist of a white metallic housing and two hemispherical glass domes inside of which a black metal plate and a thermopile sensor are located.

As solar radiation falls onto the Pyranometer it passes through the glass hemispherical domes, the purpose of these domes besides protecting the components inside from the influences of wind and rain etc, are to limit the spectral response to between $0.3\mu\text{m}$ and $2.8\mu\text{m}$ while at the same time ensuring that the Pyranometer still has a field of view of 180° . Once the solar radiation passes through the hemispherical domes it then falls onto the black metal plate absorption surface thus the solar radiation is absorbed and the radiation is converted to heat.

The heat generated within the Pyranometer therefore directly depends on the irradiance falling onto the metal plate absorber thus it can be used to determine the broadband solar radiation. This is done by the use of the component called a thermopile [43]. A thermopile is a device which consists of a number of thermocouples generally connected in series. The thermocouples allow the thermopile to convert thermal energy into electrical energy; this is done by utilizing a junction of two dissimilar metals which when heated or cooled produce a voltage which is proportional to the temperature change, through the use of a voltmeter it is then possible to calculate the total global radiation incident on the Pyranometer.

By modifying a Pyranometer it can also be used to measure diffuse radiation only. This is done by installing a shade ring which will block out the direct beam radiation coming from the sun, however some diffuse radiation is also generally blocked out

and this would need to be accounted for when using such measurements. Images of a Pyranometer and shaded Pyranometer are shown below in figure12 & 13.



Figure 13: Pyranometer [44]



Figure 12: Shaded Pyranometer [44]

2.5 - Cloud cover

When looking at solar measurements using devices such as a pyranometer it is important to note the change in intensity of the radiation recorded, although a number of things can influence the recorded values such as shadows of buildings and trees etc the predominant reasons for rapidly changing values of solar radiation is cloud cover. Clouds can be classified into many types and sub-types dependent on their height and appearance however it is possible to place most clouds into four different categories based on their altitudes, these are; High level, medium level, low level and vertical clouds. However the naming convention within these divisions is generally based on how the clouds are formed, so clouds formed by localized vertical currents where the moist air is carried upwards above the point where condensation occurs are termed “Cumuli-form” clouds, whereas clouds formed without strong localized vertical currents but instead by the lifting of entire layers of air are called “Strati-form” clouds, these cloud types are often spread out in layers [45]. In addition to the wording describing how the cloud is formed the words “nimbus” and “Fractus” often prefix or suffix. “Nimbus” indicates a cloud which is likely to precipitate and “Fractus” is used to describe a cloud which has been fractured into pieces by strong winds. The four different categories of clouds are further explained below.

High level clouds generally form above 6000 meters and consist of predominantly ice crystals rather than water droplets due to the low temperatures at these altitudes; this division includes cirrus, cirrocumulus, and cirrostratus clouds. Cirrus clouds are white clouds which have a feather like appearance and are found at heights of about 7000m, cirrocumulus clouds are small clumps of cloud which appear in rows often between heights of 6000m – 13000m. They are sometimes called “mackerel sky” due to them covering the sky in such a way that they resemble fish scales, another key attribute of this cloud is its unlikely-ness to join together to form more substantial cloud bodies. Cirrostratus clouds are clouds which often form around the sun or the moon but in such a way that the sun or moon is still visible giving a halo effect [46]. (See figure 14)

Medium level clouds form between 2000 – 6000 meters and can consist of both water and ice particles depending on the temperature and height of formation. Medium level clouds include Altostratus and Altostratus. Altostratus clouds are a grayish white and can appear as irregular cloudlets or in bands or rows, often with one side of the cloud being darker than the other. Altostratus clouds are a grey, grayish-blue which often cover the entire sky allowing the sun to shine through only dimly.

Low level clouds form below 2000 meters and generally comprise of water droplets although at colder temperatures can comprise of snow and ice particles. Again there are two types of low cloud these are Nimbostratus cloud and the stratus/stratocumulus cloud. The Nimbostratus cloud form a grey, dark layer and often blanket the entire sky and are responsible for continuous precipitation. Stratus/Stratocumulus clouds are common and widespread clouds, which form a low uniform layer with a dull grey appearance [45]

Vertical clouds can form anywhere below 2000m to above 12000m, there are two types of vertical cloud, the Cumulus cloud and the Cumulonimbus Cloud. The Cumulus cloud is a cloud which highly resembles a floating cotton-ball and are recognized by the flat bases and sharp outlining, they are often isolated or in small groups. The Cumulonimbus Cloud is a cloud which is characterised by its flat anvil like formation this cloud is also known as the “thunderhead” due to its likelihood of

producing heavy thunderstorms the higher the vertical development of such a cloud normally dictates the severity of the thunderstorm, these clouds often bring thunder and lightning as well as heavy precipitation. Figure 14 below clearly shows all of the clouds described in this section.

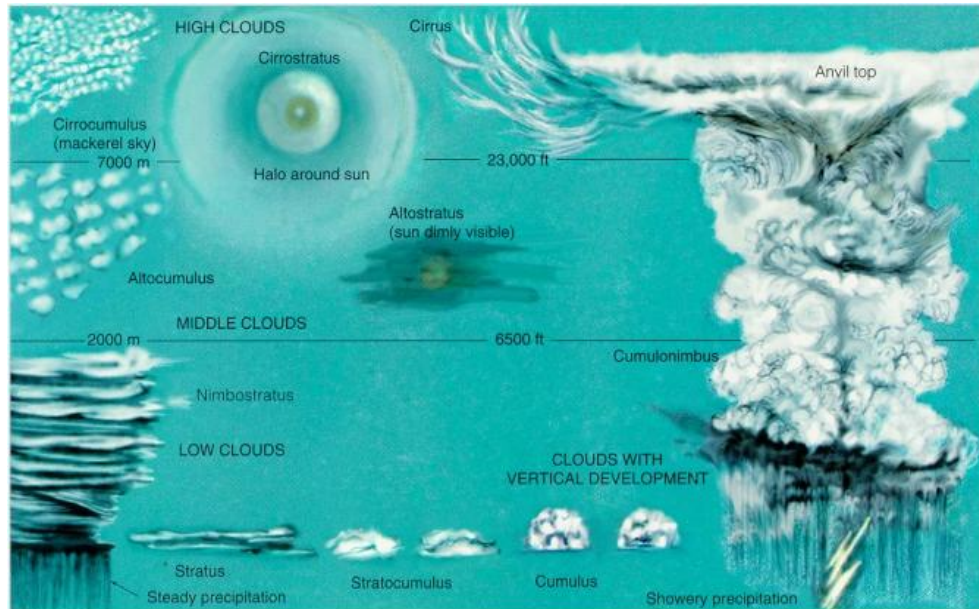


Figure 14: Types of cloud [47]

As discussed previously cloud formations are the main cause of reductions in the amount of solar radiation that is experienced at a specific geographic location, this is because as the direct beam solar radiation attempts to pass through the cloud it is scattered rather efficiently by the droplets of water and ice that make up the cloud, this scattering transforms intense direct beam radiation into less intense diffuse radiation thus the intensity of the solar radiation decreases dependent on the depth of the cloud as more scattering occurs. As each type of cloud has its own set of characteristics the amount of solar radiation which is scattered is partly dependent on which type of cloud the radiation passes through. Table 2 gives an estimation of transmittance dependent on cloud type [48].

Cloud type	Transmittance
Cirrus	0.8
Cirrocumulus	0.85
Cirrostratus	0.69
Alto cumulus	0.48
Altostratus	0.35
Nimbus	0.11
Stratocumulus	0.25
Cumulus	0.26
Cumulus congestus	0.24
Cumulonimbus	0.18
Stratus	0.15
Fractus	0.33

Table 2: Transmittance of different types of cloud [48]

However it should be noted that these figures are very dependent on a number of different parameters such as water and ice content, height, depth and solar elevation angles therefore should only be taken as a rough reference also the fact that the different types of cloud may occur at the same time thus on occasion overlapping further effecting the values given in table 2.

The sky condition may also be varying degrees of clear, overcast and partly cloudy this also has a significant impact on the amount of direct beam radiation which freely reaches the surface.

2.6 - Aim s & Objectives

The aim of this project is to establish a simple, easy to use model which will have the ability to transform hourly averages of global horizontal radiation into 5 minute resolution global horizontal radiation which is representative of the base hourly data. To achieve this a method is developed which utilises transition probability matrixes constructed from data graciously provided by the University of Reading and a Markov chain process. The model construction, operation and validation are shown in the subsequent sections of this report

Objectives

- Develop a simple, easy to use model which will have the ability to transform hourly averages of global horizontal radiation into synthetic high resolution(5 minute) global horizontal radiation.
- Validate the output of the synthetic data against measured data for different geographic locations.

3.0 -Literature review

This section aims to explore the work that has been carried out in relation to the generation, adaptation and understanding of solar data along with investigating a number of methods used presently which attempt to generate solar radiation data at varying degrees of resolution both through the use of input values of different time periods (I.e. yearly, monthly, hourly) and through methods such as the use of satellite imagery and meteorological characteristics related to weather and solar variability , this is done in an attempt to convey the amount of different methods currently used and their complexity and to show that there is a need for the authors method which aims to be simple and easy to use and will be made freely available.

3.1 - Solar models and studies of low resolution

Over the years there has been a number of studies and models produced and verified by engineers and scientists concentrated on the production of solar data, either attempting to predict the variability of solar irradiance of the future, or investigating trends in measured solar data so as to better understand the available resource and its dependencies, most of these models are predominately concerned with generating low to medium resolution data of monthly, daily and hourly time series values. Although the aim of the authors model is to generate high resolution data (5-minute resolution) the concepts and procedures introduced within these reports are extremely important therefore a number of low resolution models are discussed here. These models take a variety of forms including, statistical and numerical models, analytical models, neural network approaches and empirical models [49].

Numerical and statistical models such as the reports presented by Mustacchi. C et al [50] and Aguiar. RJ et al [51] attempt to generate stochastic global radiation data at low degrees of resolution namely hourly and daily averages respectively, these reports utilizing such methods as moving average (MA) , auto regressive (AR) , auto regressive moving average (ARMA) and Markov chains in order to analyse and generate data. These reports establish a significant amount of important aspects regarding the generation of stochastic solar data, through analysis of these reports it is also evident that approaches using auto regressive methods and similar techniques contain “*pitfalls*” in that such methods are “*unable to describe the statistical features*

of solar radiation”[50] and an inability to produce “*some of the basic features of probability occurrence of radiation values*”[51] however techniques which utilize a simple Markov transition-matrix approach provides a “*simple yet effective simulation device*”[50] for establishing solar radiation values.

Other methods for instance analytical approaches, which are contained within papers such as Paulescu. M et al [52] and approaches which utilize neural networks [53, 54, 55] have also been made use of in the past with an aim to generating low resolution global horizontal radiation data, however these methods tend to be complex in the method they use.

All of the reports mentioned above do not adequately take into account the variability of solar data at higher temporal resolutions, this on the whole is to do with when the majority of the models were written as there may not have been as great a need as there is currently for high resolution data.

However ,the low resolution models are similar in respect to the model developed within this report, in that they predominantly generate global horizontal radiation; as such they also give reference to methods and fundamental principles relating to the decomposition of global radiation into its direct and diffuse components which the author will use in subsequent sections of the report, In particular reports by Orgill and Hollands [56] and Erbs et al [57] appear constantly throughout these reports, these studies establish empirical methods to estimate the direct and diffuse components of global radiation, these fundamental principles are investigated in subsequent sections of the report along with revised and modern methods for achieving the same goal.

