

# *Operational research for urban solar development*

*“PV failure detection based on operational time series”*

*27/11/2023 - Morning*  
Alexandre Mathieu



# Agenda



**Curriculum**

**Audience**

**PV performance model steps**

# Curriculum

## Initial plan

Day	Time	Duration	Content
<b>Monday</b> <b>27/11/2023</b>	9h45-11h15 12h30-14h	1h30 + 1h30	50% Lecture / 50 % Hands-on
<b>Tuesday</b> <b>05/12/2023</b>	8h-9h30 9h45-11h15	1h30 + 1h30	50% Lecture / 50 % Hands-on
<b>Thursday</b> <b>07/12/2023</b>	8h-11h 12h45-15h45	6h	25% Lecture / 75 % Project
<b>Monday</b> <b>11/11/2023</b>	8h-11h 12h30-15h30	6h	10% Lecture / 90 % Project
<b>Friday</b> <b>22/12/2023</b>	8h-9h30	1h30	100 % Project

# Curriculum

## Content

**PV performance modeling** from weather variables (Irradiance in the plane of array, ambient temperature etc..) and system configuration, which includes:

- Module temperature
- DC power
- AC/DC efficiency
- Performance metric calculations.

**Failure detection** based on real production time series.

Hands-on with **python** notebooks.

# Curriculum

## Motivations

**PV capacity increases exponentially**  
+20%/year worldwide since 2010<sup>[1]</sup>



**Complex in-situ O & M:**  
No one-fit-all O&M<sup>[2]</sup>  
Expensive<sup>[3]</sup>



**15%<sup>[4]</sup> of underperformances:**  
« Recoverable » Energy: 5%<sup>[4,5]</sup>

[1] Masson, G., Kaizuka, I., 2021. Trends in Photovoltaic Applications (IEA-PVPS T1-41:2021). IEA PVPS

[2] U. Jahn et al., 'Guidelines for Operation and Maintenance of Photovoltaic Power Plants in Different Climates', IEA, Technical Report IEA-PVPS T13-07:2022, Oct. 2022.

[3] « Coûts et rentabilités du grand photovoltaïque en métropole continentale », CRE, 2019

[4] Leloux, J., Narvarte, L., Trebosc, D., 2011. Performance analysis of 10,000 residential PV systems in France and Belgium. Presented at the 26th European Photovoltaic Solar Energy Conference and Exhibition, Hamburg, Germany.)

[5] Raycatch, 'Solar Asset Optimization, Industry Benchmark Study', Tel Aviv, Israel, Feb. 2021.

# Curriculum

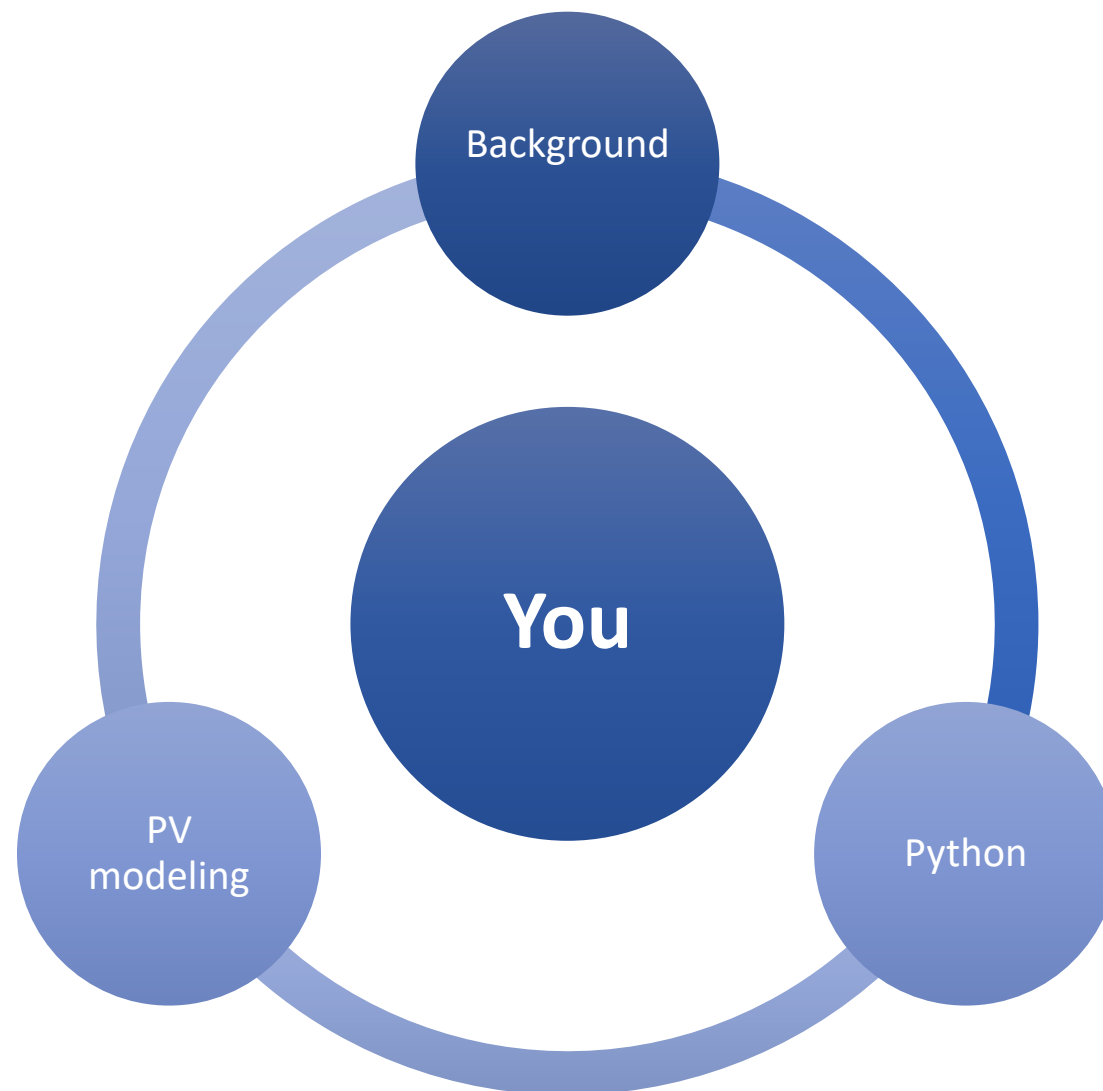
## Objectives

At the end of this course, you should be able to:

- Perform data treatment and analysis with Python on operational PV time series.
- Model and calculate the performance of a PV installation.
- Detect and quantify underperformances.

