

Operational research for urban solar development

“PV failure detection based on operational time series”

26/11/2024

Alexandre Mathieu



Curriculum Plan

Day	Time	Duration	Content
Wednesday 13/11/2024	11h15-12h45 14h15-15h45	1h30 + 1h30	50% Lecture / 50 % Hands-on
Tuesday 26/11/2024	9h45-13h00	1h30 + 1h30	25% Lecture / 75 % Hands-on
Monday 02/12/2024	13h15-16h15	3h	15% Lecture / 85 % Hands-on
Monday 09/12/2024	8h-11h 13h15-16h15	6h	10% Lecture / 90 % Hands-on/Project
Tuesday 10/12/2024	8h-11h	3h	10% Lecture / 90 % Project
Monday 16/12/2024	8-11h	3h	10% Lecture / 90 % Project
Thursday 19/12/2024	9h45-12h45	3h	10% Lecture / 90 % Project
Monday 06/01/2025	13h15-14h45	1h30	100% Project
Monday 13/01/2025	9h45-11h45	1h30	100% Project
Total		27h	

Agenda



Review notebook last week

PV performance model steps

Notebook: Shading effect on irradiance

Review first notebook

Notebook recap 13/11/2024

Google collab link: https://github.com/AlexandreHugoMathieu/pv-fault-detection-solar-academy/blob/master/notebooks/python_intro_poa.ipynb

Correction: https://github.com/AlexandreHugoMathieu/pv-fault-detection-solar-academy/blob/master/notebooks/python_intro_poa.ipynb

Review first notebook

Notebook recap 13/11/2024

Python commands 1/4

ts

a	1
b	2
c	3

index

values

df

	first_column	second_column
a	1	2
b	2	4
c	3	6

index

`import numpy as np` # import to your python instance the package "numpy" and rename it "np" (helpful for math calculations)

`import pandas as pd` # import to your python instance the package "pandas" (helpful for data structure and calculations)

`ts = pd.Series([1, 2, 3], index=['a', 'b', 'c'])` # Initiate a pandas series into variable "ts"

`ts2 = ts + ts/2 + np.cos(ts) + np.pi` # Make calculate with "ts" and store it into "ts2"

`print(ts2)` # print serie ts

`ts.plot(marker="o")` # Make a plot of ts with "o" (circle) marker

`df = pd.DataFrame()` # Initiate an empty dataframe into variable "df"

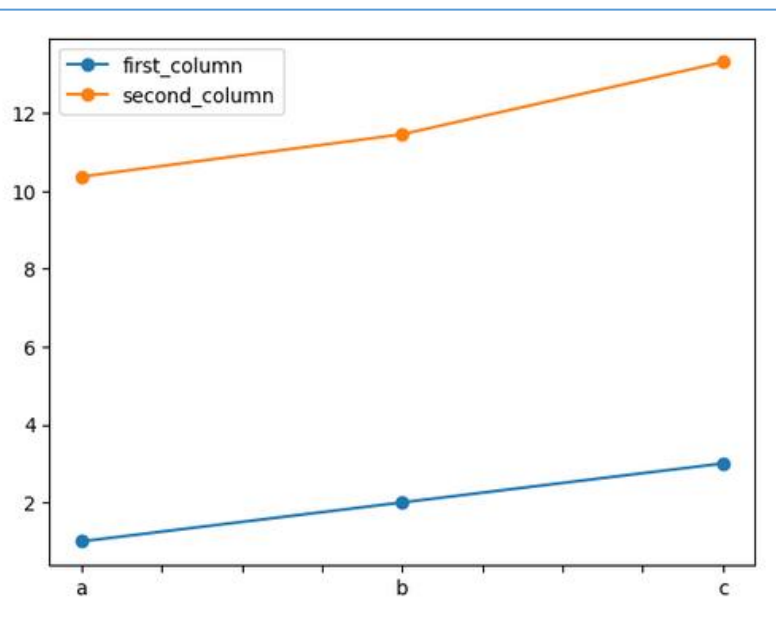
`df["first_column"] = ts` # Store "ts" serie in a column labeled "first_column"

`df["second_column"] = ts2 * 2` # Store "ts2" serie in another column labeled "second_column"

`df.plot(marker="o")` # Make a plot of df with "o" (circle) marker

`df.loc["a", :]` # Select the entire row with "a" as index

`df.loc["a", "first_column"]` # Select the value with "a" as index and "first_column" as column



Review first notebook

Notebook recap 13/11/2024

Python commands 2/4

	poa_global	poa_direct	poa_diffuse	poa_sky_diffuse	poa_ground_diffuse
2022-01-01 00:00:00+01:00	NaN	NaN	NaN	NaN	NaN
2022-01-01 01:00:00+01:00	0.000000	0.000000	0.000000	0.000000	0.000000
2022-01-01 02:00:00+01:00	0.000000	0.000000	0.000000	0.000000	0.000000
2022-01-01 03:00:00+01:00	0.000000	0.000000	0.000000	0.000000	0.000000
2022-01-01 04:00:00+01:00	0.000000	0.000000	0.000000	0.000000	0.000000
2022-01-01 05:00:00+01:00	0.000000	0.000000	0.000000	0.000000	0.000000
2022-01-01 06:00:00+01:00	0.000000	0.000000	0.000000	0.000000	0.000000
2022-01-01 07:00:00+01:00	0.000000	0.000000	0.000000	0.000000	0.000000
2022-01-01 08:00:00+01:00	0.593235	0.000000	0.593235	0.589085	0.004150
2022-01-01 09:00:00+01:00	71.788066	46.706296	25.081770	24.724777	0.356993
2022-01-01 10:00:00+01:00	210.546485	150.470436	60.076049	59.167899	0.908150
2022-01-01 11:00:00+01:00	376.976885	313.961064	63.015821	61.519000	1.496821

Calculate POA, the lazy way

```
from pvlib.irradiance import get_total_irradiance # import the function
"get_total_irradiance from pvlib"
# On another note, pvlib* is a very useful package for PV modeling with plenty of convenient
functions, do not hesitate to look it up on the web
```

beta = 20 # tilt [°]

azimuth = 180 # azimuth [°]

rho = 0.2 # albedo

values

```
solar_position = pd.read_csv("solarpos_data.csv") # Import the data file
"solarpos_data.csv" which contains the sun path (azimuth and elevation) with datetime index
weather_data = pd.read_csv("sat_data.csv", index_col=0) # Import the data file
"sat_data.csv" which irradiance (dni, ghi, dhi) with datetime index
```

```
data = get_total_irradiance(beta, azimuth, solar_position["zenith"],
solar_position["azimuth"], weather_data["dni"], weather_data["ghi"],
weather_data["dhi"], albedo=rho) # Directly apply the isotropic models
```

```
print(data.head(12)) # Show the first 12 lines of the DataFrame
```

Pvlib ref

*William F. Holmgren, Clifford W. Hansen, and Mark A. Mikofski.
"pvlib python: a python package for modeling solar energy
systems." Journal of Open Source Software, 3(29), 884, (2018).