3.2 - Existing computational models

There are at present a number of computer based databases and atlases which are used to deliver geographic and time varying solar radiation values throughout the world in general these databases comprise of ground measurements, maps, reference years and geographical information, they are sometimes referred to as “*integrated information systems*” [58]. Databases and software packages such as METEONORM™, the European solar radiation atlas and the Server Satal-light program would fall under this category each of these software packages operates in a different manor and can generate different outputs.

METEONORM™ is a commercial package which provides ground source measurements from databases which were collated between 1986 to 2005, It has the ability to display information relating to temperature, humidity, precipitation etc as well as to display a variety of solar radiation values for a particular geographic location, these values are derived from monthly values and generally displayed as hourly time steps however the METEONORM™ package does contain the ability to produce high resolution data of 1 – minute time steps, as this is a commercial package there is a cost, with each licence for this software retailing at 480 Euros.

The European solar radiation atlas (ESRA) is a computational instrument which provides knowledge and helps exploitation of the solar resource available across Europe [58] this software and database package can be run in two modes the first is “maps” mode and the second is “Station” mode. In map mode the software provides maps of average monthly global, beam and diffuse radiation, using 10 years of data as basis. In station mode the software can be used to search for data from a individual station , the maximum resolution of this data is generally hourly however some station do have data at “*half hourly resolution*”[58]. This software like METEONORM™ also comes at a cost retailing at approximately 380 Euros per licence.

The Server Satal – light program is free to use online service which provides only maps of mean daily irradiation, frequency of skies being one of three states (i.e. sunny, intermediate or cloudy) and a number of options allowing comparison between certain years. Although these maps may be useful for general information about a particular site or location they do not provide data as the other maps do and the highest resolution contained on the maps is of an average daily irradiance which in turn has been derived from monthly data.

Most of these programs currently do not provide the required resolution of data and the programs which have the ability to provide such resolution are all payment services which is unlikely to be useful for students and academics, these systems can also be complex in the user interface with the professional systems having a vast array of settings.

3.3 - Current High resolution models

As has been shown continually throughout this section, scientist and engineers have studied solar energy, its decomposition and its variability in many different reports, most of the reports however have focused on the study and generation of low to medium resolution data in monthly, daily and hourly time steps, such as those shown in section 3.1, this is mainly due to the worldwide measurement of solar data being generally stored in these low to medium resolutions.

It has also been shown that as the world moves towards more sustainable means that high resolution data will become more important, this again has been shown in a number of reports, where particular attention has been paid to exploring the difference between hourly and minutely solar radiation data such as Suehrcke et al [22] and Gansler et al [24] these reports state quite clearly that the difference in variability between hourly and minutely data when taken in the context of solar systems such as photovoltaic's will lead to the electrical energy output being "*significantly different when using minutely data*" as opposed to hourly data, "*due to the variability*" which is not accounted for [24]. Therefore the aim of the author's project is to generate stochastic high resolution data of 5-minute time steps from an hourly input thus taking account of this variability as best we can.

As the problems of solar variability and its impact on systems which rely, or are dependent on the sun have been further explored, attempts by a number of groups have been made to achieve a similar goal in respect to the generation of high resolution data, in particular reports by Skartveit and Olseth[18], Walkenhorst, O et al [23] and Richardson and Thomson[59] have laid the foundations for the work the author is attempting within this report, what these reports achieve and how this relates to the authors project will be shown here.

The report by Skartveit and Olseth [18] sets out, and satisfactorily achieves the production of realistic time series data of high resolution global and beam irradiance, with hourly averages provided as an input. It achieves this by using probability distributions of the short term irradiance data, and combining this with an autoregressive model thus producing intra-hour short term, time series outputs. Although the modelled results displayed within this paper show good correlation to the measured results, which are also displayed within the paper, it has been shown in

preceding sections of this report that autoregressive models taken as a whole can find it “*difficult to describe the statistical features of solar radiation*”[50].Furthermore the output resolution of the values produced by the model have a maximum resolution of 15 minutes, although this time step will display a large majority of the solar variability experienced it will inevitably mitigate some of the short term variability. The impact of this would be minimal when taken into consideration over a singular day but over a longer period, such as a month or year; this could have an impact on reactive and dependent systems.

The paper by Walkenhorst, O et al [23] is titled “*Dynamic annual daylight simulation based on one –hour and one-minute means of irradiance*” it aims to investigate the influence of short term solar variability on daylight availability within a building, it does so by utilizing building modelling software called DAYSIM which is a RADIANCE based dynamic daylight simulation tool, however due to the lack of available high resolution data it encounters problems and thus it uses an adapted version of the Skartveit and Olseth model to generate high resolution data of 1-minute resolution from hourly means, it deduces that the utilization on high resolution data reduces the underestimation of artificial lighting demand over the year from 27% to 8% respectively. This report typifies how high resolution data can be used outside of the field of solar power and photovoltaic’s but , as previously described the use of autoregressive models can provide “*pitfalls*”[50]. This is one of the main reasons why the authors model uses a Markov chain approach which utilizes a transition probability matrix.

The final journal and model which attempts the generation of high resolution irradiance data is a report by Richardson. I and Thomson. M [59] the irradiance model contained within thus report is only one part of what is a bigger model including an electricity demand model, occupancy model and PV model.

The irradiance model contained within is intrinsically linked to both the electricity demand model taking into account lighting and the PV model from which energy output is calculated thus an accurate portrayal of how well the photovoltaic panel matches the energy usage of the dwelling is gained.

The irradiance model exploited by Richardson and Thomson does not include any means of inputting base irradiance data , thus the high resolution data generated is totally synthetic and is not intended to represent what occurred but merely what could

have occur based on previous trends, to achieve this a first order Markov-chain is used in conjunction with a transition probability matrix.

The model described within the authors report aims to use a similar technique in that the use of a Markov chain and transition probability matrix will be utilized as opposed to any other type of approach (i.e. an autoregressive model).

This however will be accompanied by the ability for the user to input hourly values which will be used as a basis for the stochastic generation of high resolution global horizontal radiation data, furthermore the transition probability matrixes utilized by the authors model will be expanded extensively so as to include multiple transition probability matrix representing monthly periods. This approach should allow for the generation of high resolution data from hourly means which is simple and robust yet yields good accuracy in correlation to measured results.

4.0 - Model Construction

In this section of the report the methodology, equations, procedures and techniques which are used in the construction of the model discussed above are examined and explained. The spreadsheet model which is contained within this report consists of a number of worksheets and Visual basic codes which are linked within Microsoft Excel to provide an easy to use method of stochastically generating high resolution solar data from hourly means, a general overview of the entire process is shown in figure 15 below. Each of the processes listed in figure 15 requires greater explanation as to their operation and construction; these will be explained in subsequent sections of the report the layout of which will follow in the same order as they are show in figure 15 below.

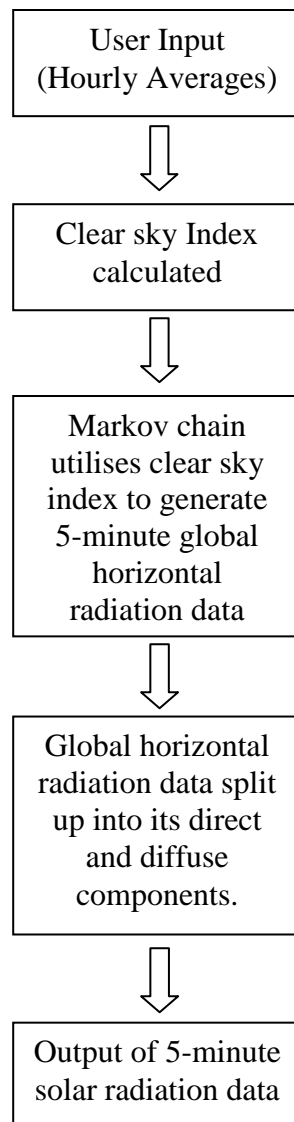


Figure 15: Overview of generation procedure

4.1 - User input

Before any calculations or operations are carried out a number of key parameters need to be set by the user, these parameters are used consistently throughout all the subsequent calculations. The information required from the user relates mostly around the entering of the global horizontal radiation hourly means which are used as a basis for the model to generate higher resolution data, the location of the site in question (i.e. latitude and longitude) and the start date and end date of the hourly data is also required as the model can generate up to one month's worth of data and the day of the year needs to be ascertained for each day of data entered to undertake the calculations required. Without this information the model will obviously not operate as desired. A further section located on the user interface allows for the changing of the start and end date of summer, this is used by subsequent sections of the model to correct time variations brought about by the change to British summer time (BST), however these should not need to be changed from the default values for operation within the UK. The final section allows for the input of average temperatures and relative humidity's for each month for the location in question this will increase the accuracy of the model for that location. Figure 16 shows a screenshot of the user interface of the model.

The screenshot shows the Microsoft Excel interface for the Synthetic High Resolution Solar Model. The spreadsheet is divided into several sections for user input and data output. Callout boxes highlight the following elements:

- Start date and End date:** Callout box pointing to cells B2 and C2, stating "Up to one month's worth of data can be entered".
- Location (Latitude & Longitude):** Callout box pointing to cells B3 and C3.
- Average hourly global horizontal radiation:** Callout box pointing to the 'Global Horizontal Hourly Solar data' table (rows 10-24).
- Monthly temperature and relative humidity:** Callout box pointing to the 'Avg temp' and 'Humidity' columns (rows 4-12).
- Click buttons that run VBA code:** Callout box pointing to the 'Row count/Clear sky Generation', 'Calculate Clear sky averages', and 'Generate 5-minute Solar radiation data' buttons.

Start date	End date	Z (elevation at site) from sea level (m)	Month	Avg temp	Humidity
01/12/2010	31/12/2010	68	1	3.71	49.4
57.43	-0.963	97	2	6.41	44.6
	304		4	9.6	44
			5	11	43
			6	16	43
			7	18	43
			8	16	43
			9	14	43
			10	11	43
			11	5.7	41
			12	0.85	57.5

Figure 16: User interface

4.2 - Generation of Clear sky index

Part of the user input consists of hourly averages of measured global horizontal radiation, previous sections of this report have explored the problems related to use hourly averages as opposed to higher resolution time series data, thus this model has been created to easily generate high resolution solar radiation data from these hourly means. This process requires a number of steps which were shown in figure 15, the first of these steps is the generation of clear sky index (**K_c**) as proposed by Skartveit, A. & Olseth, J. A [18].

The generation of clear sky index itself can be split into two parts, firstly the calculation of the Clear sky radiation (**R_{so}**) is needed, this being the radiation that would be experienced in a given location if it was a totally clear day (i.e. no cloud cover) this is calculated through the use of a series of equations which are further explored in section 4.2.2. The second step is to calculate the Clear sky index (**K_c**), the clear sky index is a simple method used to take account of the attenuation of the solar radiation due to cloud cover, this negates the need for the use of extremely complex meteorological calculations, it also allows for the analysis and generation of trends in the transition between different degrees of cloud cover which is key to this project. The clear sky index is further discussed in section 4.2.3 below.

4.2.1 - Clear Sky radiation (R_{so})

One of the most critical parts of the algorithm constructed by the author is the calculations related to the clear sky radiation or the clear sky model, as stated previously the clear sky model is a mathematical derivation used to determine the global horizontal radiation (W/m^2) that would occur on any given day at any location on a clear, cloudless day.