# Audience

# Audience





# Agenda

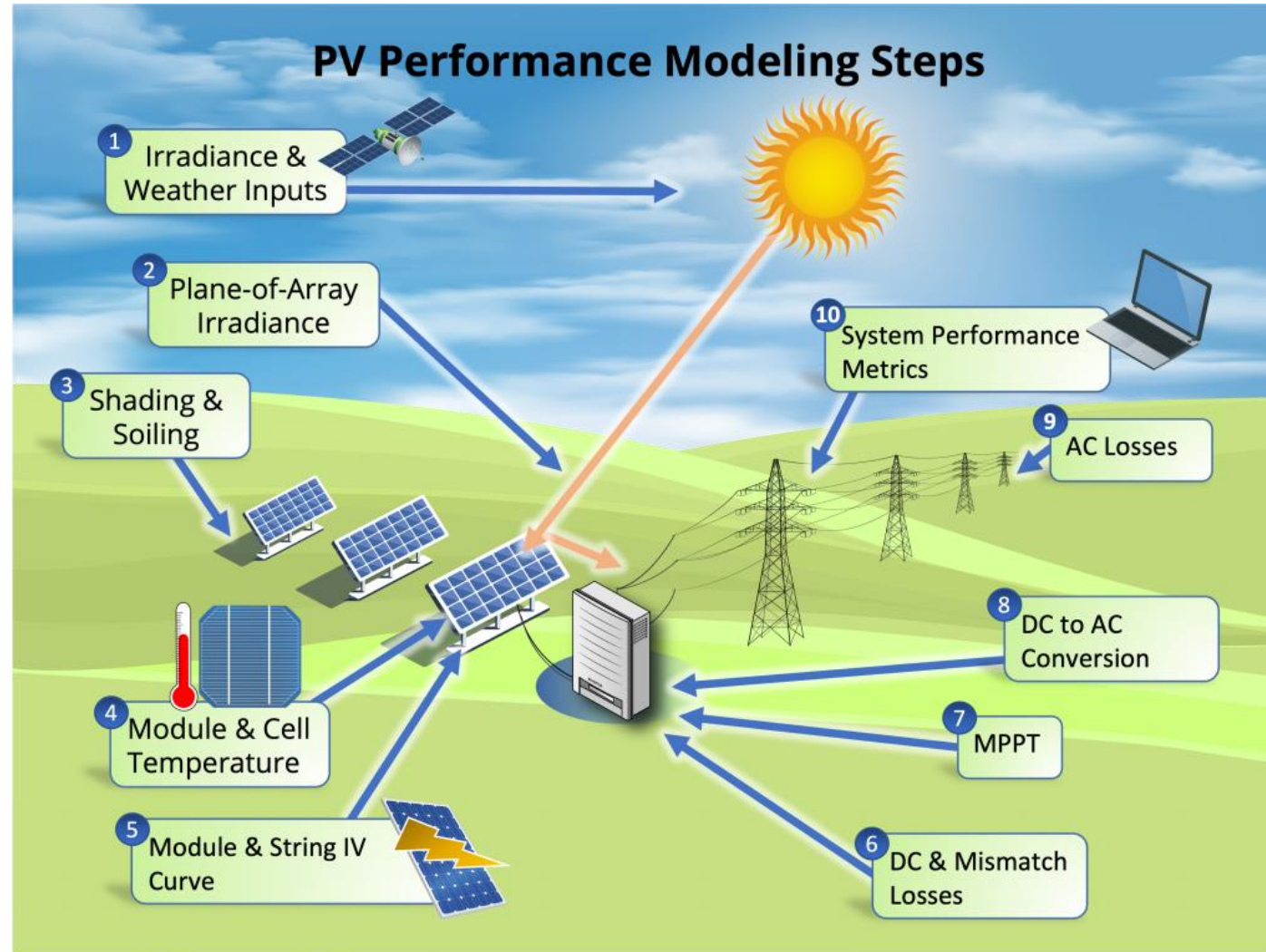


Curriculum

Audience

**PV performance model steps**

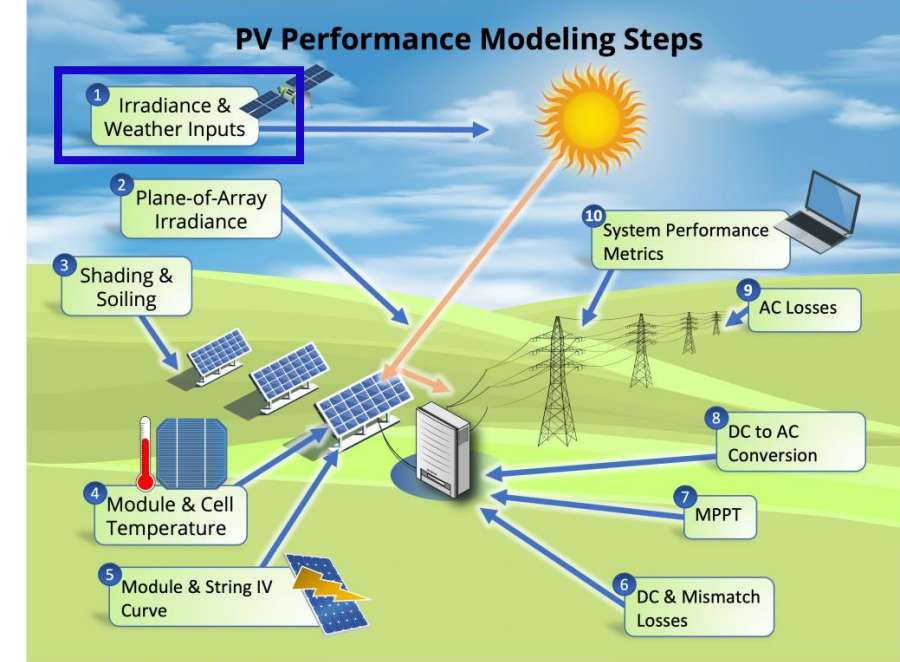
# Modeling steps



\*image from: PVSC48 python tutorial

# Modeling steps

1. **Weather data:** Irradiance, ambient temperature, humidity, rain, snow...



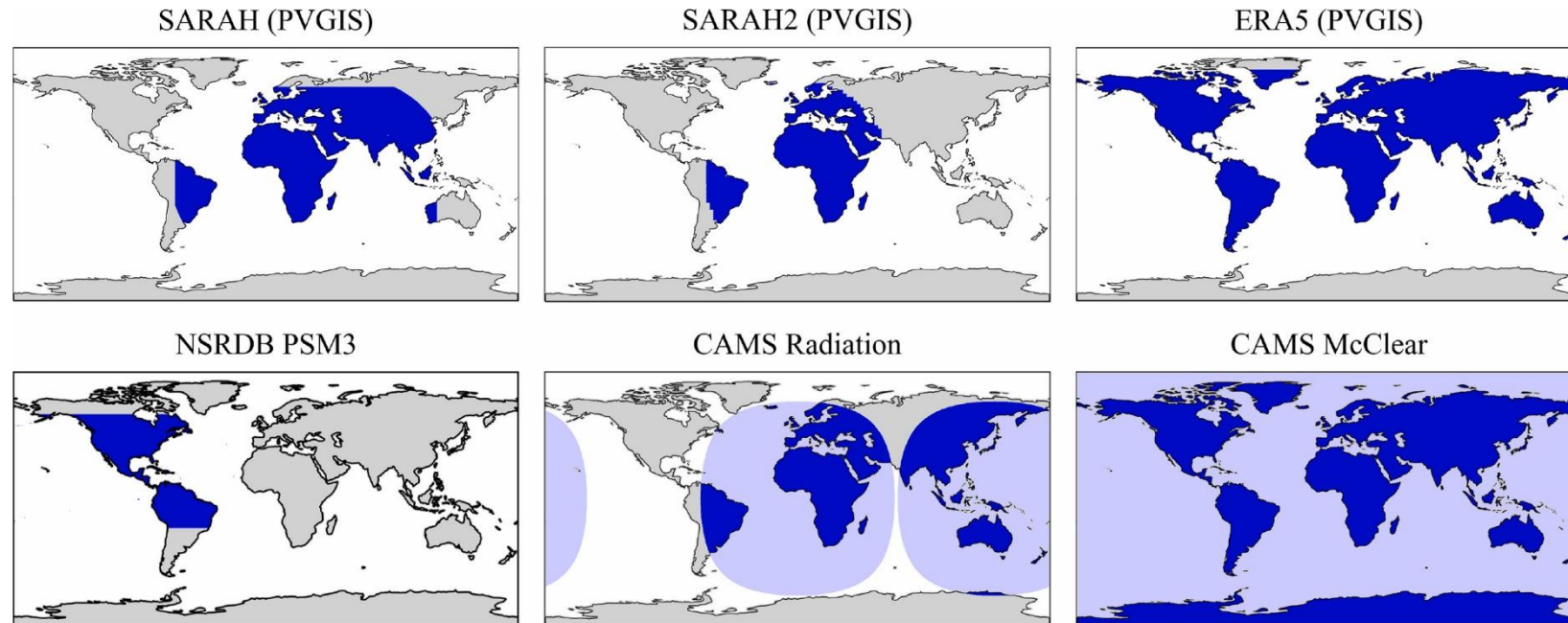
# Modeling steps

## 1. Weather data:

Irradiance, measured in  $\text{W/m}^2$ , with max instantaneous values around 1000  $\text{W/m}^2$  is obtained from:

- a. Satellite data (CAMS, NSSRDB, SolarGis...)

Some Irradiance datasets, supported by pvlib [1]



# Modeling steps

## 1. Weather data:

Irradiance, measured in  $\text{W/m}^2$ , is obtained from:

- Satellite data (CAMS, NSSRDB, SolarGis...)
- In-situ instrumentations



Pyranometer:  
Global/Inclined Irradiance

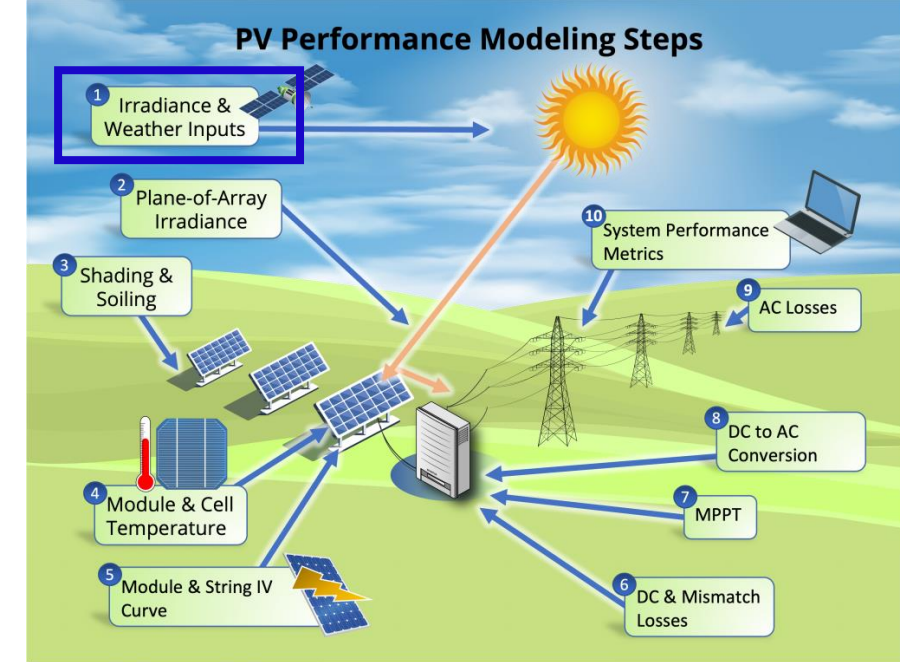


Pyrhelimeter:  
Direct irradiance



(Hukseflux product)

Pyranometer with shadow ring:  
Diffuse Irradiance





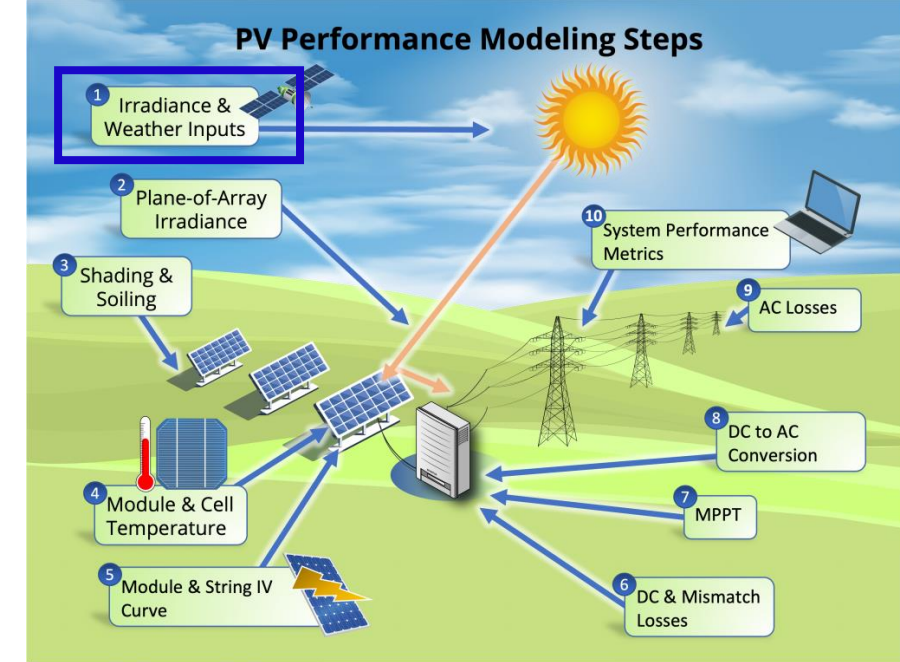
# Modeling steps

## 1. Weather data:

Irradiance, measured in  $\text{W/m}^2$ , is obtained from:

- Satellite data (CAMS, NSSRDB, SolarGis...)
- In-situ instrumentations

Reference cell: Inclined Irradiance  
(without reflection/spectral effects)