<https://doi.org/10.21105/joss.00884>

Review first notebook

Notebook recap 13/11/2024

Python commands 3/4

```
##### Function definition #####
```

```
# Function, useful to store few lines of code you want to reuse and apply with different inputs
```

```
def my_func_name(argument1, argument2):
```

```
# Define the function "my_func_name" with the "def" command and a small increment tab to the right
```

```
    y = argument1 + argument2 * 2
```

```
    return y
```

```
my_func_name(1,2) # Apply the function with the two mandatory arguments and return 5
```

```
my_func_name(1,argument2=3) # Return 7
```

```
my_func_name(argument2=3, argument1=2) # Return 8
```

Review first notebook

Notebook recap 13/11/2024

Python commands 4/4

```
##### Function definition #####
```

```
# Function, useful to store few lines of code you want to reuse and apply with different inputs
```

```
def my_func_name2(argument1, argument2, argument3=0):
```

```
# Define the function "my_func_name" with the "def" command and a small increment tab to the right
```

```
# Argument3 is optional
```

```
    y = argument1 + argument2 * 2 + argument3
```

```
    return y
```

```
my_func_name2(1,2,2) # Apply the function with three arguments and return 7
```

```
my_func_name2(1,2) # Apply the function with the two mandatory arguments and return 5
```

```
my_func_name(argument2=3, argument1=2, 3)
```

```
# Crash because all arguments need to be specified on the right of a keyword argument
```

```
-> """ SyntaxError: positional argument follows keyword argument """
```


Review first notebook

Online Documentation

Section Navigation

Position

Clear sky

Airmass and atmospheric models

Irradiance

Methods for irradiance calculations

Decomposing and combining irradiance

Transposition models

`pvlib.irradiance.get_total_irradiance`

`pvlib.irradiance.get_sky_diffuse`

`pvlib.irradiance.isotropic`

`pvlib.irradiance.perez`

`pvlib.irradiance.perez_driesse`

`pvlib.irradiance.haydavies`

`pvlib.irradiance.klucher`

`pvlib.irradiance.reindl`

`pvlib.irradiance.king`

`pvlib.irradiance.ghi_from_poa_driesse_2023`

DNI estimation models

Clearness index models

API reference > ... > Transposition models > `pvlib.irradi...`

`pvlib.irradiance.get_total_irradiance`

```
pvlib.irradiance.get_total_irradiance(surface_tilt, surface_azimuth,
solar_zenith, solar_azimuth, dni, ghi, dhi, dni_extra=None, airmass=None,
albedo=0.25, surface_type=None, model='isotropic',
model_perez='allsitescomposite1990') #
```

[\[source\]](#)

Determine total in-plane irradiance and its beam, sky diffuse and ground reflected components, using the specified sky diffuse irradiance model.

$$I_{tot} = I_{beam} + I_{skydiffuse} + I_{ground}$$

Sky diffuse models include:

- isotropic (default)
- klucher
- haydavies
- reindl
- king
- perez
- perez-driesse

Parameters:

- **`surface_tilt`** (*numeric*) – Panel tilt from horizontal. [degree]
- **`surface_azimuth`** (*numeric*) – Panel azimuth from north. [degree]

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albedo=0.25, surface_type=None, model='isotropic',
model_perez='allsitescomposite1990') #
```

[\[source\]](#)

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Mandatory arguments

Parameters:

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DNI estimation models

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API reference > Transposition models > pvlib.irradi...

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solar_zenith, solar_azimuth, dni, ghi, dhi, dni_extra=None, airmass=None,
albedo=0.25, surface_type=None, model='isotropic',
model_perez='allsitescomposite1990') #
```

[\[source\]](#)

Determine total in-plane irradiance and its beam, sky diffuse and ground reflected components, using the specified sky diffuse irradiance model.

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Sky diffuse models include:

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- haydavies
- reindl
- king
- perez
- perez-driesse

Mandatory arguments

Optional (default) arguments

Parameters:

- **surface_tilt** (*numeric*) – Panel tilt from horizontal. [degree]
- **surface_azimuth** (*numeric*) – Panel azimuth from north. [degree]

Agenda



Review notebook last week

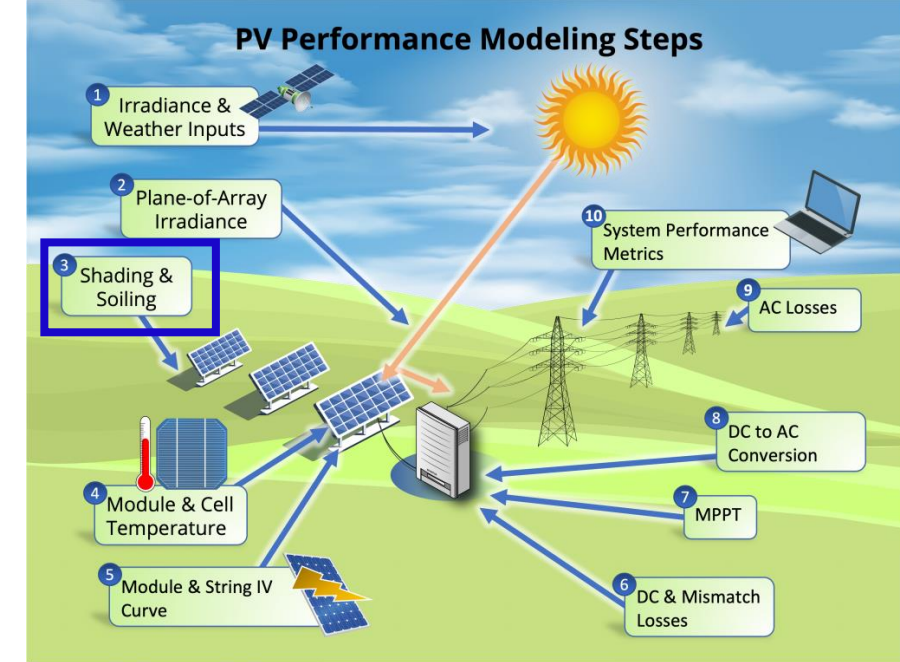
PV performance model steps

Modeling steps

3. Shading

Shadow can come from near and far elements and their impact can be distinguished into two categories

1. **Partial shading** refers to a condition where some but not all of the solar cells or panels in a PV array are exposed to sunlight while others are shaded.
2. **Full shading** occurs when the entire PV array is covered and deprived of direct sunlight.