The output of this model is used in the determined the hourly Clear sky index (**K_c**), these hourly clear sky indexes are in turn used as the input for subsequent sections of the model which determine the characteristics of the cloud cover of the proceeding hour, a detailed overview of this procedure is discussed at length later in the report however as the clear sky model is such a key component to the functionality of the model it is important to explore the method and equations used to calculate the clear sky global horizontal radiation(**R_{so}**).

Currently there are a variety of methods which can be used to generate estimations of clear sky radiation, each of these methods share some crucial equations relating to the position of the sun and solar angles however the actual calculation of the clear sky global horizontal radiation is achieved in a different manor for each, in order to ensure that the best method was used three separate methods for calculating the clear sky radiation were tested and then subsequently checked against measured data. The three methods compared were the Ashrae method [60], a method proposed by Duffie and Beckman [61] and a method proposed by the Environmental and water resource institute (EWRI) of the American society of Civil engineers (ASCE) [62 – 63]. Although all three methods gave good results for most of the year , on some occasions the Ashrae and Duffie and Beckman methods underestimated the clear sky radiation this is most likely to do with their simple procedure and the fact they do not take into account temperature and relative humidity of the location in question. As such it was determined that the method proposed by the American society of civil engineers (ASCE) consistently gave the best results. This can be clearly seen in figure 17 below which shows a comparison of all three methods tested to measured data recorded in Belfast on 1st of May 2011. Figure 18 shows the ASCE method only.

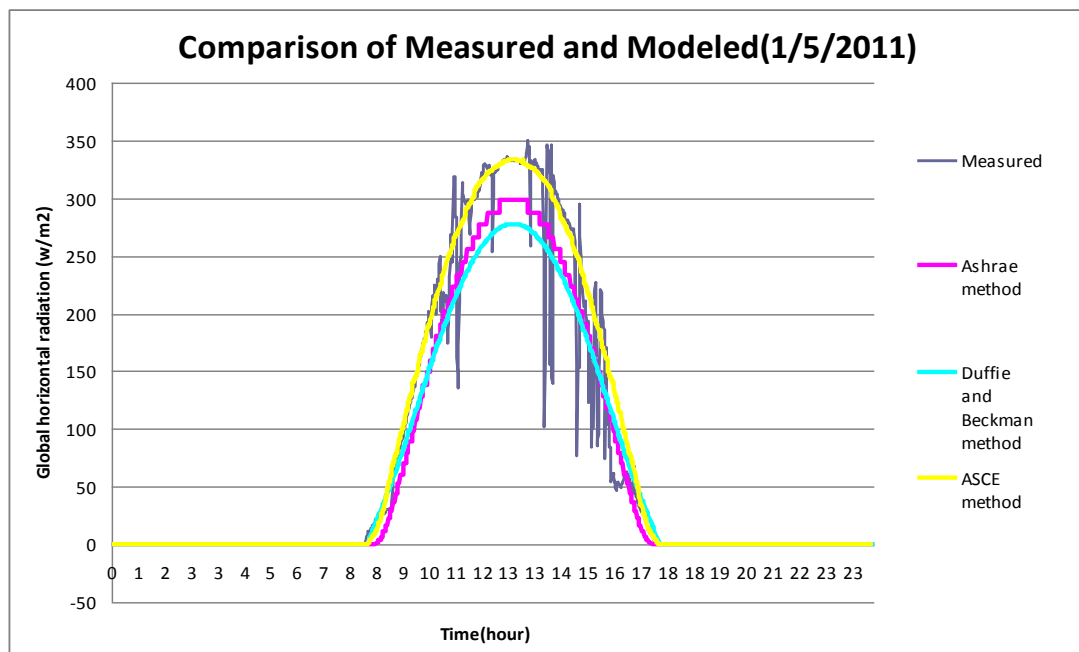


Figure 17: Comparison of measured and modelled data

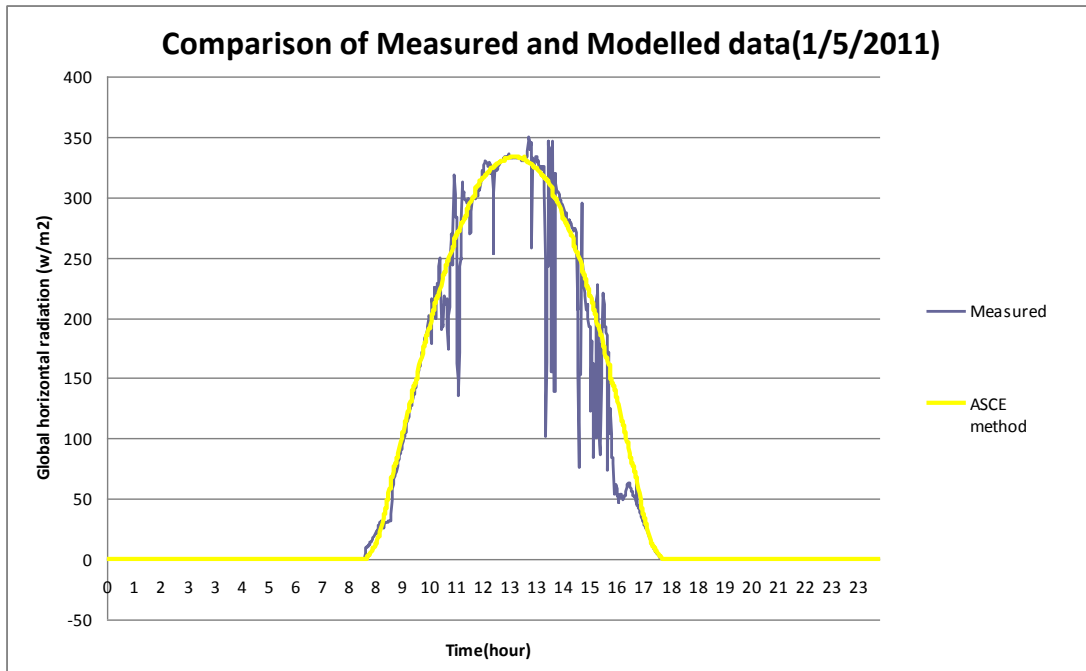


Figure 18: Comparison of measured and modelled data (ASCE) method only

Henceforth when the clear sky method/radiation is referred to the author will be referring the ASCE method as opposed to the Ashrae or Duffie and Beckman method. As the ASCE method was the only method utilised after the initial analysis it will be the only method explained in full for this report as it would be impractical and unnecessary to explain all three methods.

4.2.2 - Clear Sky radiation model equations and operation

The clear sky radiation model is used to determine the solar radiation that would occur at any given site if it was a cloudless clear day this value however does not remain constant, the value of solar radiation hitting any location will vary dependent on the location of the sun with respect to the site in question, this value further changes over time as the earth rotates, not only about the sun but about its own axis. In order to determine these values knowledge of solar angles and the geometry of the sun in relation to the earth is needed. The first thing that is required is the location of the site in question where Latitude (L_a) is the angle north or south of the equator and Longitude (L_o) is the angle east or west of the Prime meridian Greenwich, this information must be given by the user in the authors model, also needed from the user is the date which in turn is used to determine the day of the year (DOY) where 1st of January has a DOY number of 1 and 31st of December has a DOY number of 365 this information is used in a number of equations that are explored in this section. After these values have been entered key information can be calculated which will

give the solar position and angle which are then utilized in determining the clear sky radiation (**R_{so}**) on the horizontal plain. The information required to achieve this is the declination (**δ**), Hour angle (**h**), solar altitude (**β**), Solar azimuth (**γ_s**) and Zenith angle (**θ_z**) the equations used in these calculations are shown below in detail.

Declination (δ) - The angular position of the location in question, north or south of the celestial equator is called the declination. It ranges from -23.5 to +23.5 as the earth rotates about the sun and its own axis. There are a number of ways to calculate declination with varying degrees of accuracy the equation used for this report is shown in equation 2 below.

$$\delta = 0.006918 - 0.399912 \cos \Pi + 0.070257 \sin \Pi - 0.006758 \cos 2\Pi + 0.000907 \sin 2\Pi - 0.002697 \cos 3\Pi + 0.00148 \sin 3\Pi \quad (\text{Equation 2 [64]})$$

Where **Π** is the day angle calculated by: $\frac{2\pi * (DOY - 1)}{365}$ (Equation 3 [64])

And **DOY** is an abbreviation for day of the year.

Hour angle (h) - The angular position of the sun east or west as it moves across the sky from the local meridian is called the hour angle, at solar noon when the sun is directly overhead the hour angle is equal to 0°, each hour from the solar noon corresponds to an angular motion of the sun of 15°. The hour angle is calculated by firstly calculating the equation of time (EOT) (equation 4) this formula is an empirical relationship which corrects the eccentricity experienced due to the way the earth orbits the sun and earths axial tilt, sequentially a time correction factor (TC) (equation 6) is calculated thus taking into account of the variations in local solar time (LST) over the year. After the time correction factor has been calculated the local solar time (LST)(Equation 7) itself can then be calculated , finally the hour angle is derived from use of equation 8[65] these equations are shown below.

$$EOT = 9.87 \sin(2B) - 7.53 \cos(B) - 1.5 \sin(B) \quad (\text{Equation 4})$$

Where **EOT** is the equation of time

$$B = \frac{360}{365} (DOY - 81) \quad (\text{Equation 5})$$

$$TC = 4(Lo - LSTM) + EOT \quad (\text{Equation 6})$$

Where **TC** is time correction factor

Lo is Longitude

EOT is equation of time

LSTM is local standard time meridian

$$LST = LT + \frac{TC}{60} \quad (\text{Equation 7})$$

Where **LST** is Local solar time

LT is the local time

TC is the time correction factor calculated in equation 5

$$h = 15 \times (LST - 12) \quad (\text{Equation 8})$$

Where **h** is the hour angle

LST is the Local solar time calculated in equation 7.

Once the declination and hour angle has been calculated the solar altitude (β), solar azimuth (γ_s) and zenith angle (θ_z) may be calculated.

Solar altitude angle (β) – is the angle between the solar radiation and the horizontal as shown in figure 19, It is calculated by the use of equation 9.

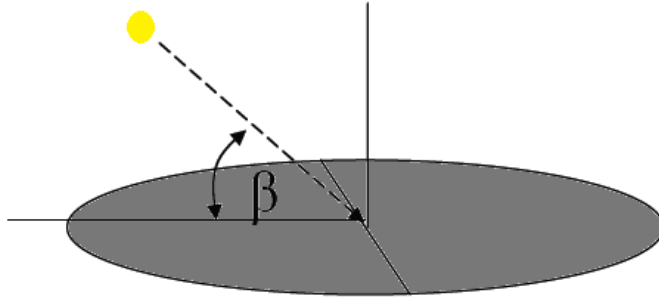


Figure 19: Solar altitude angle (β)

$$\beta = \text{Sin}^{-1}[\text{Cos}(La)\text{Cos}(h)\text{Cos}(\delta) + \text{Sin}(La)\text{Sin}(\delta)] \quad (\text{Equation 9})$$

Where **La** is latitude

h is hour angle

δ is declination

Solar azimuth (γ_s) – is the angle between the north and south and the solar radiation as shown in figure 20, it is determined by equation 10.