# Modeling steps

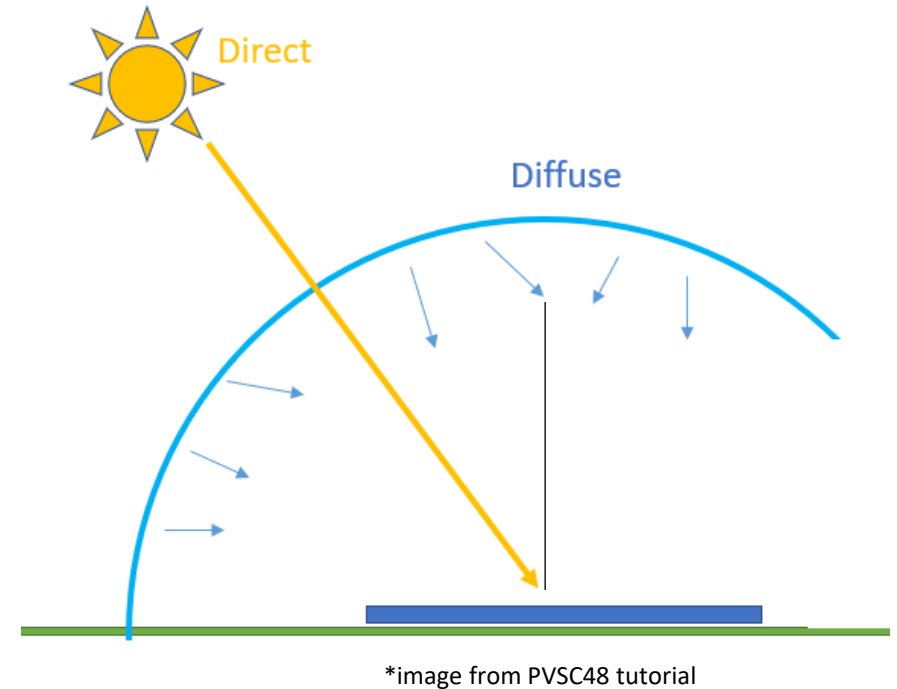
## 1. Weather data

Reminder, the Global Horizontal Irradiance (GHI) can be broken down into 2 components (ignoring reflections from other surrounding elements):

$$GHI = BHI + DHI$$

**BHI** (Beam Horizontal Irradiance): Power received from the Beam of the sun on the horizontal plan.

**DHI** (Diffuse Horizontal Irradiance): Power received from the sky diffusion of the light.



# Modeling steps

## 1. Weather data

Reminder, the Global Horizontal Irradiance (GHI) can be broken down into 2 components (ignoring reflections from other surrounding elements):

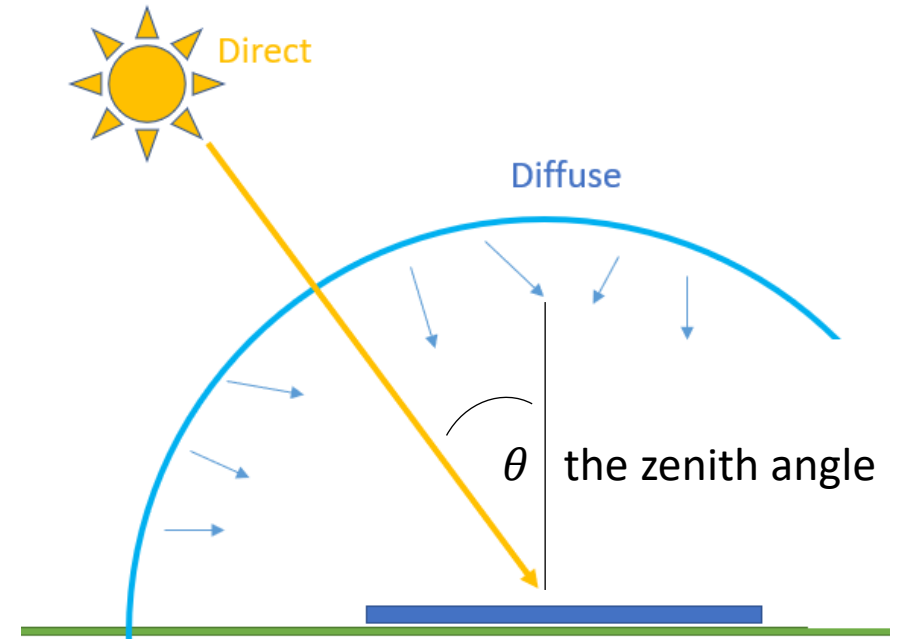
$$GHI = BHI + DHI$$

**BHI** (Beam Horizontal Irradiance): Power received from the Beam of the sun on the horizontal plan.

BHI is usually deducted from **DNI**, the Direct Normal Irradiance, which is most commonly the output of weather models.

$$BHI = DNI \cdot \cos(\theta)$$

**DHI** (Diffuse Horizontal Irradiance): Power received from the sky diffusion of the light.



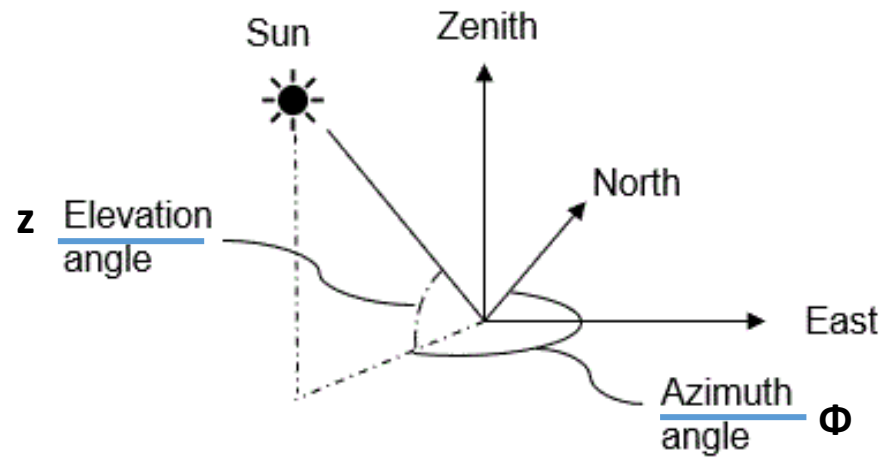
\*image adapted from PVSC48 tutorial



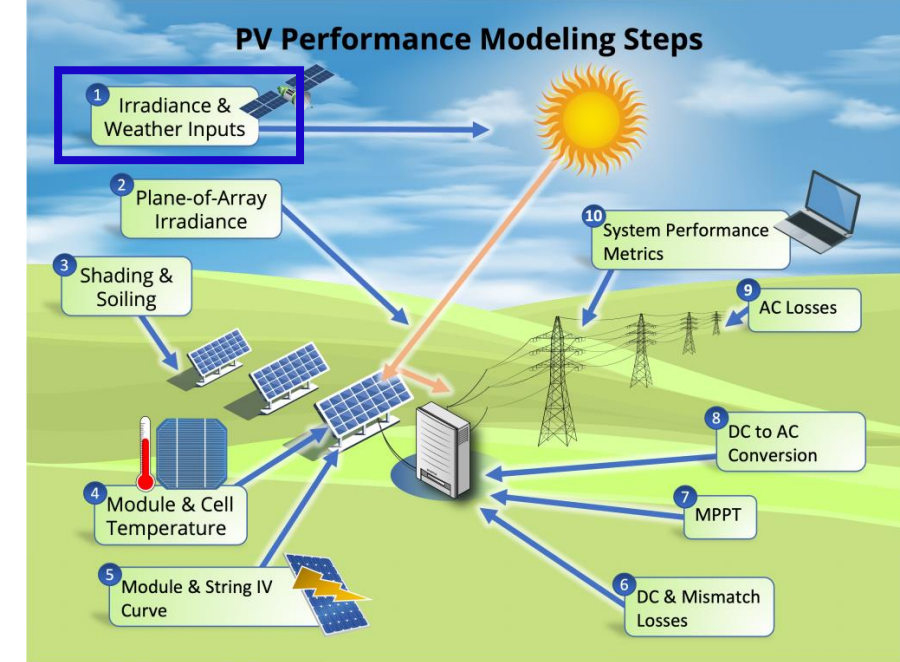
# Modeling steps

## 1. Weather data:

Sun path



azimuth angle: north=0, east=90, south=180, west=270 degree



# Modeling steps

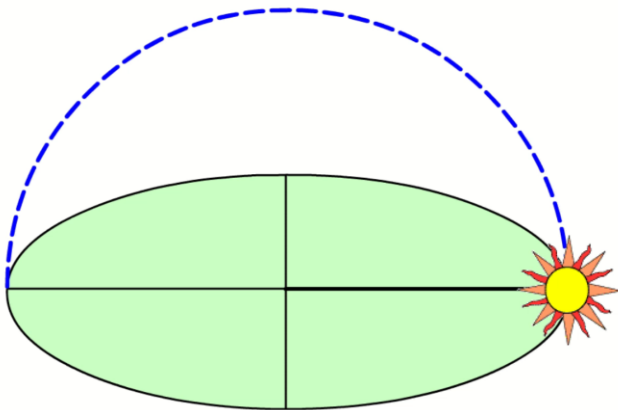
## 1. Weather data:

The sun path is characterized by:

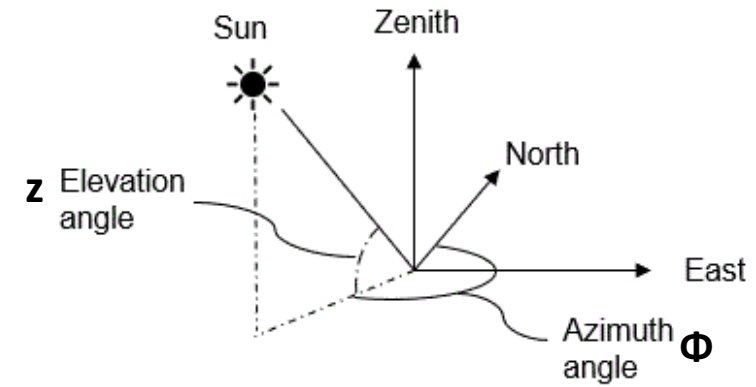
- $z$  [°], elevation angle, height of the sun in the sky

At sunrise and sunset, the elevation angle is 0.°

[Click to Start](#)



\*animation from pveducation



azimuth angle: north=0, east=90, south=180, west=270 degree

# Modeling steps

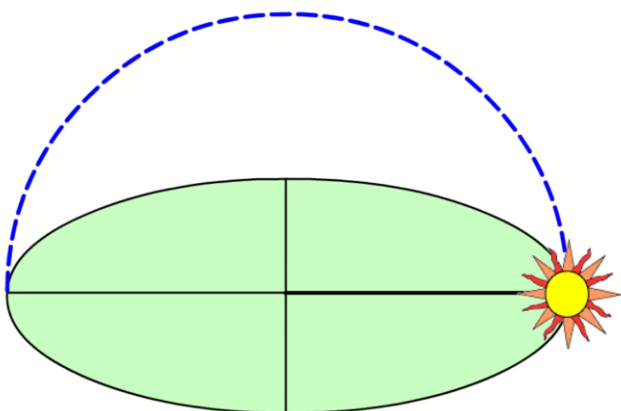
## 1. Weather data:

The sun path is characterized by:

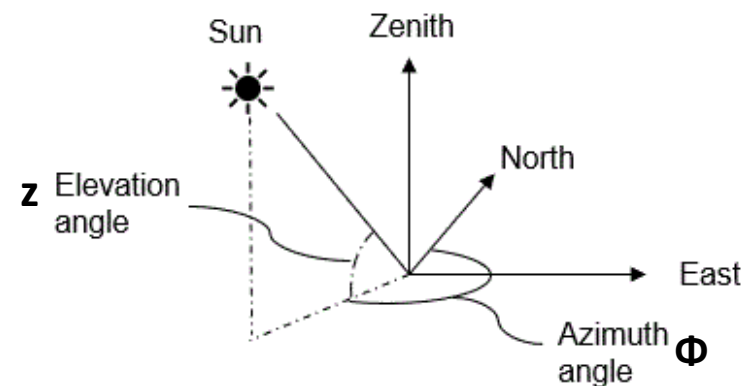
- $z$  [°], elevation angle, height of the sun in the sky