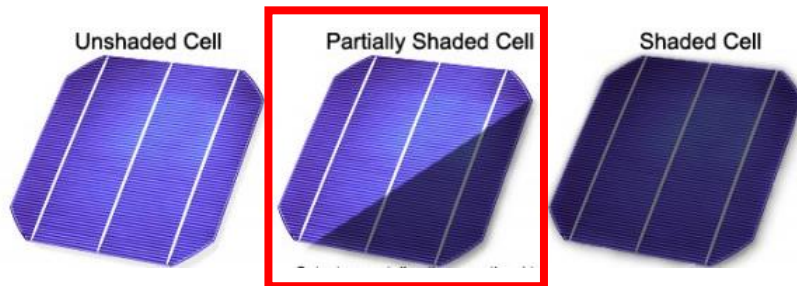


Modeling steps

3. Shading

Shadow can come from near and far elements and their impact can be distinguished into two categories

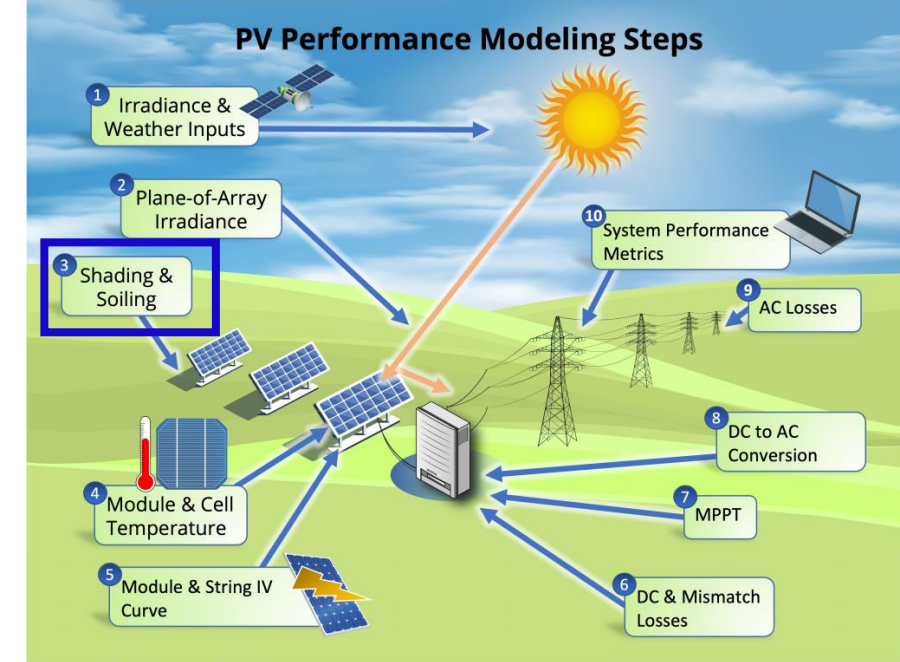
1. **Partial shading** refers to a condition where some but not all of the solar cells or panels in a PV array are exposed to sunlight while others are shaded.
2. **Full shading** occurs when the entire PV array is covered and deprived of direct sunlight.



In average, if a PV panel is around 8% covered, bypass diodes start to activate and ****might**** reduce the module power production to much lower levels.

Demonstrated in one of pvlib example:

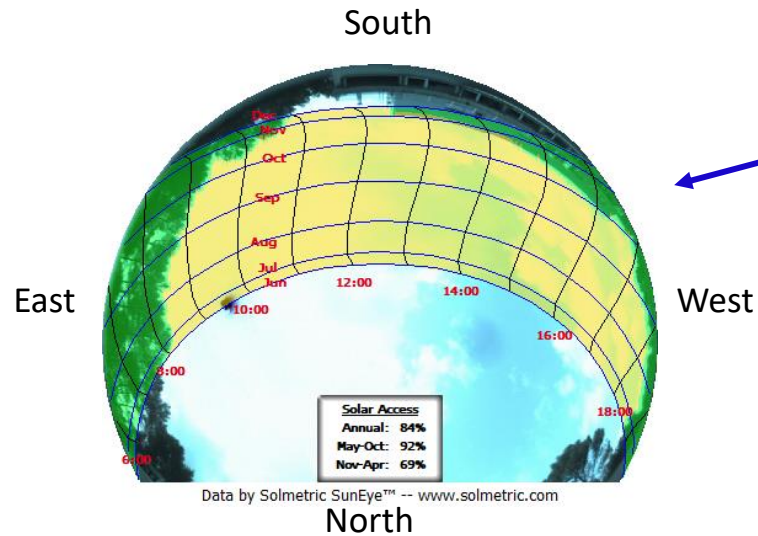
https://pvlib-python.readthedocs.io/en/stable/gallery/shading/plot_partial_module_shading_simple.html



Modeling steps

3. Shading

The Fisheye camera enables to account for the in-situ (full) shading in PV modeling.