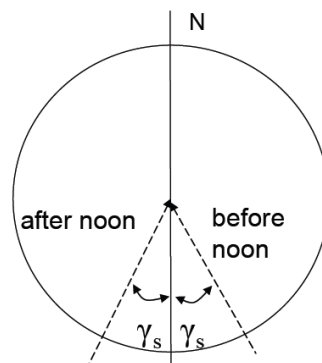


Figure 20: Solar Azimuth (γ_s)

$$\gamma_s = \frac{\sin(La) \cos(h) \cos(\delta) - \cos(La) \sin(\delta)}{\cos(\beta)} \quad (\text{Equation 10})$$

Where **La** is latitude

h is hour angle

δ is declination

and **β** is the solar altitude angle

Zenith angle (θ_z) – Is the angle between the solar radiation and the vertical as shown in figure 21. The zenith angle is calculated using the equation 11.

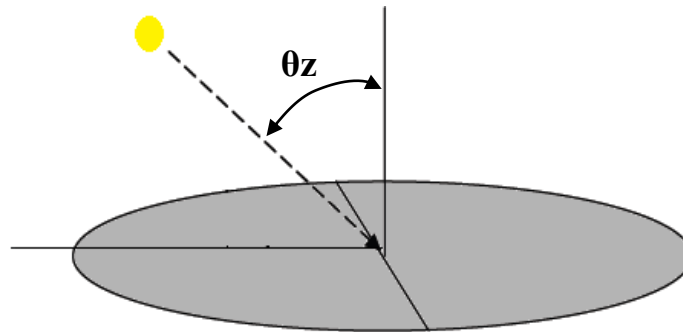


Figure 21: Solar Zenith angle (θ_z)

$$\theta_z = \cos^{-1}[\cos(La)\cos(h)\cos(\delta) + \sin(La)\sin(\delta)]$$

OR

$$90 - \beta$$

(Equation 11)

Where **La** is latitude

h is hour angle

δ is declination

The cosine of the above solar zenith angle (θ_z) is another key parameter which is utilised in a number of equations, used to calculate the clear sky radiation; it is calculated by the method shown in equation 12 below.

$$\cos \theta = \sin(La)\sin(\delta) + \cos(La)\cos(h)\cos(\delta) \quad (\text{Equation 12})$$

Clear sky radiation (**R_{so}**) - Through the use of the equations above, the position and angle of the sun can be determined at each time step this information is then used in the calculation of Clear sky radiation (**R_{so}**) using the method proposed by the American society of civil engineers as discussed previously [62].

$$R_{so} = (K_B + K_D)Ra$$

(Equation 13)

Where:

K_B is Clearness index for the direct beam component of the solar radiation

K_D is the transmissibility index for the diffuse component of the solar radiation

R_a is the extraterrestrial radiation [W/m²]

The derivation of **R_a**, **K_B** and **K_D** are further explored below.

Extraterrestrial radiation (**R_a**) - is the radiation that would be experienced if there were no atmosphere present and is calculated by:

$$Ra = \frac{G_{sc} \cos\theta}{d^2}$$

(Equation 14)

Where :

G_{sc} is the solar constant of 1367 W/m²

Cosθ is the cosine of the solar zenith as calculated in equation 12

And **d²** is a function of the day of the year (DOY) as specified by Duffie and Beckman [61] and is calculated using equation 15.

$$d^2 = \frac{1}{1 + 0.033 \cos\left(\frac{DOY 2\pi}{365}\right)}$$

(Equation 15)

Clearness index for the direct beam component (K_B) is calculated by:

$$K_B = 0.98 \exp \left[\frac{-0.00146 P}{K_t \sin \beta} - 0.075 \left(\frac{W}{\sin \beta} \right)^{0.4} \right]$$

(Equation 16)

Where:

Kt is the turbidity coefficient where $0 < K_t < 1$, however for a clear day $K_t = 1$

W is Perceptible water [mm], calculated by equation 17

P is Atmospheric pressure [kPa] and is calculated by equation 19

And **Sinβ** is the sun angle β above the horizontal

Through the use of equations 17 – 21, K_B takes into account the elevation of the site, sun angle, water vapour and atmospheric pressure and how these factors influence both the absorption and scattering of solar radiation.

The derivation of **W**, **P** and **Sinβ** are further explored below.

Perceptible water (W) - is calculated by equation 17 and determines the volume of perceptible water in the atmosphere in millimetres [mm]:

$$W = 0.14 e^a P + 2.1 \quad (\text{Equation 17})$$

Where e^a is the Actual vapour pressure [kPa]

$$e^a = e^o \frac{Rh}{100} \quad (\text{Equation 18})$$

Rh is relative humidity

And e^o is a saturated vapour pressure coefficient, which analysis the ability of the air to hold water vapour.

$$e^o = 0.618 \exp \left(\frac{17.27 T}{T + 237.5} \right) \quad (\text{Equation 19})$$

Where **T** is temperature

Atmospheric pressure (P) - is calculated by equation 20 and determines the mean atmospheric pressure at the site utilising the site elevation and a simplified version of the Universal gas law [62].

$$P = 101.3 \left(\frac{293 - 0.0065 Z}{293} \right)^{5.26} \quad (\text{Equation 20})$$

Where

Z is site elevation above sea level (m)

Sin β - is the Sin of the sun angle β above the horizontal calculated by use of equation 9

Transmissibility index for the diffuse component (K_D) - is a function of the clearness index of the direct beam component K_B and is calculated using equations 21 & 22 dependent on the value of K_B .

$$K_D = 0.35 - 0.36K_B \quad \text{For } K_B > 0.15 \quad (\text{Equation 21})$$

$$K_D = 0.18 - 0.82K_B \quad \text{For } K_B < 0.15 \quad (\text{Equation 22})$$

4.2.3 - Clear sky index calculations

Once the clear sky radiation R_{so} has been calculated the clear sky index can be generated by the use of equation 24 below. As stated before the clear sky index is a useful method which is utilised to identify the amount of cloud cover in a location and alleviates the need for more complex meteorological calculations. In this model the clear sky index ranges from 0 to 1, with 1 representing a clear day with no cloud cover and 0 representing times where no solar radiation is incident on the site (i.e. night-time), a clear sky index of 0.5 for example would represent a day where only half the global radiation that could have potentially reached the site in question actually did due to attenuation caused by the clouds. The clear sky index (K_c) is defined as the ratio of measured global horizontal radiation and the calculated Clear sky radiation on the horizontal plain.

$$K_c = \frac{I_{gh}}{R_{so}} \quad (\text{Equation 24})$$

Where:

I_{gh} is the measured global horizontal radiation [W/m^2]

R_{so} is the clear sky radiation [W/m^2]

For the model which accompanies this report the hourly clear sky index is calculated using the hourly means of measured data, provided by the user and the hourly mean of the clear sky radiation which is calculated through the use of equations 2 – 23. These hourly clearness index are used as the input into the third section of the model as shown in figure 15. In this section the hourly clear sky index generated by the process explained above are used in combination with a Markov chain process and transition probability matrix in the generation of 5-minute global horizontal radiation, However the initial construction of the transition probability matrixes also requires the use of the clear sky index although at a higher temporal resolution, how these processes operate and their initial construction is explained in detail in section 4.3 below.

4.3 - Generation of 5-minute resolution Global horizontal radiation data

As has been discussed previously the process used to generate 5 minute resolution global horizontal data from hourly means utilizes a method which uses transition probability matrixes in combination with a first order Markov chain process. In order to make the model operational the transition probability matrixes first had to be constructed, to achieve this the clear sky index explained in section 4.2.3 was used in combination with measured 5 minute resolution global horizontal radiation data which was provided by the meteorological department of the University of Reading, thus meaning that unlike the process described in section 4.2.3 the clear sky index was generated at 5 minute time steps as opposed to hourly. This section aims to explore how the transition probability matrix were constructed and operate, and how the Markov chain process is used to generate the 5 minute resolution data from hourly means.

4.3.1 - Transition probability matrix (TPM) construction

In order to generate stochastic solar radiation data a transition probability matrix has to be constructed. A transition probability matrix is a method which is used to take into account the probability of transition from one possible state to any other possible state, in this model the transition probability matrix takes account of the likelihood of transition between the clear sky index's at 5 minute time intervals which are generated using the process described in section 4.2.3 in conjunction with 5 minute resolution measured global horizontal radiation data as shown in figure 22.

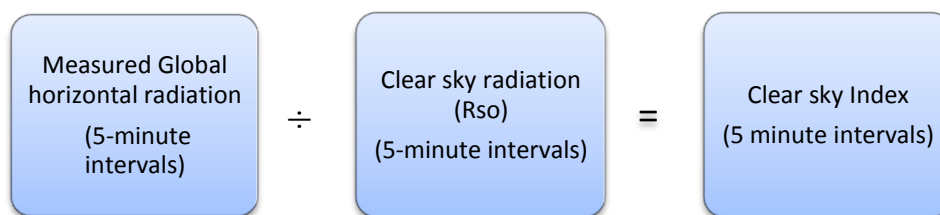


Figure 22: *Clear sky Index (5 minute intervals)*

The global horizontal radiation observations used for this process were obtained by the automatic measurement system of the Department of Meteorology at University of Reading using a Kipp & Zonen CM11 Pyranometer, in total 4 years of data was used to build the transition probability matrixes resulting in a long time series of data, this long time series of 420480 entries was then split up into 12 separate data series corresponding to the months of the year with each separate data series consisting of between 32256 to 35712 values dependent on the month. This approach was used as it is not expected that every month would have the same transition probabilities therefore the aim of this approach was to take account of any changes in probability of transition that would occur dependent on the month and there different characteristics in weather patterns which ultimately has an effect on cloud cover. The use of 4 years worth of data is also important in an aim to average out the likelihood of a single year with abnormal weather patterns (i.e. a very hot summer or cold winter) overly effecting the transition probabilities between states thus a more standard picture of the transition probabilities is gained using multiple years.

After the long time series of data has been split into monthly data sets, the probability of transition is then determined by counting the number of times a value is followed by another number and dividing by the number of times the value appears in the time series of data. For instance the probability that a clear sky index of 1 will remain at 1 (i.e. clear sky) at the next time interval (i.e. 5 minutes) is determined by counting the number of times a clear sky index of 1 is directly followed in the data series by another clear sky index of 1 and then dividing by the number of times the clear sky index of 1 appears throughout the entire data series thus determining the likelihood of transition, each transition probability matrix is constructed with a range of 0.01 to 1 along the horizontal and vertical axis of the matrix. The probability of transition between the values on the vertical axis and the values on the horizontal axis are then determined, thus generating a transition probability matrix were the probability of transition between each possible clear sky index is quantified. A shortened example of the transition probability matrix for the month of March can be seen in figure 23 below, as can be seen the highest probability is either for the clear sky index to remain the same or for it to increase or decrease only slightly, giving the TPM on the whole a trend were the largest probabilities occur diagonal to one another, towards the clear

sky end of the TPM this trend is also evident however there is also a greater probability that the clear sky index at the next time step will be 1 (clear sky).