At sunrise and sunset, the elevation angle is 0.°

[Click to Start](#)



\*animation from pveducation

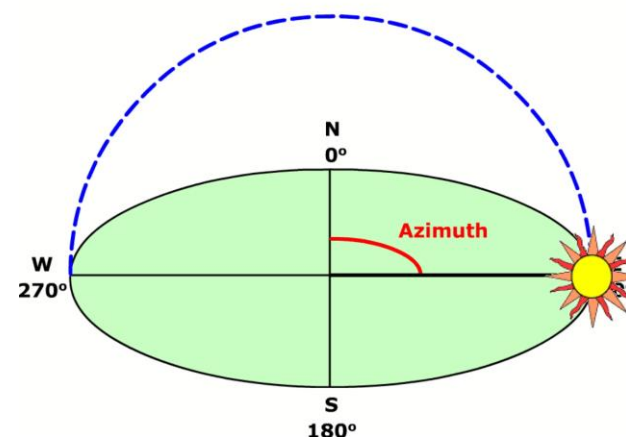


azimuth angle: north=0, east=90, south=180, west=270 degree

- $\Phi$  [°], the azimuth angle: angle of sun between north and sun direction

At sunrise on the spring and fall equinox, the azimuth angle is 90.°

[Click to Start](#)



\*animation from pveducation

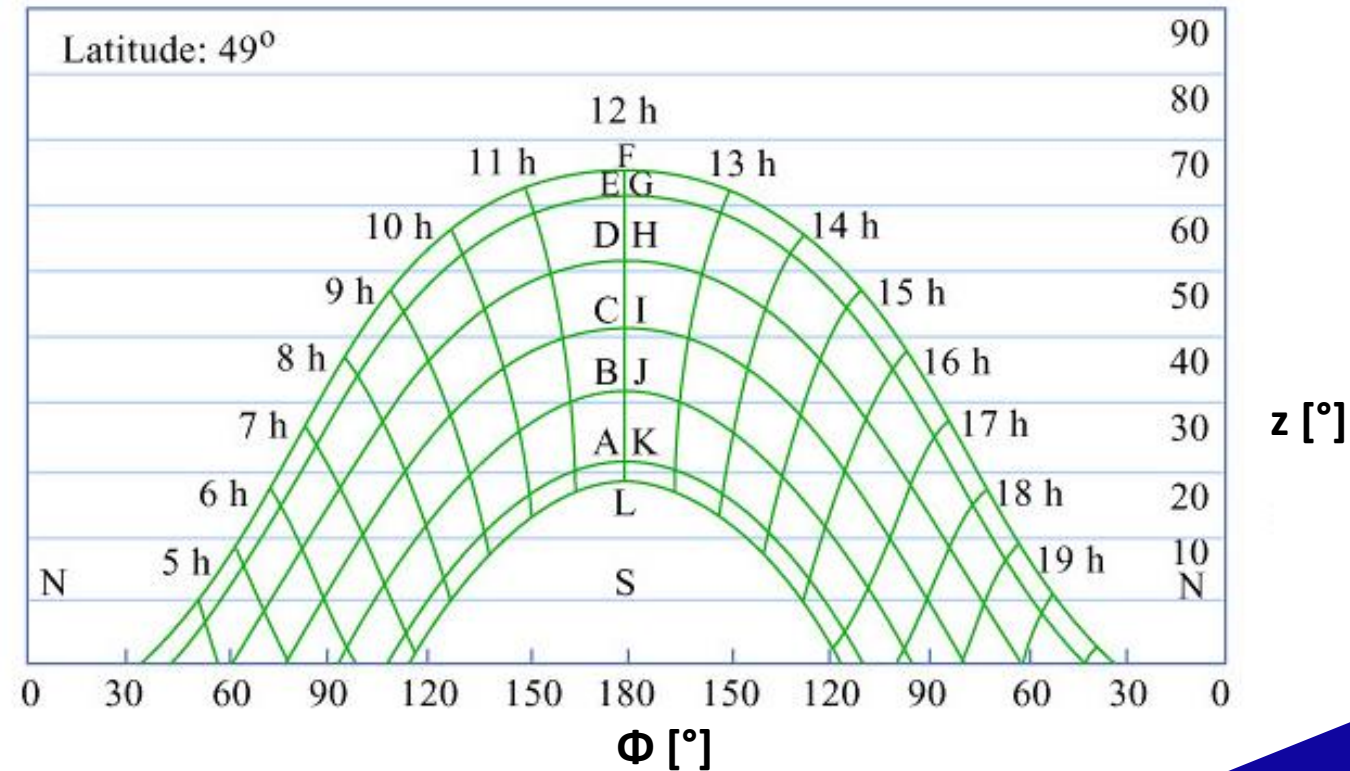
## Modeling steps

### 1. Weather data:

The sun path is characterized by:

- $z$  [°], elevation angle, height of the sun in the sky
- $\Phi$  [°], the azimuth angle: angle between north and the sun direction

Typical sun path diagram



# Modeling steps

## 1. Weather data:

The sun path is characterized by:

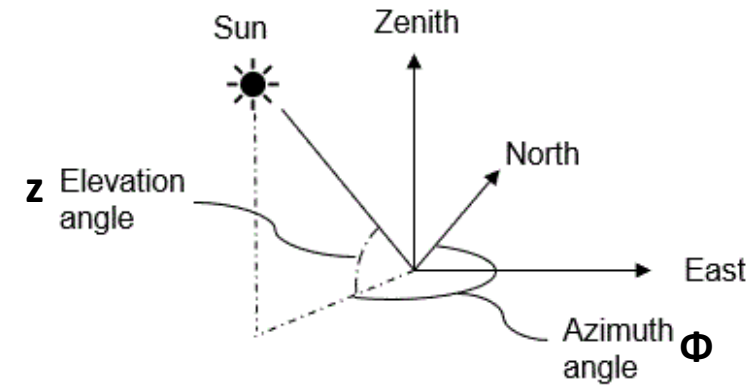
- $z$  [°], elevation angle, height of the sun in the sky
- $\Phi$  [°], the azimuth angle: angle between north and the sun direction

Remember ? To calculate BHI:

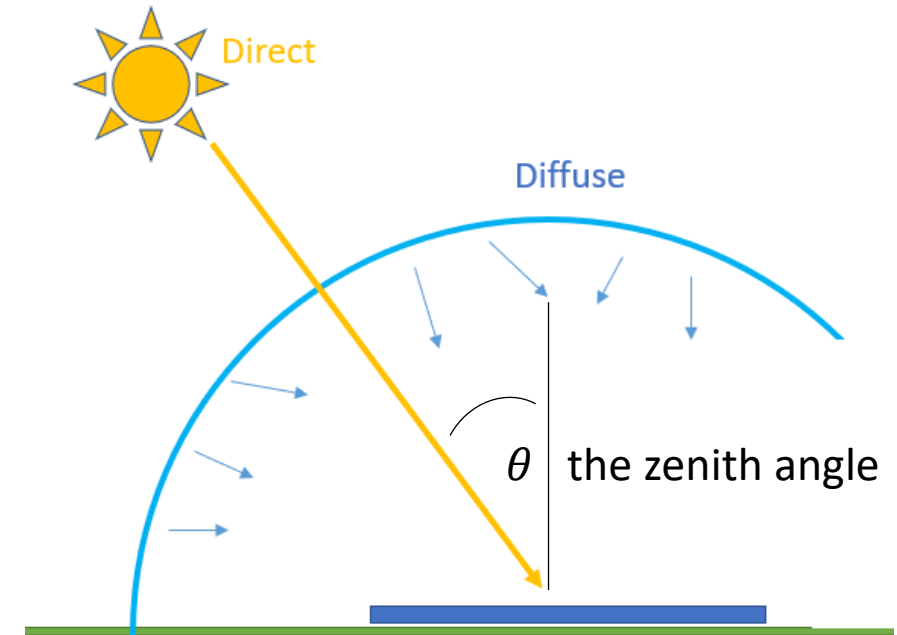
$$BHI = DNI \cdot \cos(\theta)$$

With  $\theta$ , the zenith angle

$$\theta = 90^\circ - z$$



azimuth angle: north=0, east=90, south=180, west=270 degree

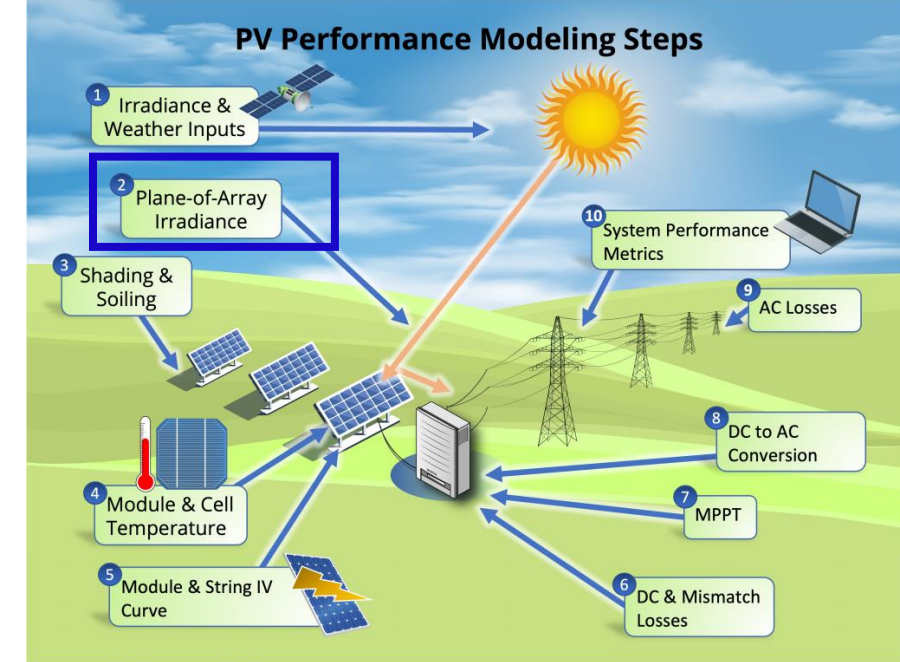


\*image adapted from PVSC48 tutorial

# Modeling steps

## 2. Plan-of-Array Irradiance

*Reminder, the Global Plane-Of-Array (GPOA) irradiance can be calculated from **DHI**, **GHI** and **DNI**.*