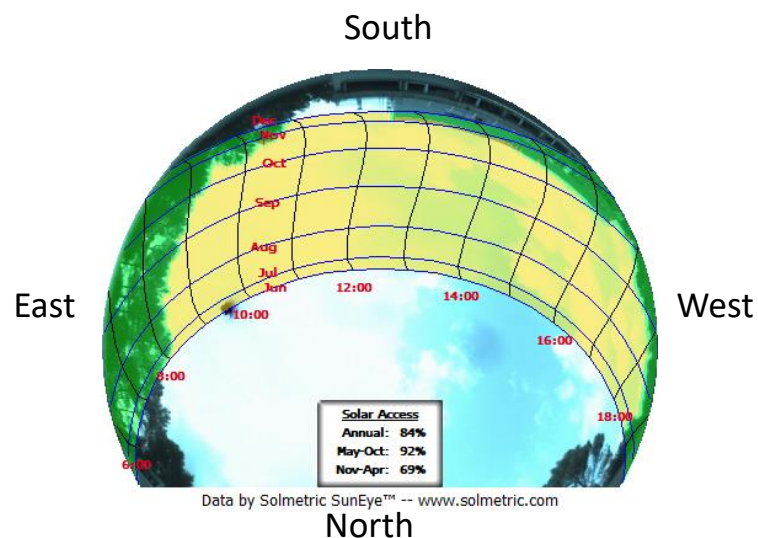
Example in South of France



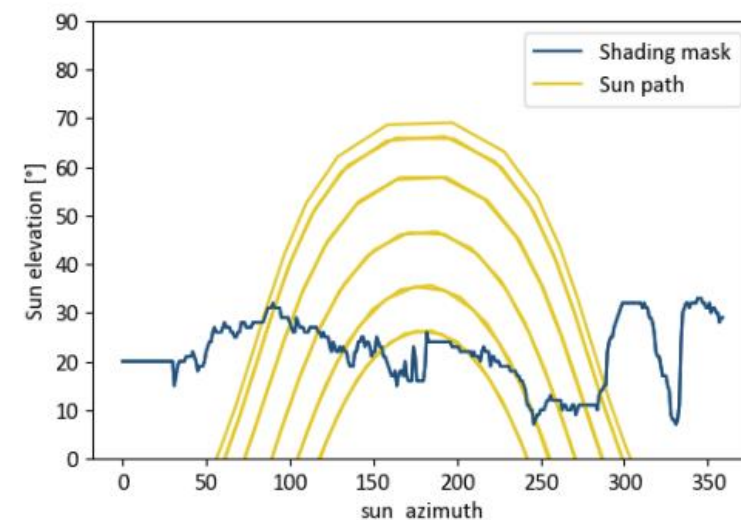
Modeling steps

3. Shading

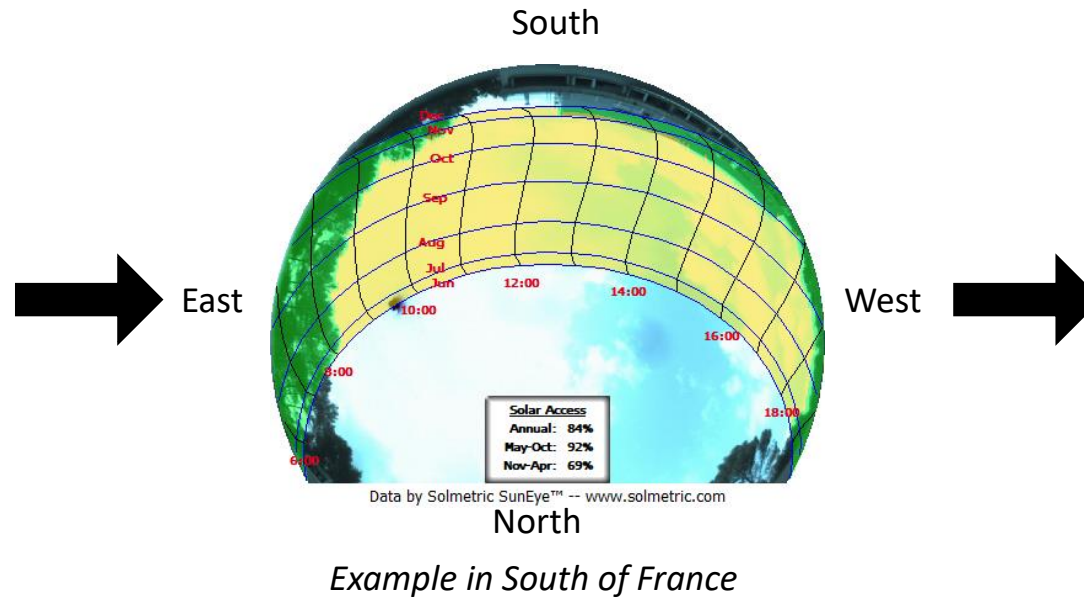
The Fisheye camera enables to account for the in-situ (full) shading in PV modeling.



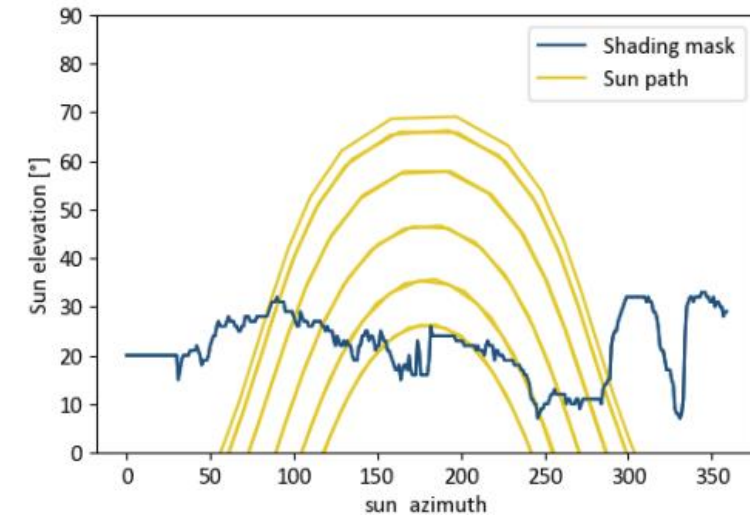
Example in South of France



Modeling steps



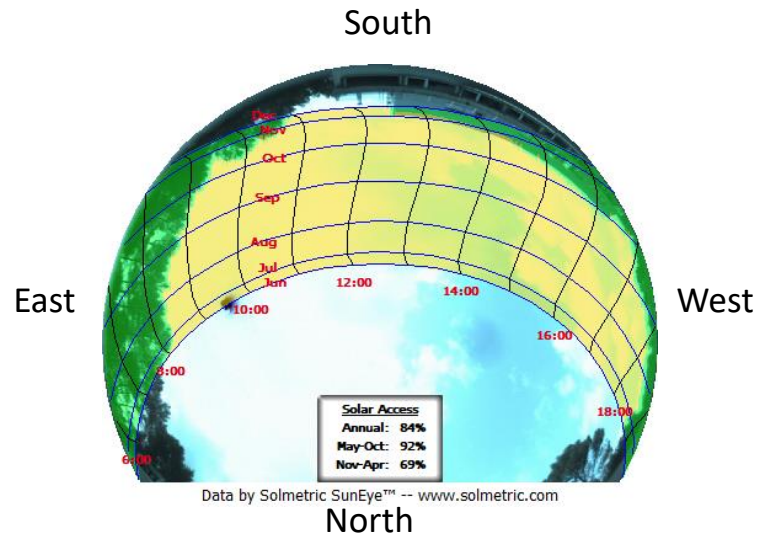
Example in South of France



Modeling steps

3. Shading

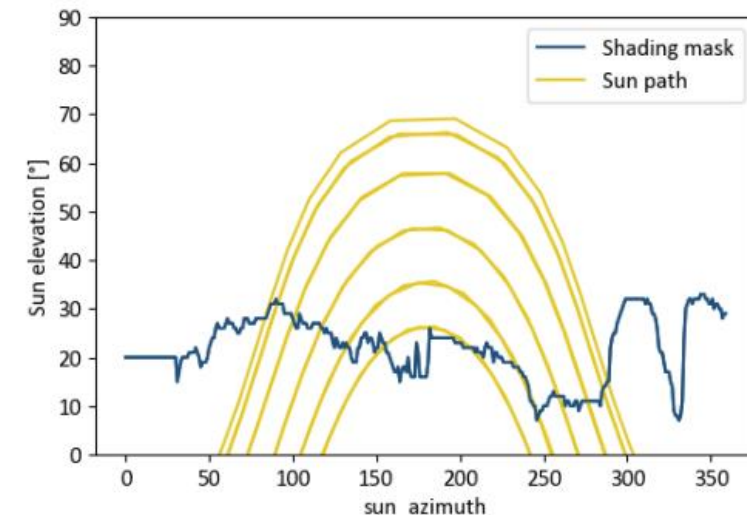
The Fisheye camera enables to account for the in-situ (full) shading in PV modeling.