		Transition probability matrix (TPM)											
		Next clear sky index											
		0.01	0.02	0.03	0.04	0.05	-	0.95	0.96	0.97	0.98	0.99	1
Present Clear sky index	0.01	0	0	0	0	0	-	0	0	0	0	0	0
	0.02	0	0.666667	0.222222	0	0	-	0	0	0	0	0	0
	0.03	0.166667	0	0.166667	0.5	0	-	0	0	0	0	0	0
	0.04	0	0.076923	0.076923	0.230769	0.230769	-	0	0	0	0	0	0
	0.05	0	0	0.058824	0.117647	0.294118	-	0	0	0	0	0	0
	0.06	0	0	0.026316	0	0.105263	-	0	0	0	0	0	0
	0.07	0	0.016949	0	0.050847	0.016949	-	0	0	0	0	0	0
	0.08	0	0	0	0	0.012048	-	0	0	0	0	0	0
	0.09	0	0	0	0.008621	0	-	0	0	0	0	0	0
	0.1	0	0	0	0	0	-	0	0	0	0	0	0
	-	-	-	-	-	-	-	-	-	-	-	-	-
	0.9	0	0	0	0	0	-	0.035176	0.015075	0.025126	0.005025	0.01005	0.075377
	0.91	0	0	0	0	0	-	0.034091	0.015152	0.018939	0.022727	0.015152	0.102273
	0.92	0	0	0	0	0	-	0.052174	0.026087	0.043478	0.004348	0.017391	0.121739
	0.93	0	0	0	0	0	-	0.034364	0.030928	0.020619	0.024055	0.013746	0.099656
	0.94	0	0	0	0	0	-	0.131313	0.050505	0.030303	0.026936	0.016835	0.124579
	0.95	0	0	0	0	0	-	0.281899	0.145401	0.068249	0.017804	0.023739	0.130564
	0.96	0	0	0	0	0	-	0.102273	0.267045	0.173295	0.056818	0.042614	0.130682
	0.97	0	0	0	0	0	-	0.030233	0.104651	0.327907	0.155814	0.05814	0.165116
	0.98	0	0	0	0	0	-	0.013514	0.040541	0.117117	0.36036	0.184685	0.130631
0.99	0	0	0	0	0	-	0.016064	0.022088	0.056225	0.118474	0.381526	0.2751	
1	0	0	0	0	0	-	0.009041	0.009923	0.011466	0.014553	0.026461	0.741566	

Figure 23: Transition probability matrix example

4.3.2 - First order Markov Chain process

The transition probability matrix's which are constructed using the method discussed above are used in conjunction with a first order Markov chain so as to generate 5 minute resolution data. A Markov chain is a process which is said to have the "Markov property", that being a memory-less property were the present state is only defined by a specified degree of the past. The concept of a first order Markov chain (as opposed to a second, third... order) is a technique were the current state is determined based only on the previous state and the probability of transition, which in the authors model are held within the transition probability matrix explained above. Whereas a second and third order Markov chain would determine the current state based on the previous two or three states respectively. The first order Markov chain process has been widely and successfully used to generate stochastic data in the past [59] and generally provides good correlations to the distributions which are

experienced in measured data. This was discussed in the literature review where it was shown that a Markov chain process provide a “*simple yet effective simulation device*” [50] for the generation of solar radiation data.

The particular process used for the authors model utilises a random number (between 0 and 1), together with the above mentioned transition probability matrix and the current clear sky value in its operation and generation of data, this type of first order Markov chain is more commonly referred to as a Markov Chain Monte Carlo (MCMC). The detailed operation of this process is explained below in section 4.3.3.

4.3.3 - The generation of 5-minute resolution global horizontal radiation data

The model discussed within this report is constructed using Microsoft Excel, and visual basic for applications (VBA) for excel is used extensively throughout, The method within this project builds upon the initial work of Richardson, I & Thomson, M [59].

The VBA code used by the model is freely available to examine within the model itself but can also be viewed in the appendix section of the report (Appendix 1 - 3). Figure 24 below shows the operational process of the model, and the steps that are involved, the model in this flow chart is colour coded with each colour representing a different part of the model **Blue** is represented as the user input, **Orange** is related to the calculation of clear sky radiation and Hourly clear sky index which is used as the input for the third section of the model which is represented in **green** were the hourly input is turned into a 5 minute resolution data through use of the Markov chain process and transition probability matrix.

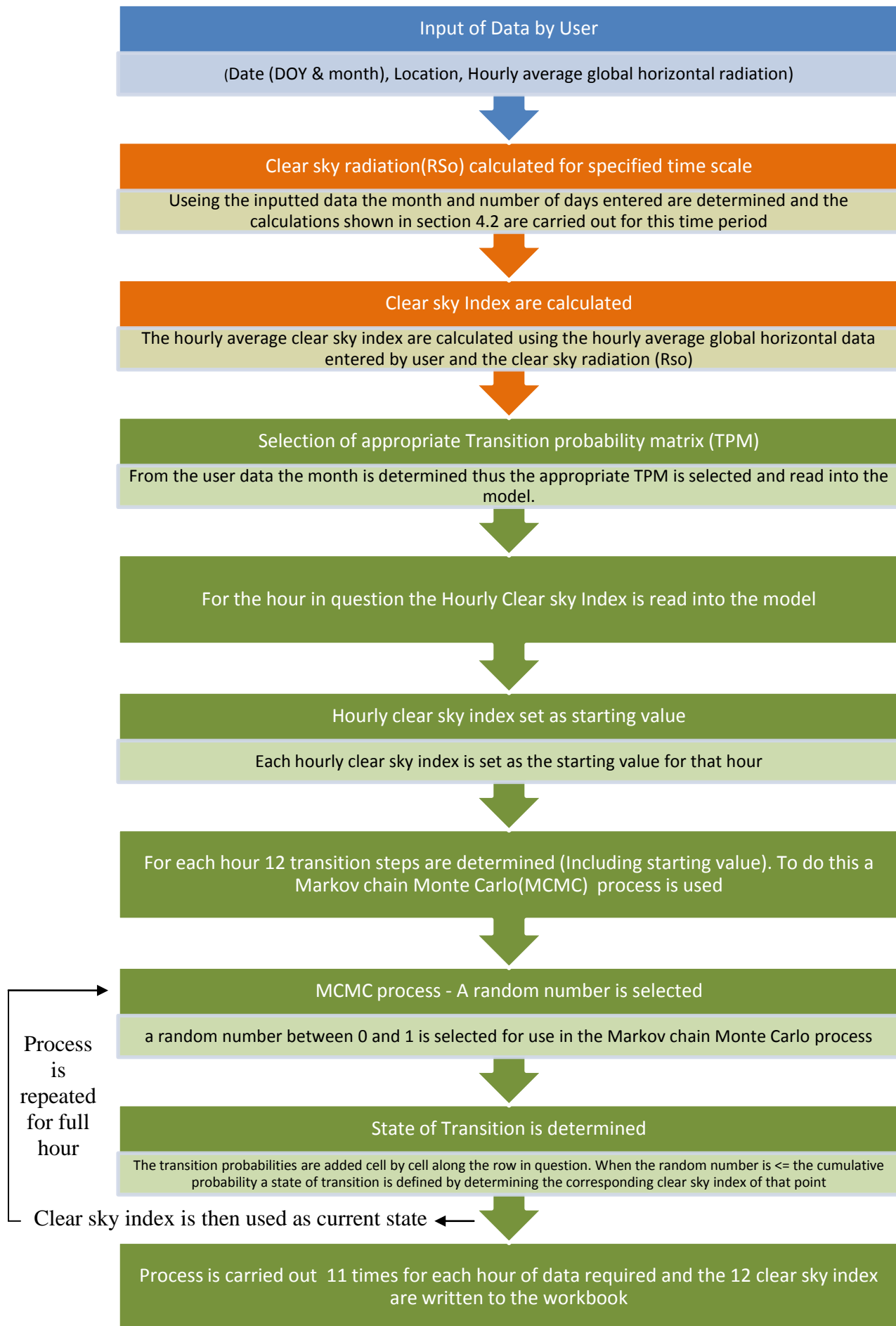


Figure 24: Operational process of model

The user input, calculation of clear sky radiation (R_{so}) and calculation of clear sky index (blue and orange sections) have been discussed at length in previous sections of this report. The operation of the green section however which relates to how the hourly clear sky index is turned into 5 minute resolution data has only briefly been mention and requires further explanation.

The first steps of the code highlighted by the green section are fairly self explanatory from figure 24 and have been discussed above in section 4.3.1: “transition probability matrix construction” in that the transition probability matrix is selected dependent on the month, therefore from the user input the month is determined and the corresponding TPM is read into the model were it is stored as an array. Taking the process one hour at a time the model progresses to the next step were the hourly clear sky index that have been generated in the previous sections of the model using the calculations and process described in section 4.2 are read into the model, every hour clear sky index calculated is used as the starting position from where the model generates 11 further clear sky index this therefore means that each hour will have in total 12 clear sky index (i.e. one every 5 minutes of the hour).

To achieve this a Markov Chain Monte Carlo process is used, this relies on the selection of a random number, and the use of the transition probability matrix. Once a random number (between 0 & 1) has been generated the code cycles through the row which corresponds to the previous 5-minute clear sky index or if at the start of the hour the starting clear sky index taken from the section above, as it does this it adds up the probabilities, when the cumulative probabilities is greater than or equal to the random number generated at the start of the process the corresponding clear sky value for that column is used as the current time step clear sky index, this is written to the sheet and the process continues and repeats until all values for that hour have been generated , Once that hour is complete the process is repeated for each subsequent hour until all hours have been processed. Therefore at the end of this process all hours initially entered by the user will have 12 clear sky indexes generated.

To calculate the Global horizontal radiation at each time step the generated clear sky indexes at 5 minute intervals are multiplied by the clear sky radiation also calculated for 5 minute intervals through the equations shown in section 4.2 to give global horizontal radiation where the attenuation of solar radiation due to cloud cover has been taken account of by the clear sky index, and the input hourly data has been used as a basis for the process.



Figure 25: Global horizontal radiation calculation procedure

4.4 - Splitting Global horizontal radiation into direct and diffuse components

The global horizontal radiation output from the above process is useful in a number of applications, and resembles the type of readings that would be measured by a Pyranometer. Global horizontal radiation however is made up of two constituents these being direct beam radiation (I_b), the portion of the radiation direct from the sun and diffuse radiation (I_d) the portion of global radiation which is scattered in the atmosphere by clouds and particulates such that .

$$I_t = I_b + I_d \left(\frac{W}{m^2} \right)$$

Where:

I_t = Global radiation

I_b = Direct beam radiation

I_d = Diffuse radiation

For other applications, such as calculations involving photovoltaic systems it is useful if the direct and diffuse components of the horizontal radiation are known, as this is a common problem, in that the most common type of measured data is global horizontal radiation and there is a requirement to know the ratios of direct and diffuse radiation which make up this global radiation there have been a number of studies which have developed equations which aim to split up Global radiation into its direct and diffuse

components. As each of these studies relies on measured data from pyranometers and shaded pyranometers from a variety of locations the equations derived are in some ways dependent on the characteristics of the site the measured data was collected, as such it was decided to explore a number of these methods and choose the most accurate for the UK.

In total three methods were selected for analysis these being the methods reported in Erbs et al [57], Orgil and Holland [56] , and Louch et al[66] these methods calculate the direct and diffuse component in a slightly different manner to each other. However, all of these methods utilise a factor called the “Clearness index”, the clearness index is very similar to the clear sky index used extensively throughout this project and in some reports the name Clearness index is used in place of the Clear sky index however in this report there is a distinct difference between the two. The clearness index (K_t) is represented by equation 25 where extraterrestrial radiation R_a on the horizontal plain (I.e. the radiation on the horizontal plain as if there was no atmosphere affecting it) is used rather than the clear sky radiation R_{so} .

Where:

$$K_t = \frac{I_{gh}}{Ra} \quad (\text{Equation 25})$$

K_t is the Clearness index

I_{gh} is the global horizontal radiation [W/m^2]

R_a is the Extraterrestrial radiation calculated using equation 13 [W/m^2]

The three methods analysed are further explained below, the Erbs et al and Orgill and Holland correlations are similar in respect that both calculate the diffuse component and subtract this from the global radiation to calculate the direct component, the Louche correlation on the other hand calculates the direct component and then by subtracting this from the global radiation the diffuse component is derived.