# Modeling steps

## 2. Plan-of-Array Irradiance

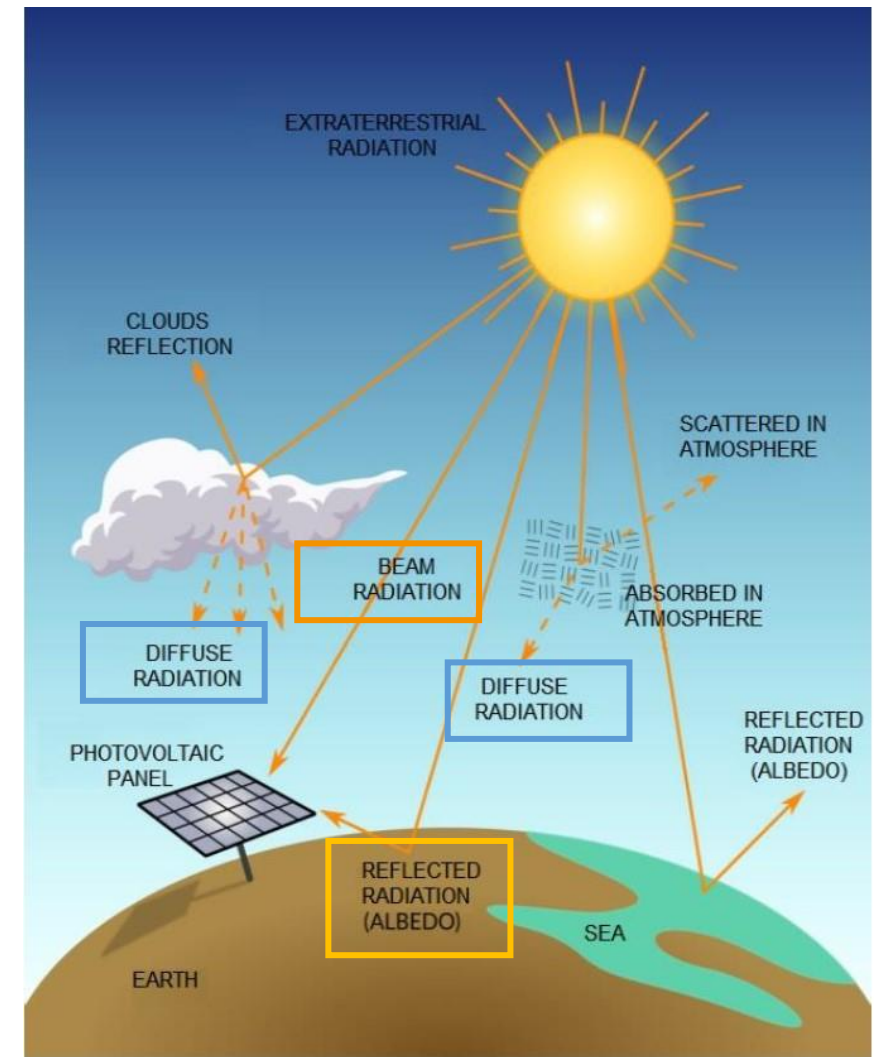
Reminder, the Global Plane-Of-Array (GPOA) irradiance can be calculated from **DHI**, **GHI** and **DNI** and are broken down into 3 components:

$$GPOA = POA_b + POA_d + POA_{grd}$$

**$POA_b$** : The direct component - Power received from the Beam of the sun on the horizontal plan

**$POA_d$** : The diffuse component - Power received from the sky diffusion of the light,

**$POA_{grd}$** : The ground-reflected component - Power received from reflections from the ground



Souza, Muriele et al. (2019). Determination of Diffused Irradiation from Horizontal Global Irradiation - Study for the City of Curitiba.

# Modeling steps

## 2. Plan-of-Array Irradiance

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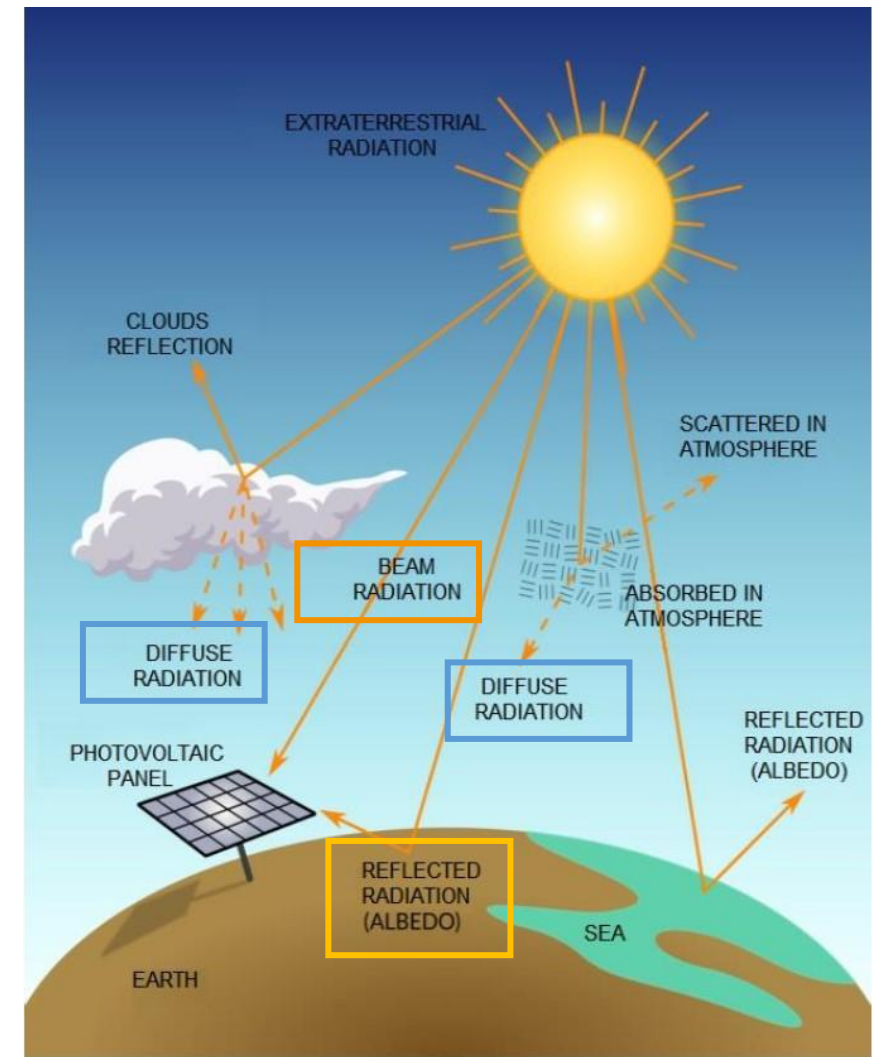
$$GPOA = POA_b + POA_d + POA_{grd}$$

- $POA_b = DNI \cdot \cos(AOI)$

Under the isotropic/no-shading assumption

- $POA_d = DHI \cdot \frac{1 + \cos(\beta)}{2}$

- $POA_{grd} = GHI \cdot \rho \cdot \frac{1 - \cos(\beta)}{2}$



Souza, Muriele et al. (2019). Determination of Diffused Irradiation from Horizontal Global Irradiation - Study for the City of Curitiba.

With:

- $\beta$ , PV installation tilt



# Modeling steps

## 2. Plan-of-Array Irradiance

*Reminder*, the Global Plane-Of-Array (GPOA) irradiance can be calculated from **DHI**, **GHI** and **DNI** and are broken down into 3 components:

$$GPOA = POA_b + POA_d + POA_{grd}$$

- $POA_b = DNI \cdot \cos(AOI)$   $\longleftrightarrow \cos(AOI) = [\cos(\beta) \cdot \sin(z) + \sin(\beta) \cdot \cos(z) \cdot \cos(\Phi - \Phi_{install})]$

Under the isotropic/no-shading assumption

- $POA_d = DHI \cdot \frac{1+\cos(\beta)}{2}$
- $POA_{grd} = GHI \cdot \rho \cdot \frac{1-\cos(\beta)}{2}$

With:

- $AOI$ , angle of incidence
- $\beta$ , PV installation tilt
- $\Phi_{install}$ , PV installation azimuth
- $z$ , Sun elevation
- $\Phi$ , sun azimuth

# Modeling steps

## 2. Plan-of-Array Irradiance

Time for some  
hands-on exercises



Use the notebook:

<https://github.com/AlexandreHugoMathieu/pv-fault-detection-solar-academy/blob/master/notebooks/python-intro-poa.ipynb>

Follow the python tutorial and calculate the 3 POA components for:

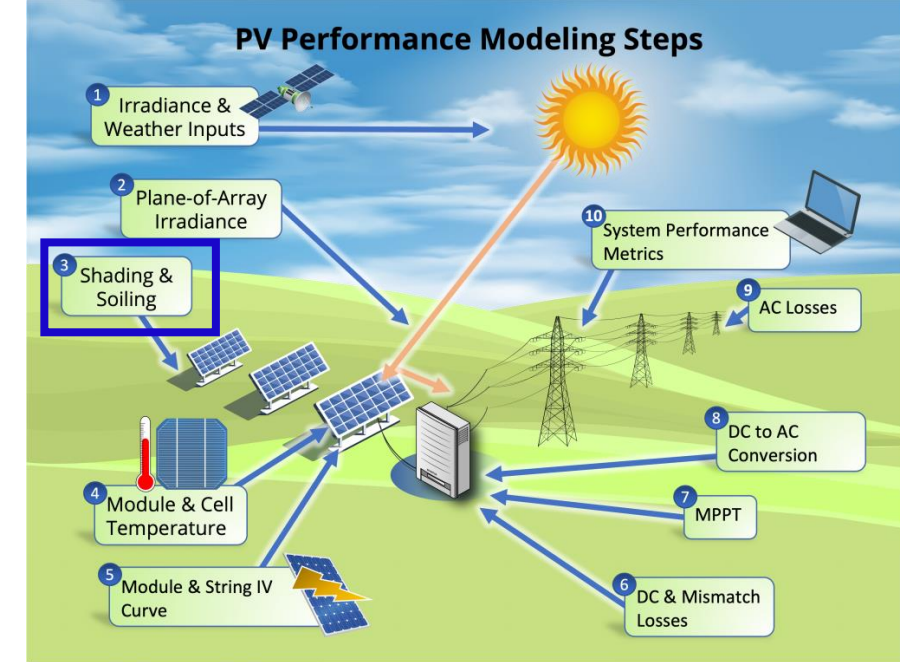
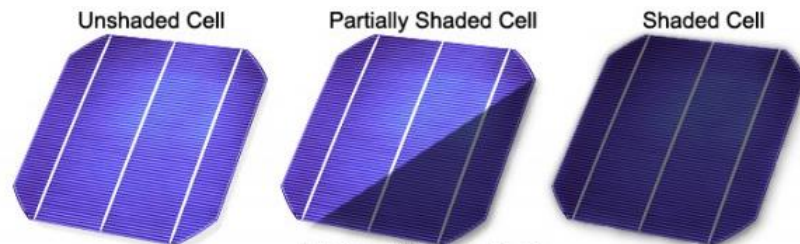
- One timestep
- Over one year