Example in South of France

Under the black curve, simple assumptions lead to:

- $POA_b = 0 \frac{W}{m^2}$
- POA_d, POA_{grd} are not significantly modified



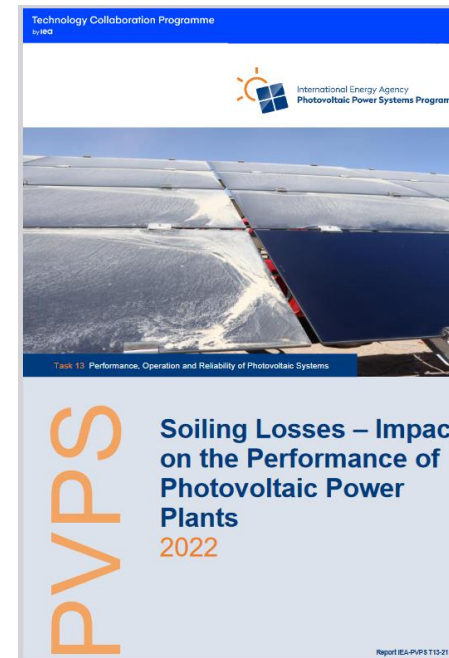
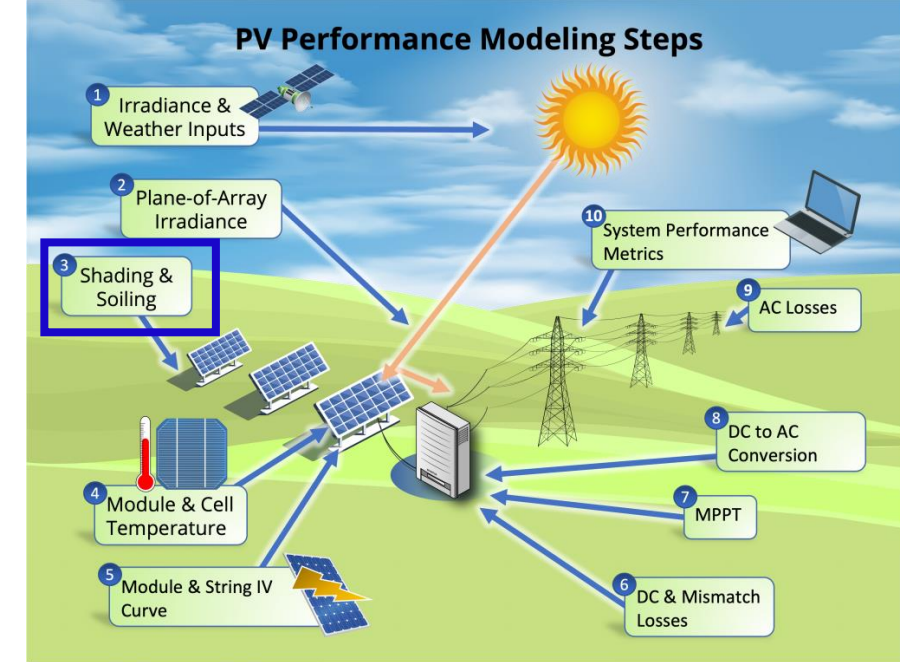
Modeling steps

3. Soiling losses

Big subject in the PV industry:

“Soiling losses refer to loss in power resulting from snow, dirt, dust and other particles that cover the surface of the PV module.” (Maghami et al., 2016)

Report of 130 pages on PV soiling: IEA PVPS, Soiling Losses – Impact on the Performance of Photovoltaic Power Plants 2022



Modeling steps

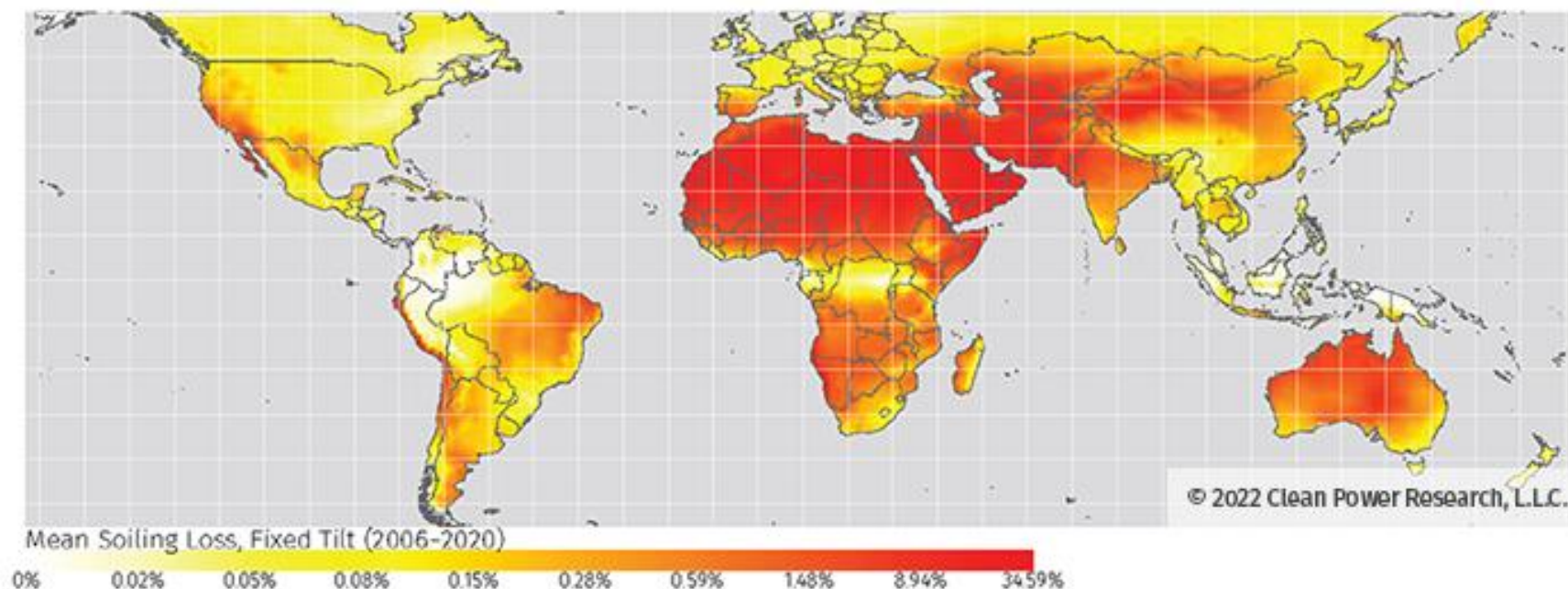
3. Soiling losses

Can roughly be ignored in France according to the HSU model.