Erbs et al Method

$$\frac{I_d}{I_{gh}} = \begin{cases} 1.0 - 0.09Kt & \text{For } Kt < 0.22 \\ 0.9511 - 0.1604Kt + 4.388Kt^2 - 16.638Kt^3 + 12.336Kt^4 & \text{For } 0.22 < Kt < 0.80 \\ 0.165 & \text{For } Kt > 0.80 \end{cases} \quad (\text{Equation 26})$$

Orgill and Holland Method

$$\frac{I_d}{I_{gh}} = \begin{cases} 1.0 - 0.249Kt & \text{For } Kt < 0.35 \\ 1.55 - 1.84Kt & \text{For } 0.35 < Kt < 0.75 \\ 0.177 & \text{For } Kt > 0.75 \end{cases} \quad (\text{Equation 27})$$

Where:

I_d is the diffuse radiation,

I_{gh} the Global horizontal radiation

And **Kt** is the clearness index.

Louche Method

The Louche method unlike the Erbs and Orgill and Holland methods calculates the direct beam component as opposed to the diffuse component. This is done through the use of equation 28 & 29

$$Kb = -10.627Kt^5 + 15.307Kt^4 - 5.205Kt^3 + 0.994Kt^2 - 0.059Kt + 0.002 \quad (\text{Equation 28})$$

The Direct beam radiation is then calculated by:

$$I_b = Kb \times Ra \quad (\text{Equation 29})$$

Where:

I_b is the Direct beam radiation on the horizontal plain [W/m²]

Ra is the Extraterrestrial radiation [W/m²]

For all three methods the direct beam normal radiation (I_{b_n}) can be calculated by use of equation 30 below this is useful for a number of applications and as such is provided within the model.

$$I_{b_n} = \frac{I_b}{\sin\beta}$$

Where:

(Equation 30)

I_b is direct beam radiation on the horizontal plain [W/m^2]

$\sin\beta$ is the sun angle β above the horizontal

After analysis it was decided that the most consistent best fit was achieved by the use of the Louche method, this was further verified by analysis carried out by Battles et al [67] which arrived at the same conclusion. However none of the three methods provided exact fits, this is mainly to do with the fact that the Direct and Diffuse components are derived from the global horizontal radiation this means that any peaks and troughs which are experienced in the global radiation are propagated through to both the diffuse and direct radiation, in measured data anomalies caused by the passing of clouds may not necessarily have an impact on the measured diffuse component at such short time intervals, a comparison of measured data compared to the data calculated using each of the three methods is shown in figures 26 & 27 below. Taking into account that the Pyranometer used to measure the data can have an error of up to +- 10% [68] then it is shown that Louche method generally provides the best fit.

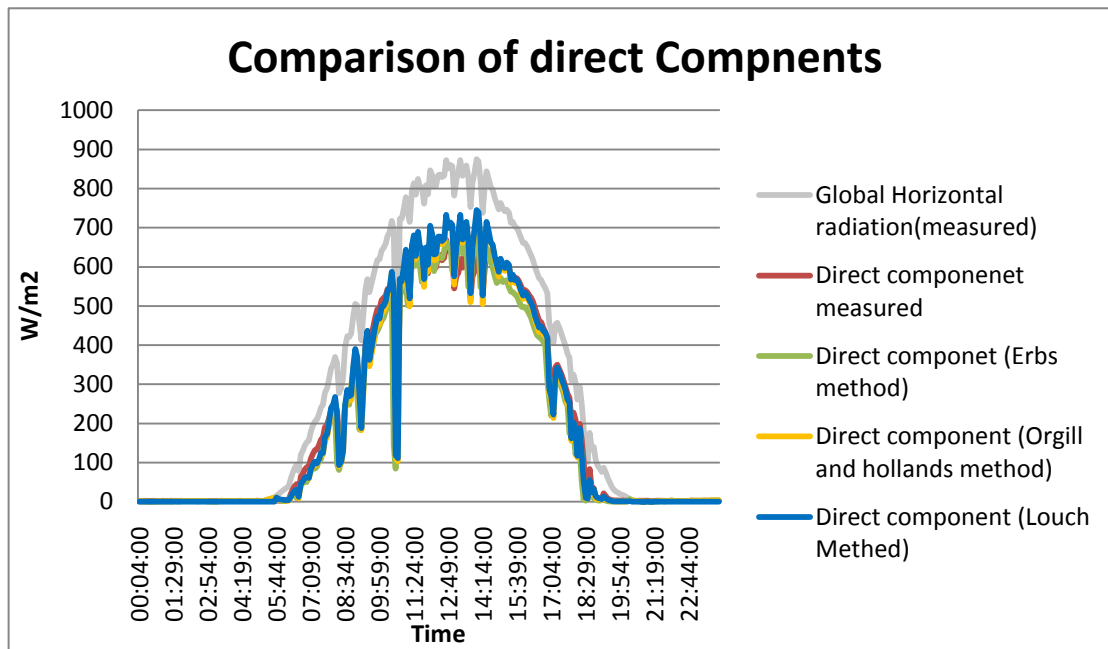


Figure 26: Comparison of calculated and measured direct components

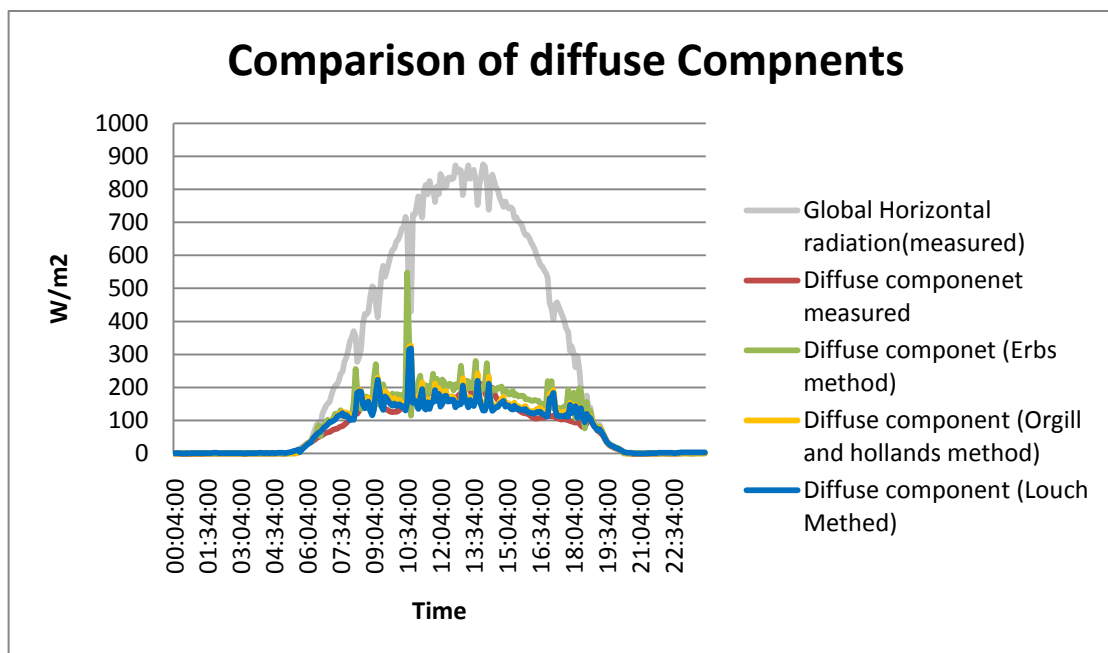


Figure 27: Comparison of calculated and measured diffuse components

5.0 - Model Validation

The model presented in this paper generates 5-minute resolution solar data based on hourly inputs using a stochastic process, therefore the output of the model will not exactly match what actually occurred at 5 minute time steps on that day as such it is difficult to gain an idea of accuracy of the model based on comparison between Generated data and measured data over short time periods such as individual days. For example dependent on the random numbers chosen by the program during the Markov chain Monte Carlo process the eventual outcome of the solar radiation data may well differ, this of course is one of the key points of the model as it is not merely a tool which replicates what happens on a given day, it is a tool which looks at the characteristics of the cloud cover on a particular hour and generates radiation values accordingly based on 4 years of data, thus mitigating any abnormalities in the weather for that particular period (i.e. Extreme summer and winters) . Figure 28 and 29 show two profiles of measured data compared to generated data, both these graphs have the same average hourly input values but it can be clearly seen that the output values can be very different due to the stochastic nature of the model, with figure 28 having generated data which matches the measured data well, and figure 29, having some over and under estimations of global horizontal radiation at points throughout the day.

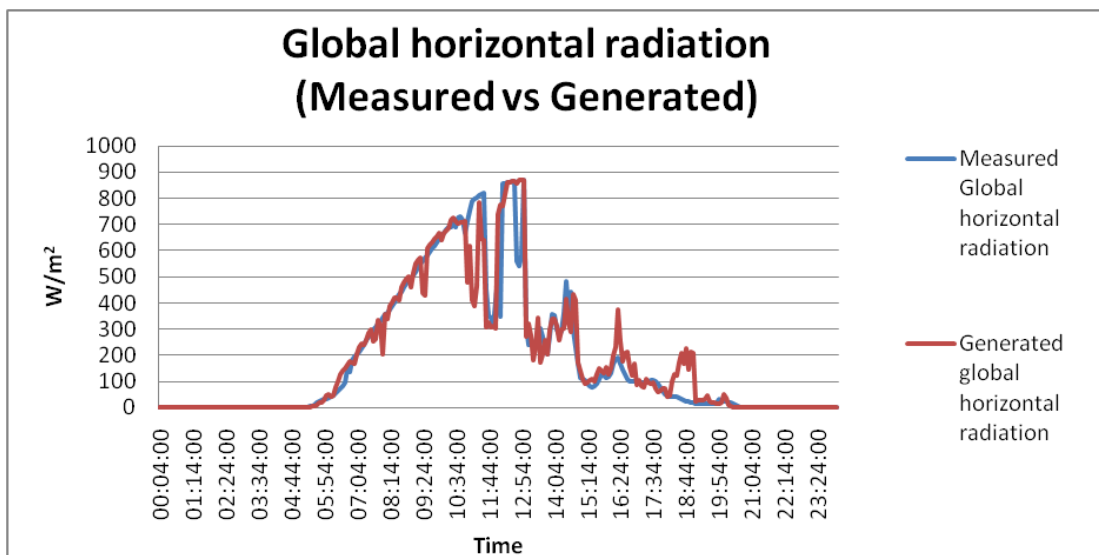


Figure 28: Stochastic generation difference (Good fit to measured data)

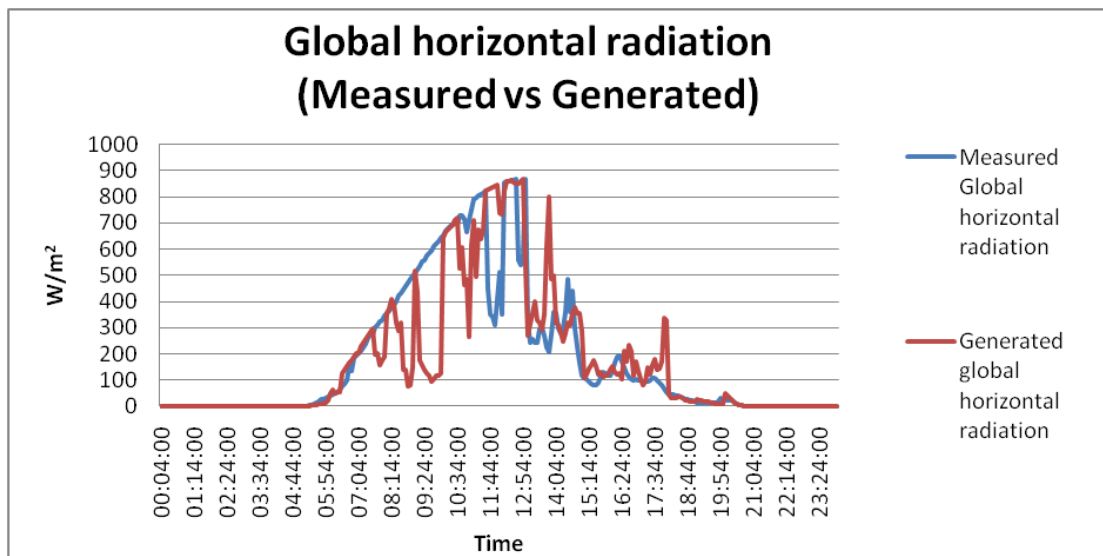


Figure 29: Stochastic generation difference
(Over an underestimation compared to measured data)

As individual days are not comparable directly the accuracy of the generated data has to be considered using a frequency based approach over a longer time span . As such it was decided to compare the generated data to measured data over the period of a year, the idea being that the over and under estimations which occur in individual days would average out over the year and thus the yearly frequency distribution of global horizontal radiation values should match.