## 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.

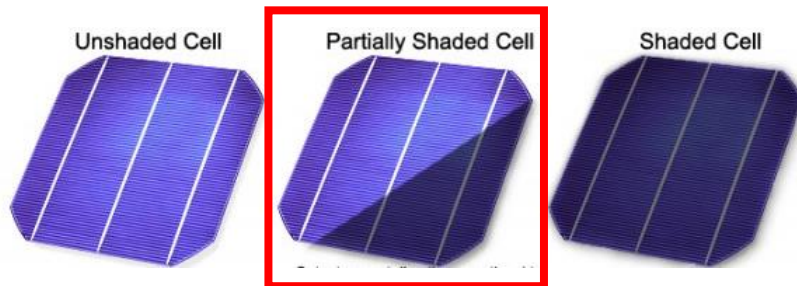


# Modeling steps

## 3. Shading

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

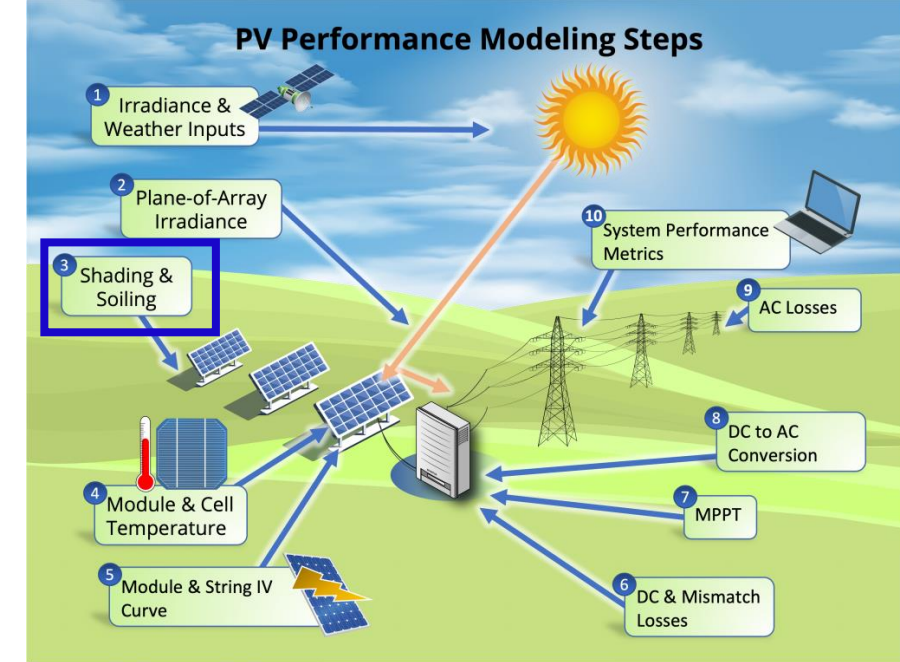
1. **Partial shading** refers to a condition where some but not all of the solar cells or panels in a PV array are exposed to sunlight while others are shaded.
2. **Full shading** occurs when the entire PV array is covered and deprived of direct sunlight.



Not treated in this course but, in average, if a PV panel is around 8% covered, the whole PV panel is bypassed and then, do not produce.

Demonstrated in one of pvlib example:

[https://pvlib-python.readthedocs.io/en/stable/gallery/shading/plot\\_partial\\_module\\_shading\\_simple.html](https://pvlib-python.readthedocs.io/en/stable/gallery/shading/plot_partial_module_shading_simple.html)

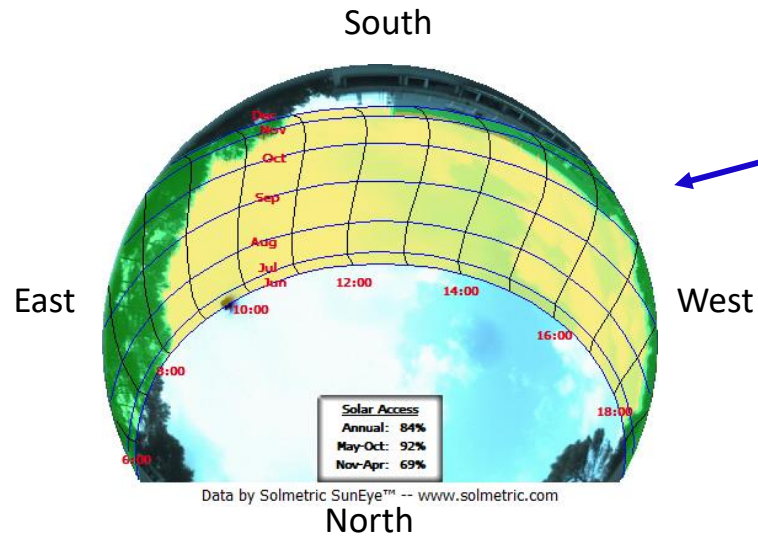




## Modeling steps

### 3. Shading

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



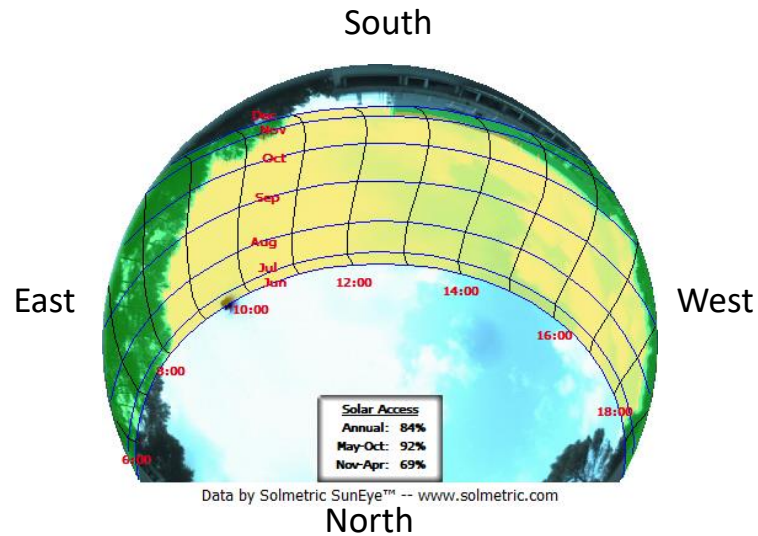
*Example in South of France*



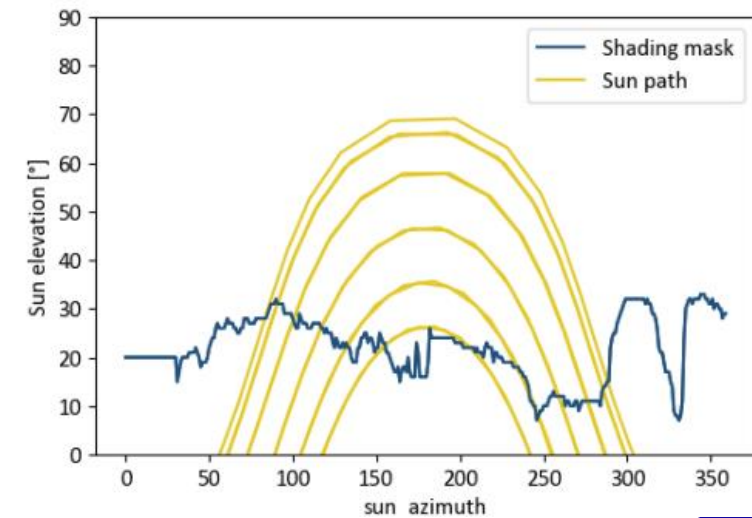
## Modeling steps

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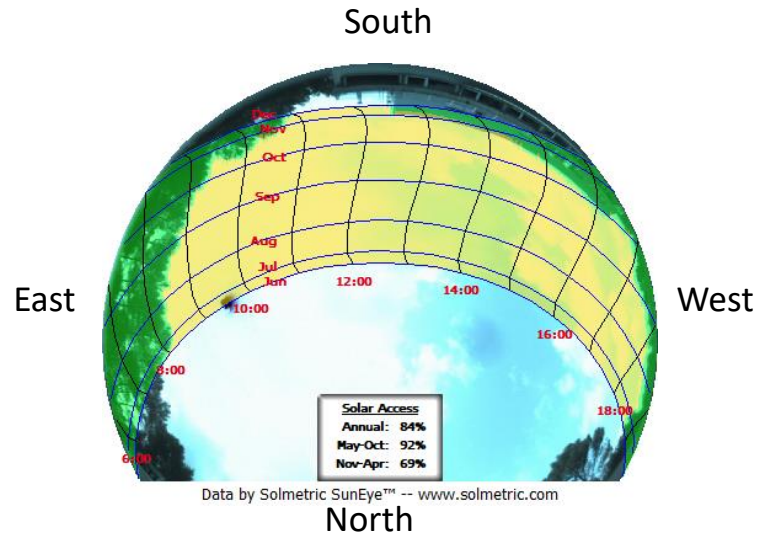
*Example in South of France*



# Modeling steps

## 3. Shading

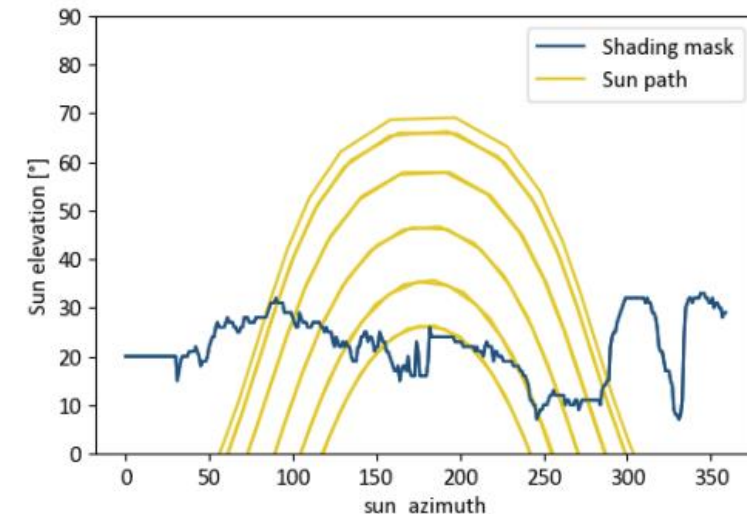
The Fisheye camera enables to account for the in-situ (full) shading in PV modeling.