Figure 1: Soiling Loss Map Based on SolarAnywhere Data and HSU Soiling Model

Annual Mean Soiling Loss (2006-2020); Fixed-tilt PV System

Annual
Soiling Loss



Modeling steps

3. Shading / Terrain horizon mask

Time for some
hands-on exercises,
Again!



Modeling steps

3. Shading / Terrain horizon mask

PVGIS: Website/Online Tool to estimate power production:

https://re.jrc.ec.europa.eu/pvg_tools/en/

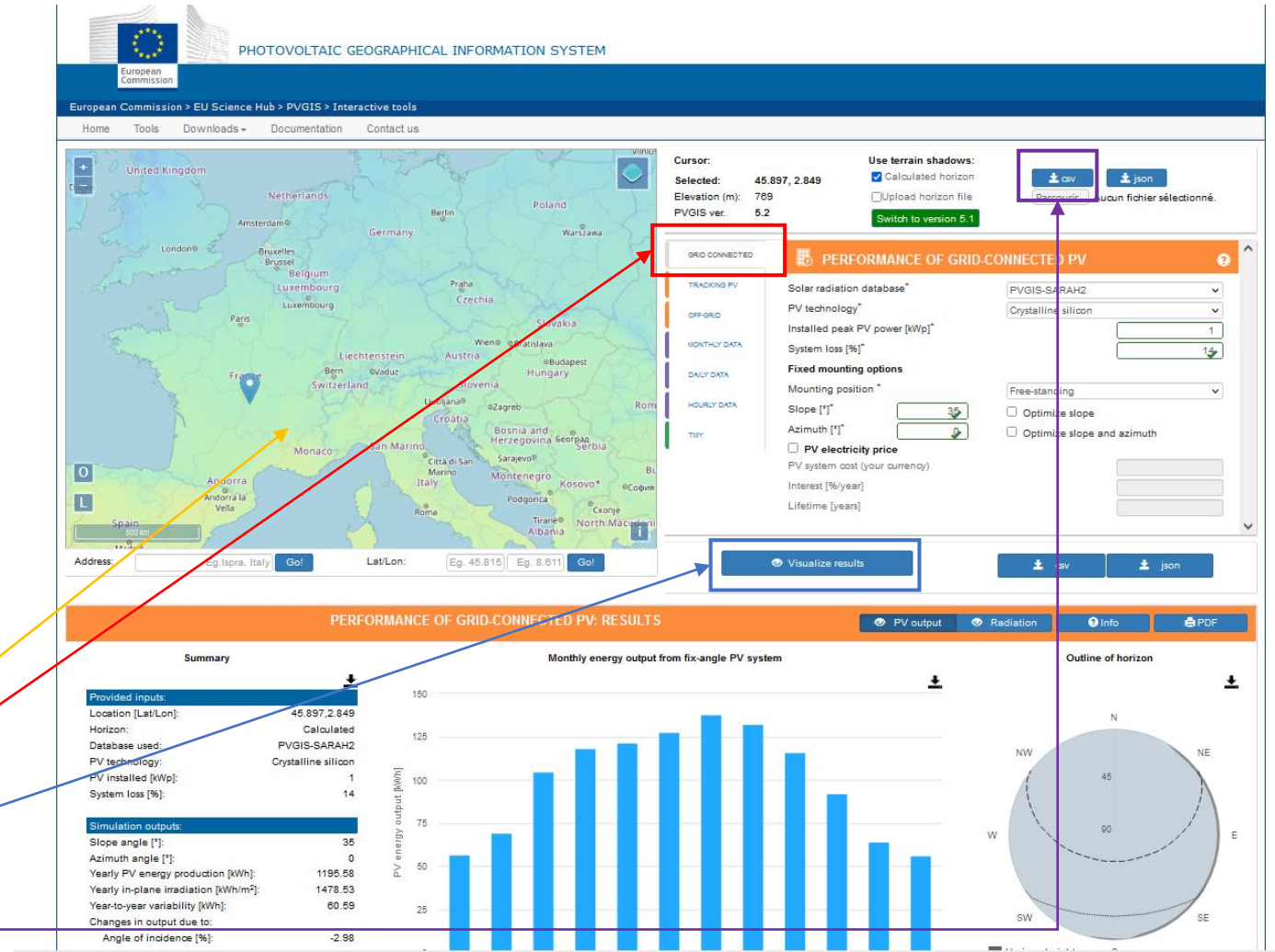
- Enables to extract the horizon mask with a Digital Surface Model (DSM).

Instructions:

- Generate a simulation on PVGIS
 - Click on the map on Grenoble and select the « Grid connected tab »
 - Vizualize
 - Extract the horizon file in csv format

2. Follow the instructions on the jupyter notebook and calculate the modified POA on one year.

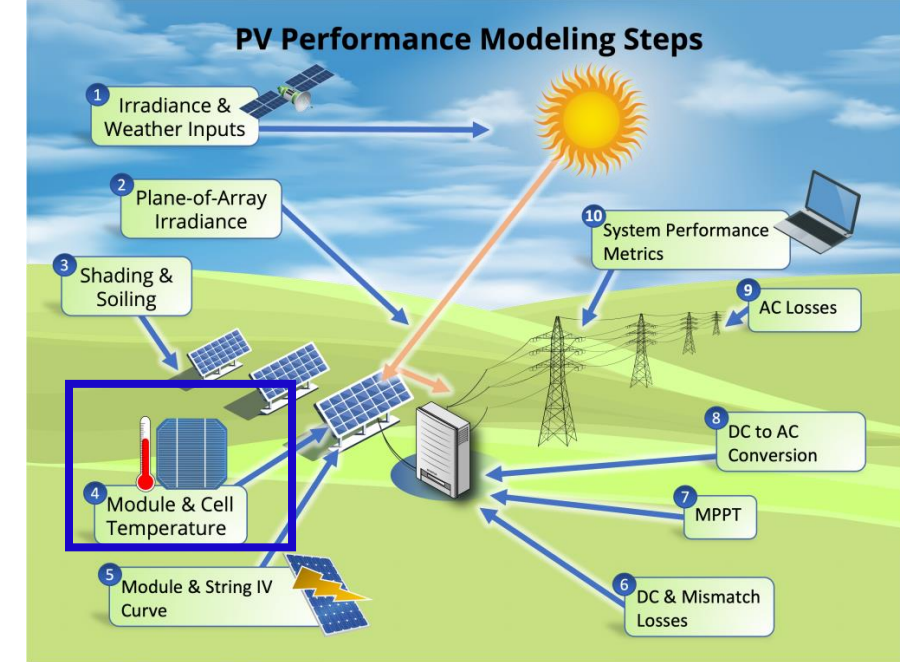
<https://colab.research.google.com/drive/1hB1pmBw-n7RiS99vCHQj2OKicfQxpcO?usp=sharing>