5.1 – Method used and frequency distribution graphs

In order to validate the output of the model the frequency distribution of 1 years worth of generated 5-minute global horizontal radiation is compared, via a histogram to the measured 5-minute global horizontal radiation for the same year for the location of the University of Reading. This validation process was carried out for two separate years namely 2008 & 2010 , using two separate years for validation helped ensure that the results given by the model were consistent and not only accurate because of the input values used, the input values used in these validation runs are the hourly averages of the measured 5 minute global horizontal radiation for the years 2008 & 2010.

Throughout this comparison the measured results used were constrained by the calculation of the clear sky radiation **R_{so}** as sometimes for instance when the sun is appearing from or entering behind a cloud the global horizontal radiation measured

can be enhanced due to increased reflected radiation, in these cases the measured radiation is set equal to the clear sky radiation R_{so} at that time as currently the model presented within this paper is unable to take into account these cloud enhancements.

The graphs of frequency distribution are shown below in figures 30 & 31. The histograms presented show frequency of occurrence VS solar radiation in bins of 20 W/m² (i.e. each column represents a range of 20 W/m² for instance the first column is the amount of occurrences between 1 – 20 W/m², the second column is the amount of occurrences between 11-40 W/m² and so on).

Looking at figure 30 & 31 it can be seen that the modelled results generated provide a very accurate representation of the measured global horizontal radiation for both 2008 and 2010 with R^2 values 0.97 for each. The model results generated do not exactly match the measured results however it was not expected that they would, there is a tendency for the model to slightly underestimate the lower end of the spectrum however when analysing the data there is no definitive trend to this and it seems to be an artefact of the stochastic nature of the model.

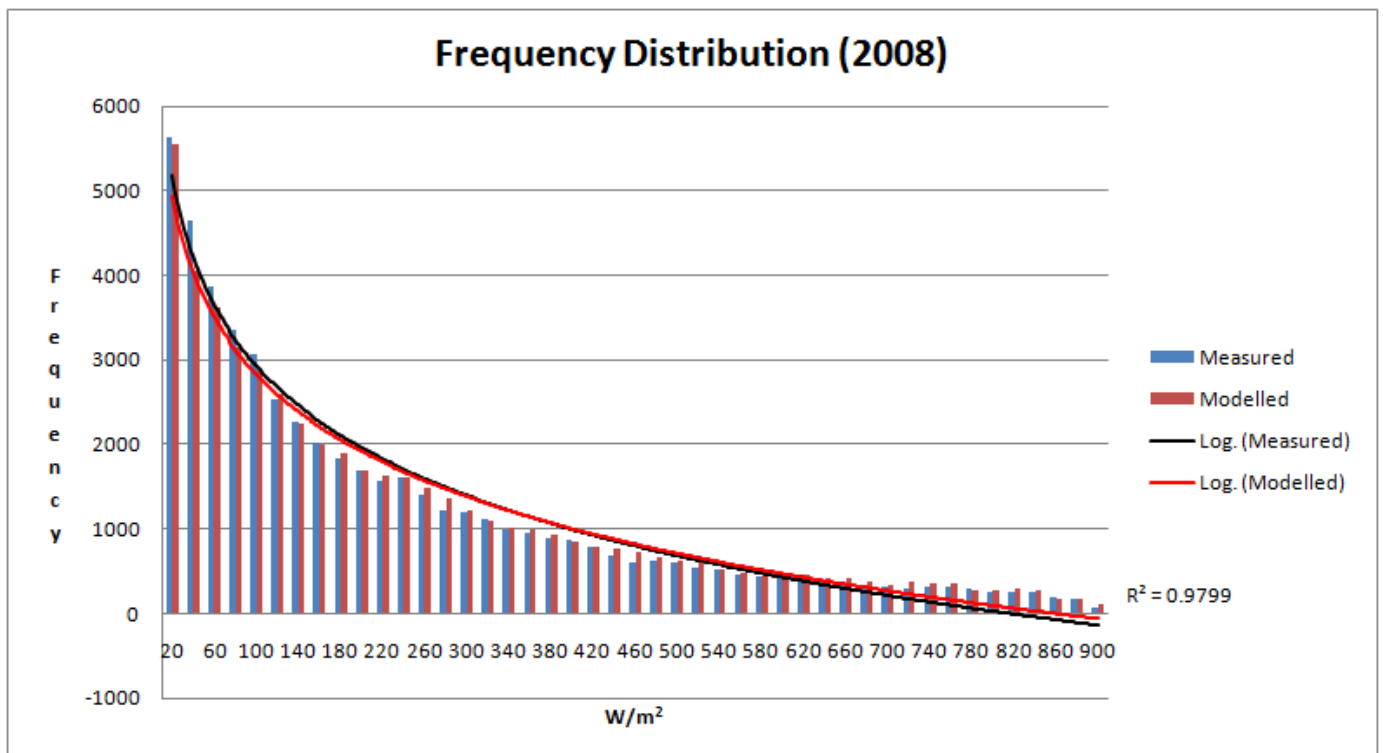


Figure 30: Frequency distribution for Reading (2008)

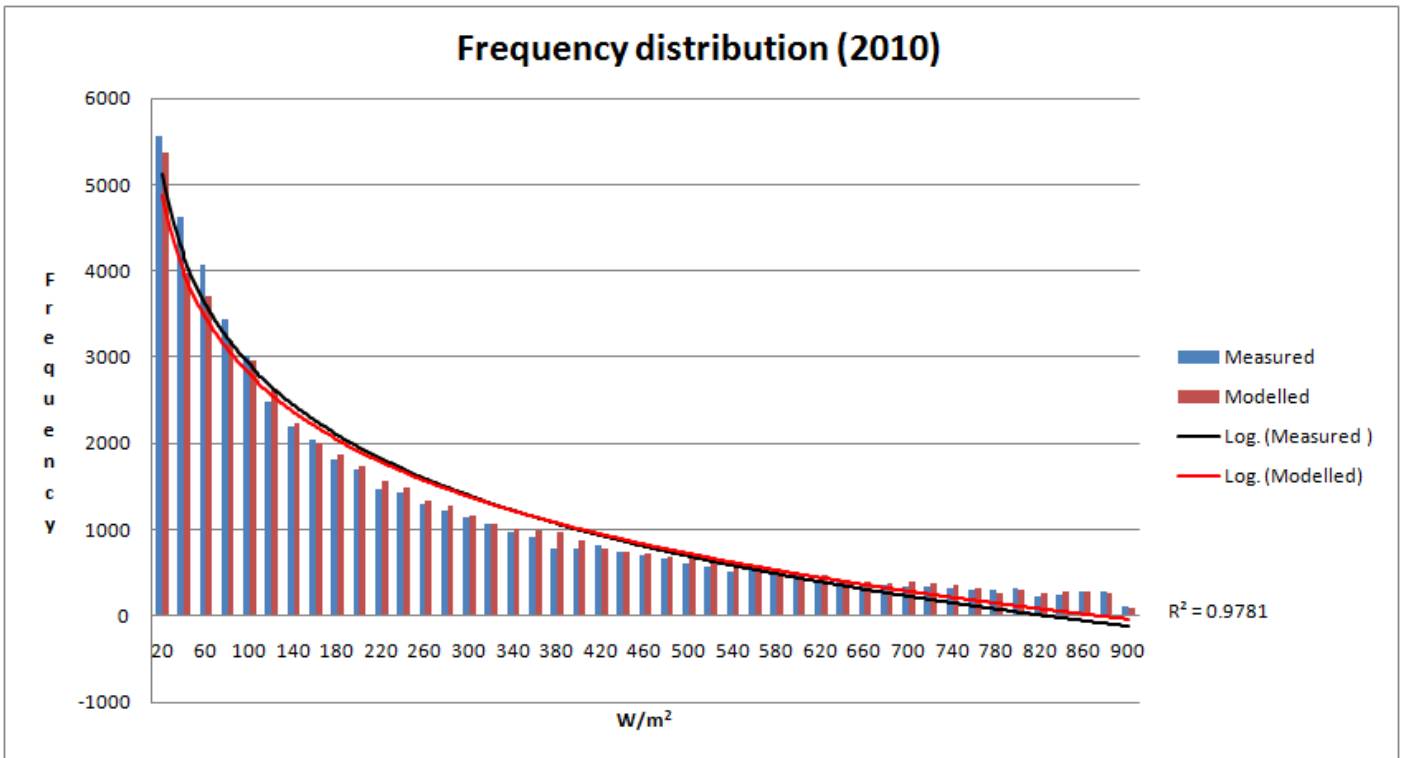


Figure 31: Frequency distribution for Reading (2010)

5.2 – Input data from different location

The above frequency distribution graphs use hourly averages calculated from the data which was provided by the University of Reading, in other words the input data is averages of the same data which was used to construct the transition probability matrix which make up part of the model therefore if the model approach gives good correlations it would be expected that a good correlation between measured and modelled data for Reading should be seen.

However, as part of the further analysis of the author's model it was important to determine how accurately solar radiation data could be generated for locations outside of Reading based on the utilisation of the transition probability matrix constructed from the Reading data. In order to investigate this, data from the Isle of Lewis, 515 miles North West of Reading [69] was used as an input, this data was graciously provided by Hebrides weather (hebwx.co.uk) and a similar process of calculating the frequency distribution over a year was carried out. The results of this process are shown in figure 32 below. As can be seen even though the model is not constructed

from any of the data the correlations between the measured and generated data are still very accurate with an R^2 value of 0.942 this proves that the model constructed can be used for locations other than Reading. It is the authors opinion that the method used in that the hourly inputs are used as a base for the generation of data and the fact that the model utilises clear sky index rather than the generation of actual global horizontal radiation values allows the model to accurately reproduce values for outside of Reading, as although the maximum and minimum values of solar radiation experienced at the locations will be different the transition between cloud conditions is likely to be similar on the whole throughout the UK.