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



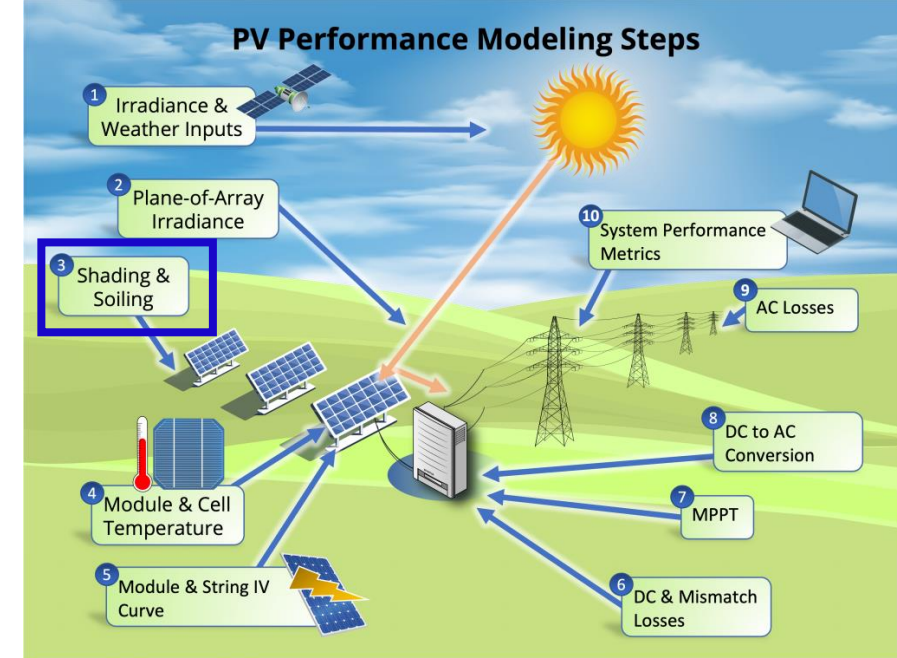


# Modeling steps

## 3. Soiling losses

Big subject in the PV industry:

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





## Modeling steps

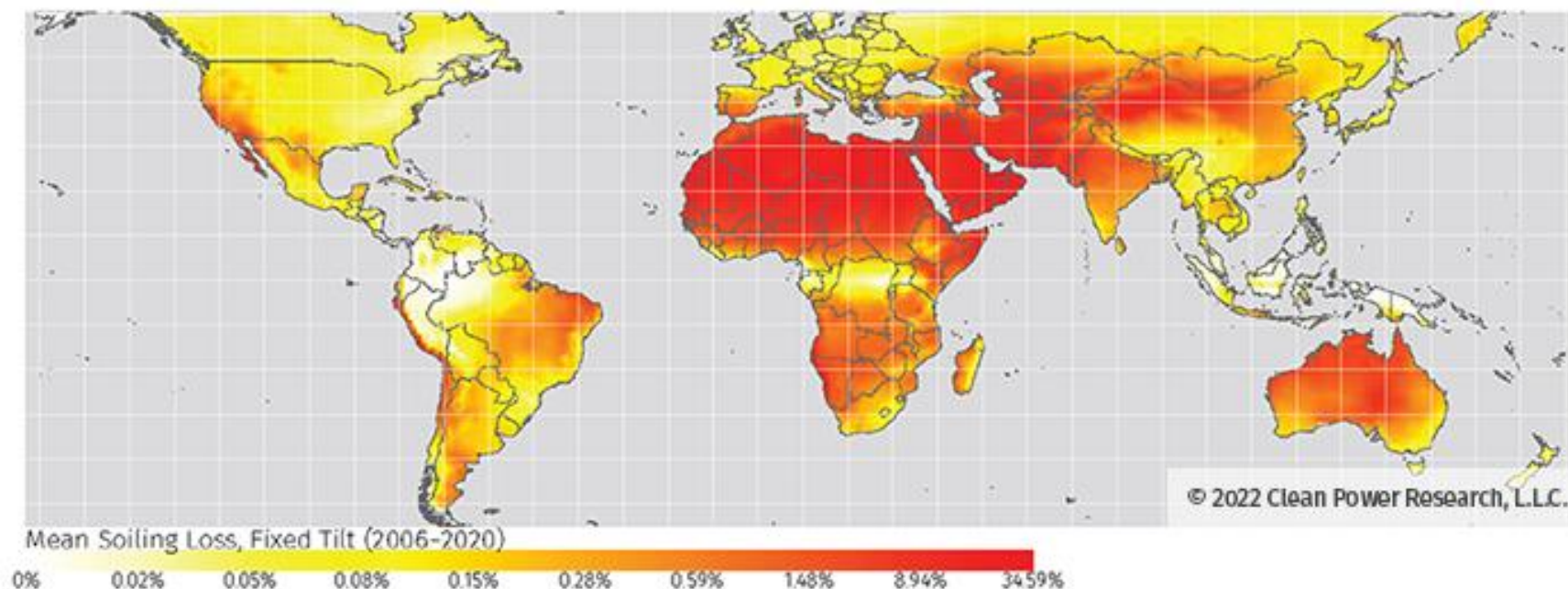
### 3. Soiling losses

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:

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

Go to: [https://re.jrc.ec.europa.eu/pvg\\_tools/en/](https://re.jrc.ec.europa.eu/pvg_tools/en/)

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/](https://re.jrc.ec.europa.eu/pvg_tools/en/)

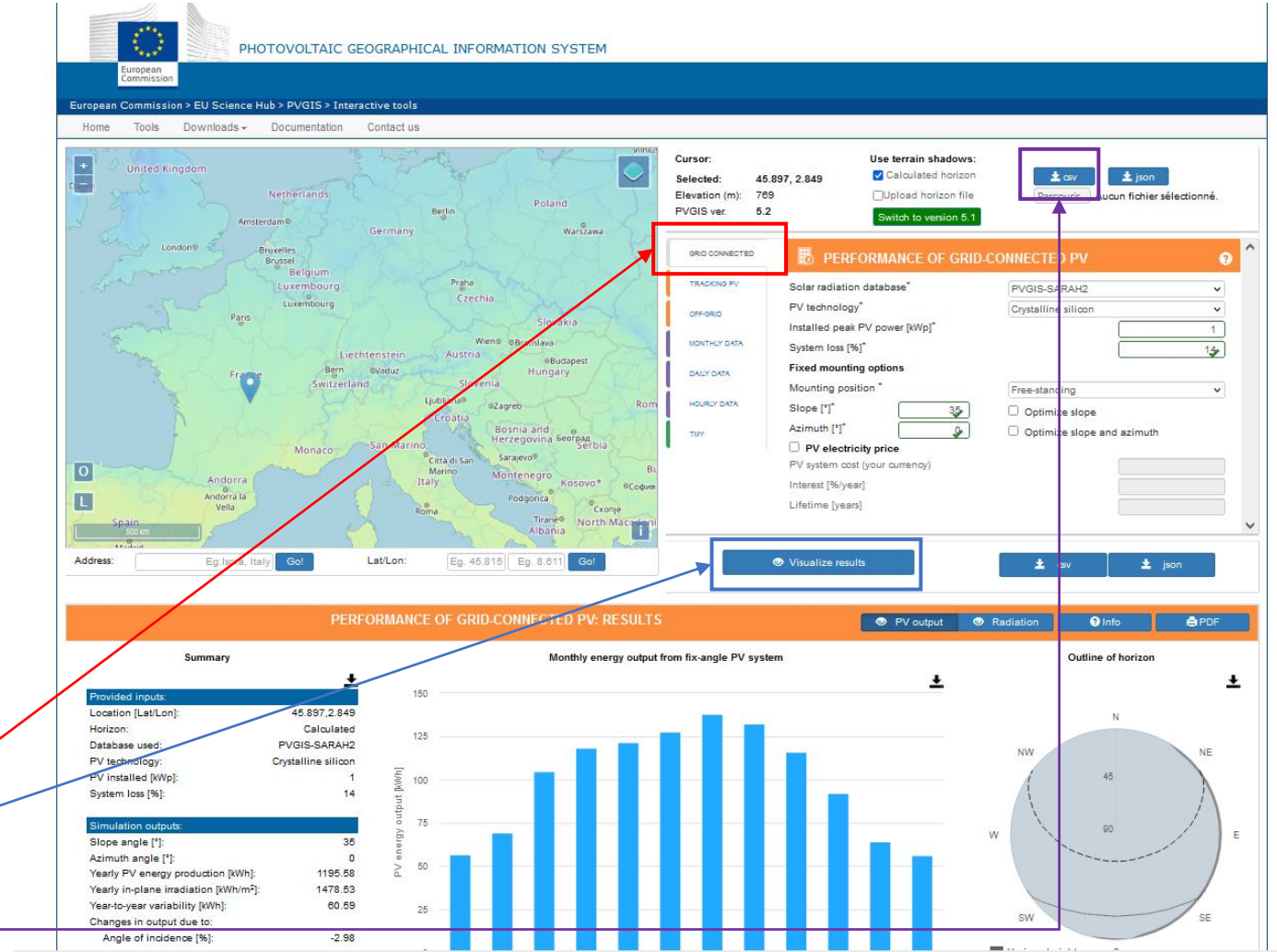
- Enables to extract the horizon mask with a Digital Surface Model (DSM).

## 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

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

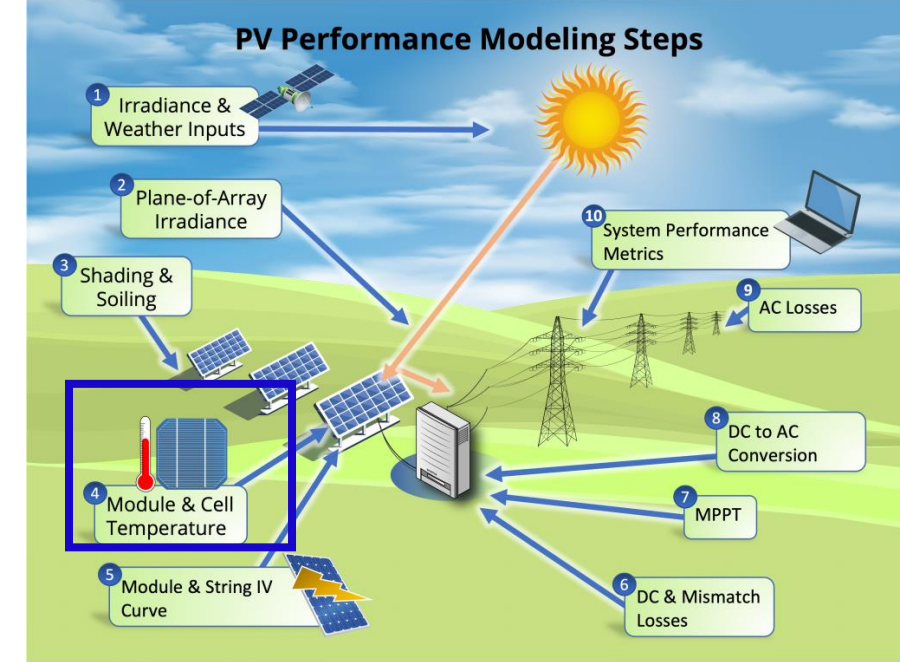
[https://github.com/AlexandreHugoMathieu/pvfault\\_detection\\_solar\\_academy/blob/master/notebooks/python\\_intro2\\_horizon\\_mask.ipynb](https://github.com/AlexandreHugoMathieu/pvfault_detection_solar_academy/blob/master/notebooks/python_intro2_horizon_mask.ipynb)