Modeling steps

4. Module and Cell temperature

The hotter a module is, the less efficient it is !



Modeling steps

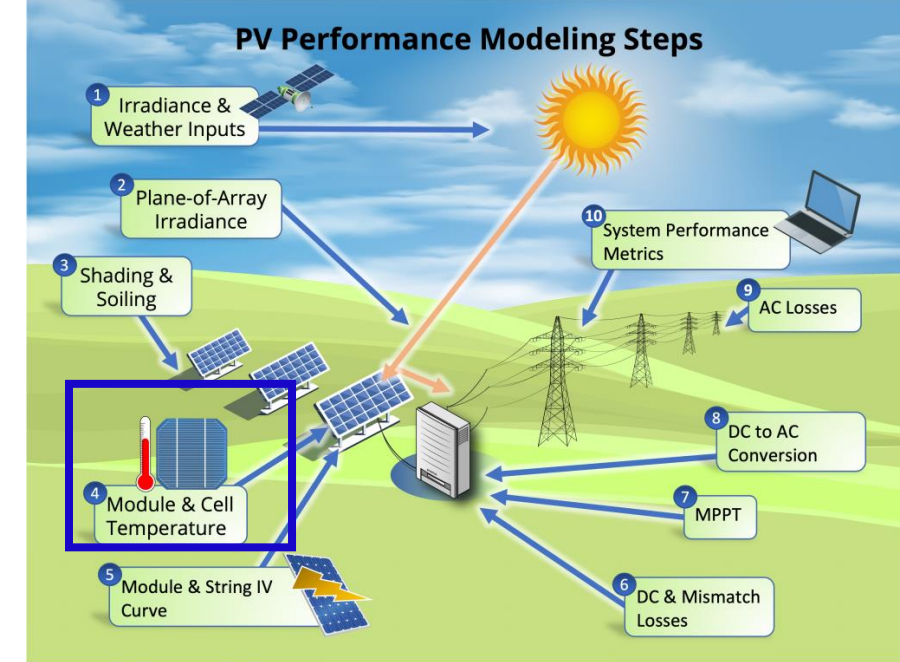
4. Cell temperature

Ross model:

Model to estimate the cell temperature T_c [°C] as function of ambient temperature and irradiance G_{POA} [W/m²].

$$T_c = T_a + G_{POA} \cdot k_{Ross}$$

k_{Ross} , typically in the range 0.02-0.05 K/m²/W.



Modeling steps

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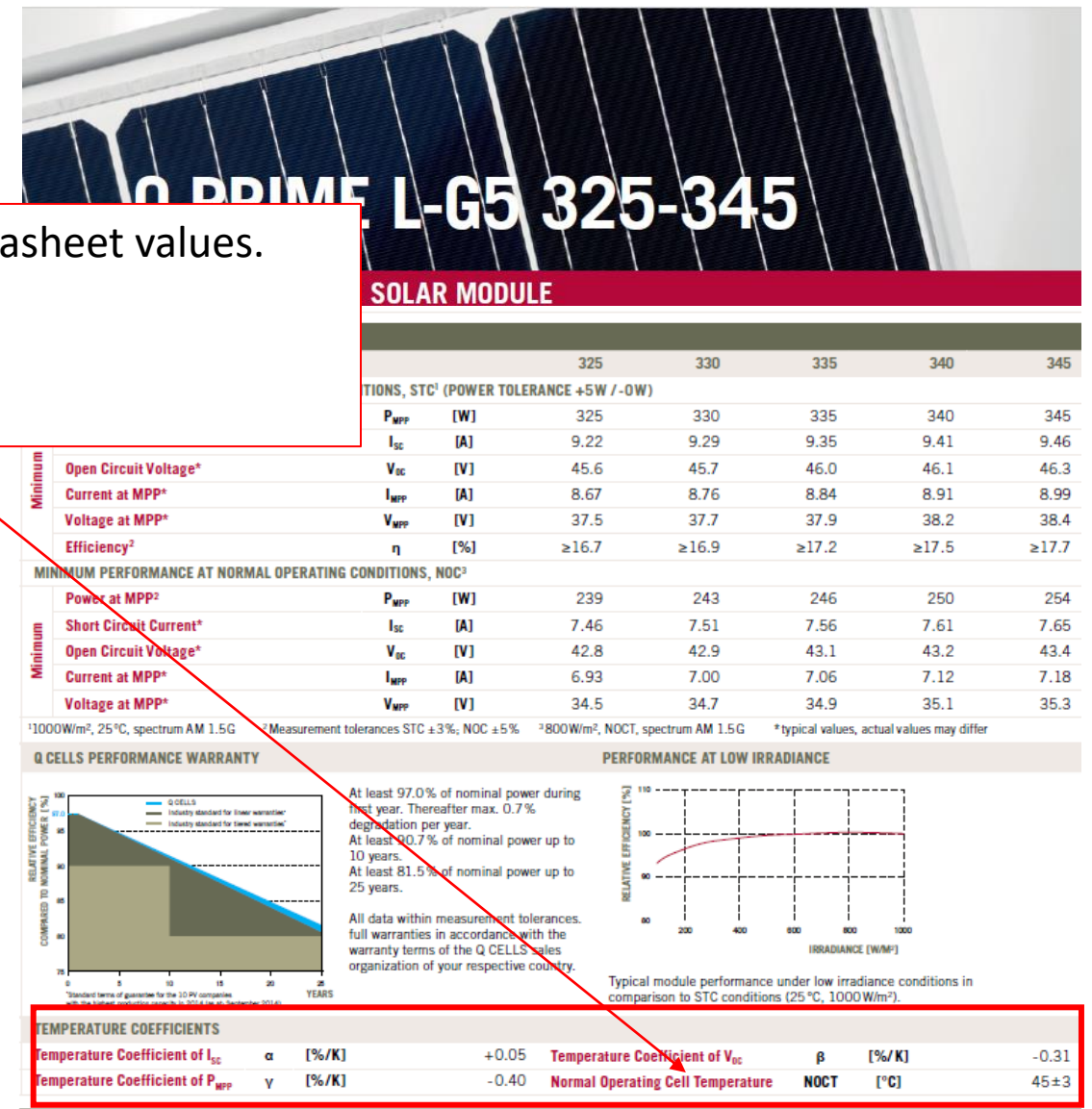
k_{Ross} , typically in the range 0.02-0.05 K/m²/W.

k_{Ross} can be fitted from datasheet values.