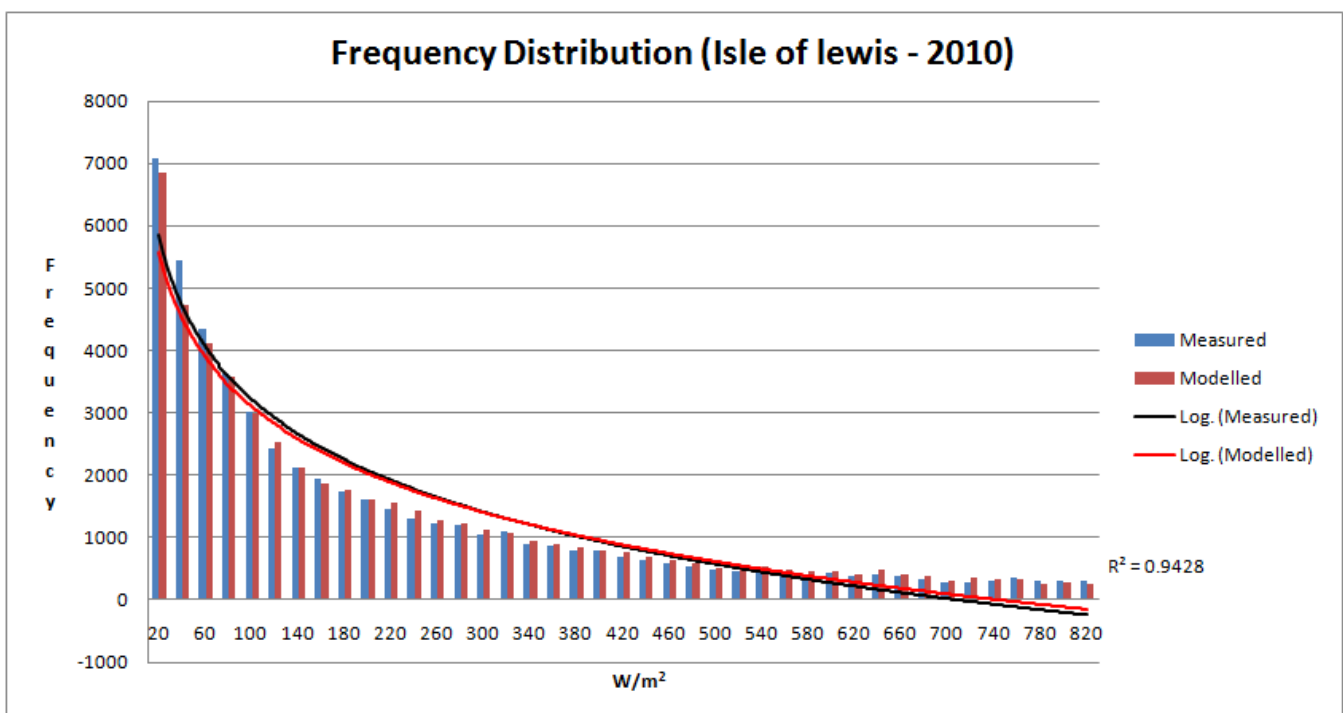


Figure 32: Frequency distribution for Lewis (2010)

6.0 – Conclusion and Further work

6.1 – Conclusion

In the modern world one of the key components to understanding and improving any field is the collation of data relating to that field of study, for years the generation of power via photovoltaic's and the understanding of energy usage within the built environment and its relationship with solar radiation has been based on the usage of hourly averages of solar data, this for a long time has thought to have been adequate. However as the increase in power generation from solar photovoltaic's has grown and the popularity of energy efficient construction and design within the built environment along with ideas of how to best utilise natural lighting and the usage of solar gains to heat a structure have become more mainstream it has become abundantly clear that the use of these hourly averages is no longer adequate. Only now have institutions and universities started to measure solar data with high enough resolutions so as to be useful in these fields of study, and even so only a handful of places are doing so, this in some ways is severely affecting the ability for the science and engineering community to truly understand the effects of solar variability on these subjects.

The aim of this project was to establish a simple, easy to use model which would have the ability to transform hourly averages of global horizontal radiation into 5 minute resolution global horizontal radiation which is representative of the base hourly data. To achieve this a method was developed which utilised transition probability matrix constructed from data graciously provided by the University of Reading and a Markov chain process, this being a well understood and proven technique for generating stochastic data which is representative of an input.

It has been shown that the method used in the construction of this model works well when considering the frequency distribution of solar radiation over a year, for the same location were the transition probability matrix was derived, However it has also been shown that this method can work equally well for locations which are vastly different in terms of overall solar duration and maximum solar radiation which is

experienced this is thought to be a result of the utilisation of clear sky index rather than the use of radiation measurements directly as discussed above. However this is based on limited testing carried out using a single data set from the Isle of Lewis therefore to truly establish how robust the approach taken actually is may require further testing and modelling using different data sets.

The model described within this report is provided freely and is available for download with a creative commons licence from the Energy Systems Research Unit (ESRU), University of Strathclyde website and may be freely used and adapted by anyone who wishes to do so.

6.2 – Further work

From the outset of this project it has been evident that the project lends itself to being used as bases for further work, both in the improvement of the model itself and in the utilisation of the model in investigating the variability of solar radiation and its impact on a number of sectors.

This section of the report aims to explore the authors opinions on where this model could be best utilised and further improved.

Higher resolution – The current model constructed generates 5 minute resolution global horizontal radiation; this in many ways is high enough resolution to take account of the vast majority of variability experienced within an average day. However the model could be further improved to generate up to a resolution of 1 minute this would take into account very fast moving cloud which may not be fully taken account of in the 5 minute resolution model constructed, However the major problem with this is the availability of accurate and consistent 1 minute resolution data to construct the transition probability matrix, at the time of writing of this report the 1 minute data which is freely available is either not of sufficient quality or there is not enough of it to accurately construct a good transition probability matrix (i.e. there is only a maximum of 1 years worth of data), as more institutions and universities increase their measurement resolutions the construction of a similar model with a 1 minute resolution would be easily achievable.

Ability to account for cloud enhancement - The current model does not have the ability to account for so called cloud enhancement, this being were the transition between clear sky and clouds actually increases the solar radiation measured above the point of clear sky radiation due to the extra reflected radiation experienced due to the nearby clouds albedo. A further project could investigate how/if this could be taken account in the model.

Investigate PV model - A further investigation into the variability of photovoltaic output with the use of high resolution data compared to the use of hourly averages could be implemented into the model, in particular a PV model in its own right could be implemented into the model were the user could input the size, location and efficiency of a planned PV array along with the hourly averages of solar radiation and the variability of this radiation could be taken account of. This could also include an investigation into the impact of the variability of the solar radiation on the inverter .

Implementation of model into ESP-r and Merit - The implementation of the model into modelling software such as ESP-r and merit could be advantageous, as this would allow the use of hourly weather files to be converted into high resolution data if the user wished thus easily investigating the impact of solar variability on their building model.

Investigate worldwide accuracy - During the model validation section it was shown that although the models transition probability matrixes were constructed using only data from the location of Reading, the model also provided a good fit when the location and input data from the Isle of Lewis was used. Further investigation could be undertaken to see how far afield this good correlation is achieved and if the model still achieves these results for a location which has drastically different weather from the UK, it could also be explored how this model could take account of these differences for instance with different transition probability matrix for different locations (i.e. tropical, temperate regions etc)

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9.0 - Appendixes

Appendix 1

```
Sub Row_count()
Dim count As Integer, myRange As Range
Dim B As Integer
Dim i As Integer
Dim t As Integer

Sheets("Data_Input").Select
Set myRange = Columns("c:c")

count = Application.WorksheetFunction.CountA(myRange)

B = count - 5
i = (B * 12) + 34
t = 35
Worksheets("Clear").Range("D37:BN8975").ClearContents
Sheets("Clear").Select
Range("D" & t & ":BN" & t).Select
Selection.AutoFill Destination:=Range("D" & t & ":BN" & i), Type:=xlFillDefault
Range("D" & t & ":BN" & t).Select
Sheets("Data_Input").Select

End Sub
```

Appendix 2

```
Sub avg_1()
Dim i As Long
Dim count As Integer, myRange As Range
Dim B As Integer
Dim s As Integer
Dim t As Integer
Dim U As Integer

Worksheets("Hourly_Calculations").Range("D:D").ClearContents

Sheets("Data_Input").Select
Set myRange = Columns("c:c")

count = Application.WorksheetFunction.CountA(myRange)

B = count - 5
s = (B * 12) + 34
t = B + 1
U = 3

For i = 35 To s Step 12

Worksheets("Hourly_Calculations").Range("D"           &
Rows.count).End(xlUp).Offset(1).Value              =
WorksheetFunction.Average(Worksheets("Clear").Range("AR" & i).Resize(12))

Next i
Worksheets("Hourly_Calculations").Range("E4:G745").ClearContents

Sheets("Hourly_Calculations").Select
Range("E3:G3").Select
Selection.AutoFill Destination:=Range("E" & U & ":G" & t), Type:=xlFillDefault
Range("E" & U & ":G" & U).Select
Sheets("Data_Input").Select

End Sub
```

Appendix 3

Sub SimulateClearnessIndex_month_Click()

Randomize

```
Dim NxtTimeInt As Integer
Dim fRand As Double
Dim fRand2 As Double
Dim CumulativeProb As Double
Dim iRow As Integer
Dim Colref As String
Dim x As Integer
Dim y As Integer
Dim iMonth As Integer
```

```
Dim dKArray(1 To 100, 1 To 100) As Double
For y = 1 To 100
  For x = 1 To 100
```

```
    Colref = Cells(1, x + 2).Address(True, False, xlA1)
    Colref = Left(Colref, InStr(Colref, "$") - 1)
```

```
    iMonth = Worksheets("Clear").Range("E4").Value
```

```
    dKArray(y, x) = Range("TPM" & CStr(iMonth) + "!" + Colref + CStr(y +
9)).Value
  Next x
Next y
```

```
Dim c As Variant
Dim B As Variant
Dim off1 As Variant
Dim off2 As Variant
Dim off3 As Variant
Dim count As Integer, myRange As Range
Dim s As Integer
```

```
Sheets("Data_Input").Select
Set myRange = Columns("c:c")
```

```
count = Application.WorksheetFunction.CountA(myRange)
```

```
s = count - 5
```

```
For c = 1 To s
```

```
off1 = (c * 12) - 10
```

```
off2 = (c * 12)
```

```
off3 = (c * 12) - 11
```

```
Dim vSimulationArray(1 To 12, 1 To 1) As Variant
```

```
Dim CurrentCSI As Integer
```

```
CurrentCSI = Worksheets("Hourly_Calculations").Range("G" & c + 1) * 100
```

```
If CurrentCSI <= 0 Then
```

```
With Worksheets("5min_Output")
```

```
.Range(.Cells(off3, 9), .Cells(off2, 9)).Value = 0
```

```
End With
```

```
Else
```

```
Worksheets("5min_Output").Range("I" & off3).Value = CurrentCSI / 100
```

```
For NxtTimeInt = 1 To 12
```

```
fRand = Rnd()
```

```
CumulativeProb = 0
```

```
Dim i As Integer
```

```
For i = 1 To 100
```

```
CumulativeProb = CumulativeProb + dKArray(CurrentCSI, i)
```

```
If fRand <= CumulativeProb Then
```

```

        CurrentCSI = i
        Exit For

    End If
Next i

Dim dk As Double
If CurrentCSI = 100 Then
    dk = 1
Else
    dk = (CurrentCSI / 100)
End If

vSimulationArray(NxtTimeInt, 1) = dk

Next NxtTimeInt

With Worksheets("5min_Output")
.Range(.Cells(off1, 9), .Cells(off2, 9)).Value = vSimulationArray
End With
End If
Next c

End Sub

```