NOCT conditions:

$$G_{POA} = 800 \text{ W/m}^2$$

$$T_a = 20^\circ\text{C}$$



Modeling steps

4. Cell temperature

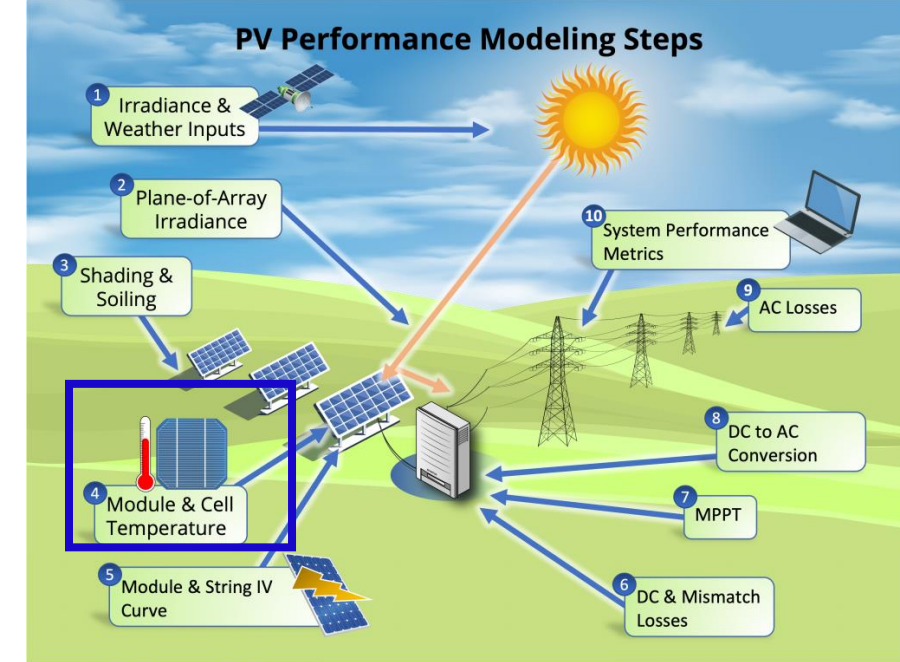
Faiman model:

Model to estimate the module temperature T_m [°C] as function of ambient temperature and irradiance G_{POA} [W/m²]
AND wind WS [$\frac{m}{s}$].

$$T_m = T_a + \frac{G_{POA}}{U_0 + U_1 \cdot WS}$$

U_0 is the constant heat transfer component [$\frac{W}{Km^2}$]

U_1 is the convective heat transfer component [$\frac{W}{Km^2(\frac{m}{s})}$]



Modeling steps

4. Cell temperature

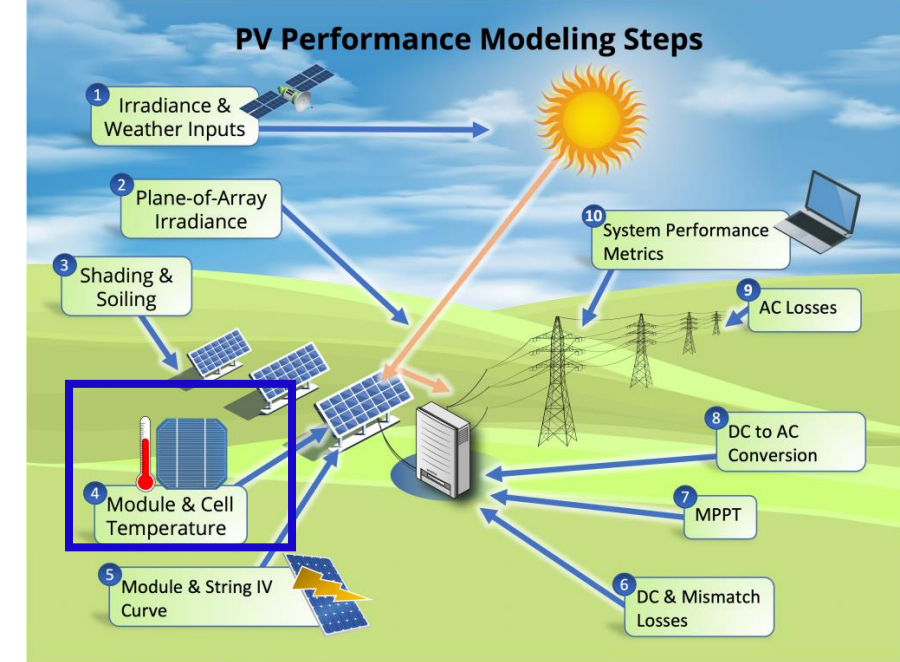
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U_0 is the constant heat transfer component [$\frac{W}{Km^2}$]

U_1 is the convective heat transfer component [$\frac{W}{Km^2(\frac{m}{s})}$]



In some cases, $T_c \simeq T_m$ can be assumed
Between T_c and T_m , only few degrees of difference

Resources

- Modeling guide PVPMC: <https://pvpmc.sandia.gov/modeling-guide/>
- Python / Pvlb tutorial: <https://pvsc-python-tutorials.github.io/PVSC48-Python-Tutorial/>
- To go further:
 - The Use of Advanced Algorithms in PV Failure Monitoring: https://iea-pvps.org/wp-content/uploads/2021/10/Final-Report-IEA-PVPS-T13-19_2021_PV-Failure-Monitoring.pdf

Shading

PV education 7.2, widget: <https://www.pveducation.org/pvcdrom/modules-and-arrays/shading>

- How to take into account shading from LIDAR data:
Solar Energy Potential Assessment on Rooftops and Facades in Large Built Environments Based on LiDAR Data, Image Processing, and Cloud Computing. Methodological Background, Application, and Validation in Geneva (Solar Cadaster), Desthieux et al.
<https://www.frontiersin.org/journals/built-environment/articles/10.3389/fbuil.2018.00014/full>

That's it