# Modeling steps

## 4. Module and Cell temperature

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





# Modeling steps

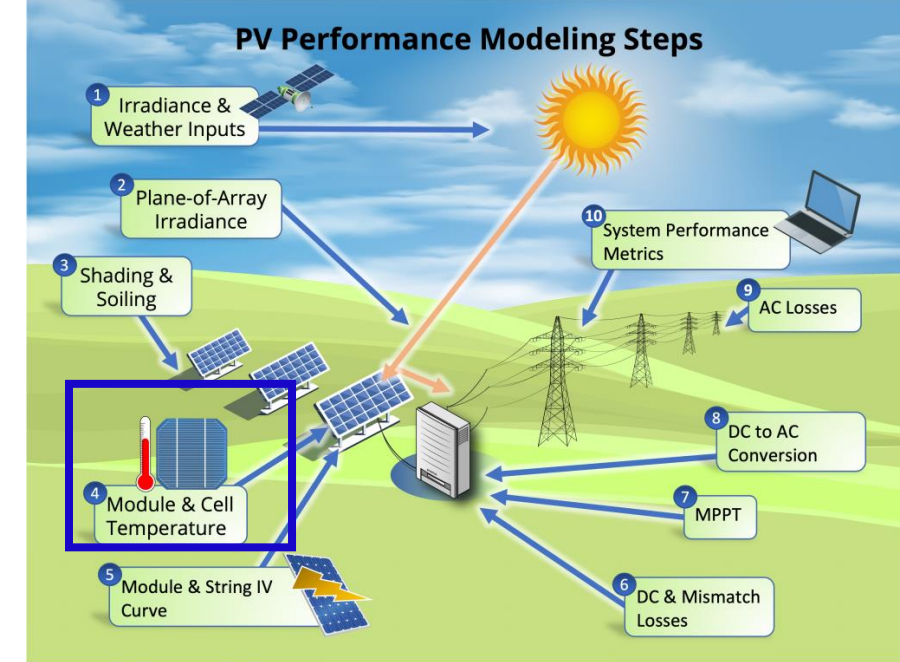
## 4. Cell temperature

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

### 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].

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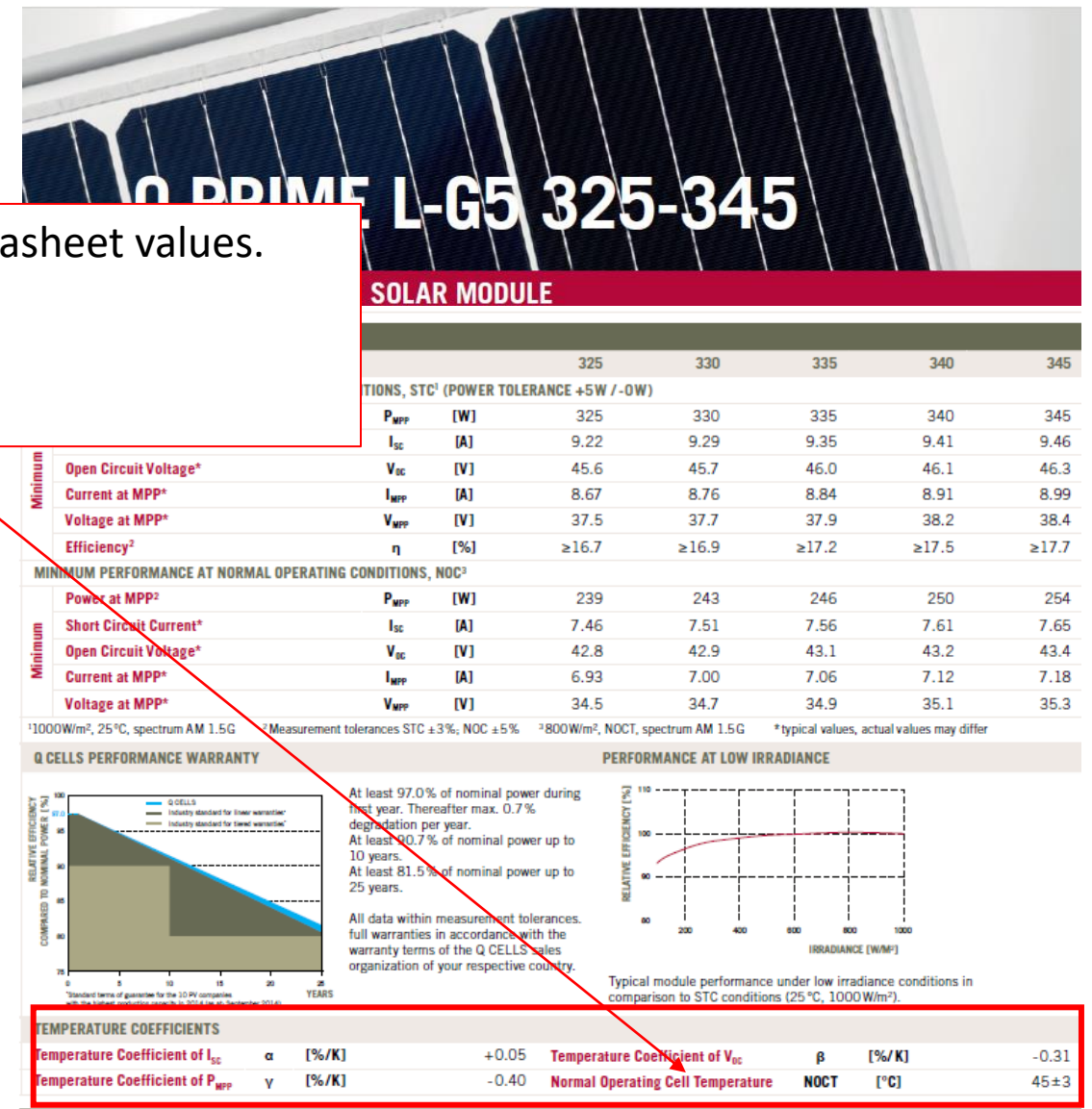
$k_{ROSS}$ , typically in the range 0.02-0.05 K/m²/W.

$k_{ROSS}$  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

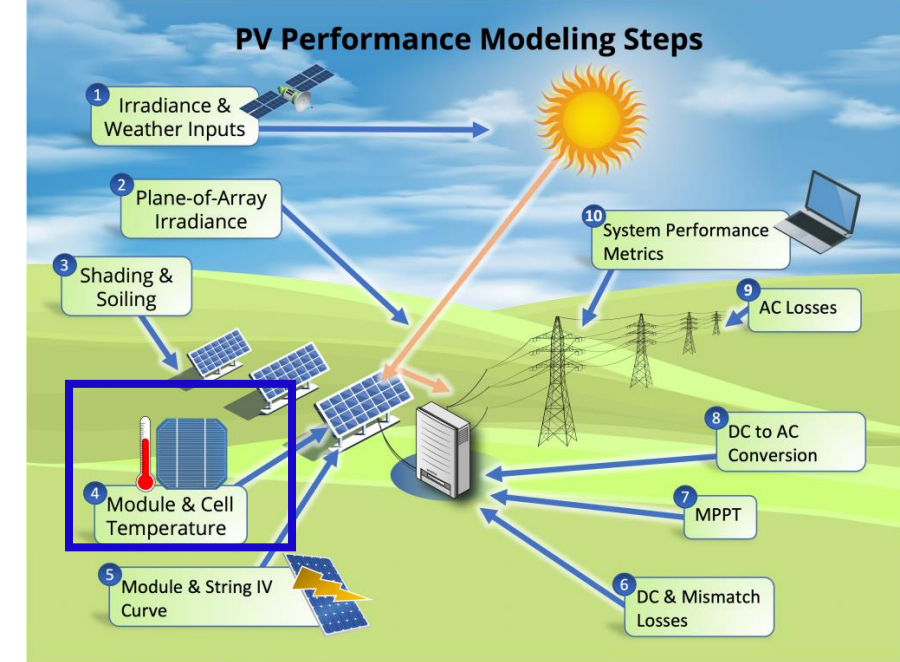
#### Faiman model:

Model to estimate the cell temperature  $T_c$  [°C] as function of ambient temperature and irradiance  $G_{POA}$  [W/m<sup>2</sup>] 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

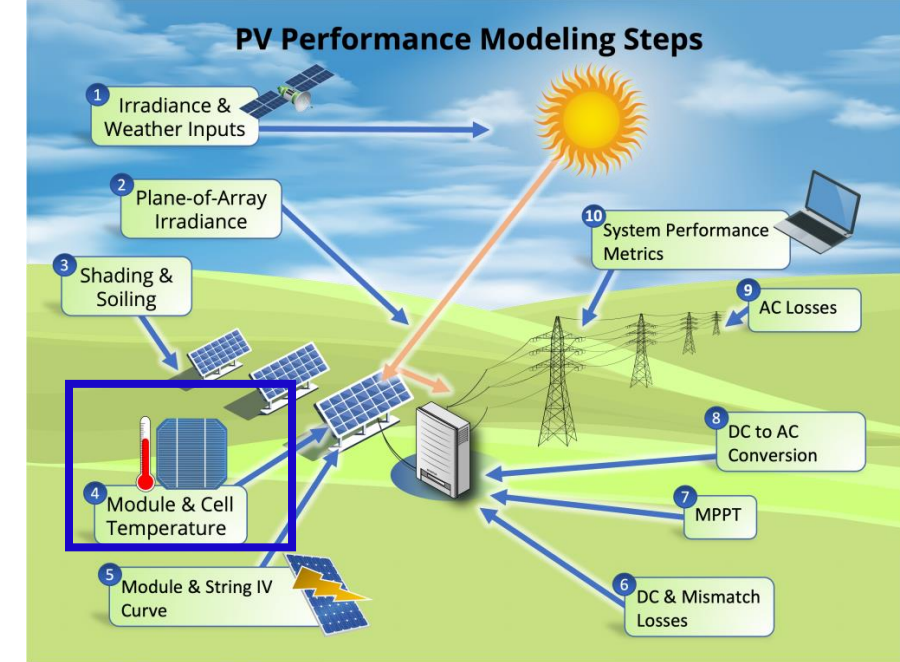
#### Faiman model:

Model to estimate the cell temperature  $T_c$  [°C] as function of ambient temperature and irradiance  $G_{POA}$  [W/m<sup>2</sup>] 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 [ $\frac{W}{Km^2(\frac{m}{s})}$ ]



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

## Modeling steps

### 4. Cell temperature

Use the following notebook:

[https://github.com/AlexandreHugoMathieu/pvfault\\_detection\\_solar\\_academy/blob/master/notebooks/dc\\_power\\_estimation.ipynb](https://github.com/AlexandreHugoMathieu/pvfault_detection_solar_academy/blob/master/notebooks/dc_power_estimation.ipynb)

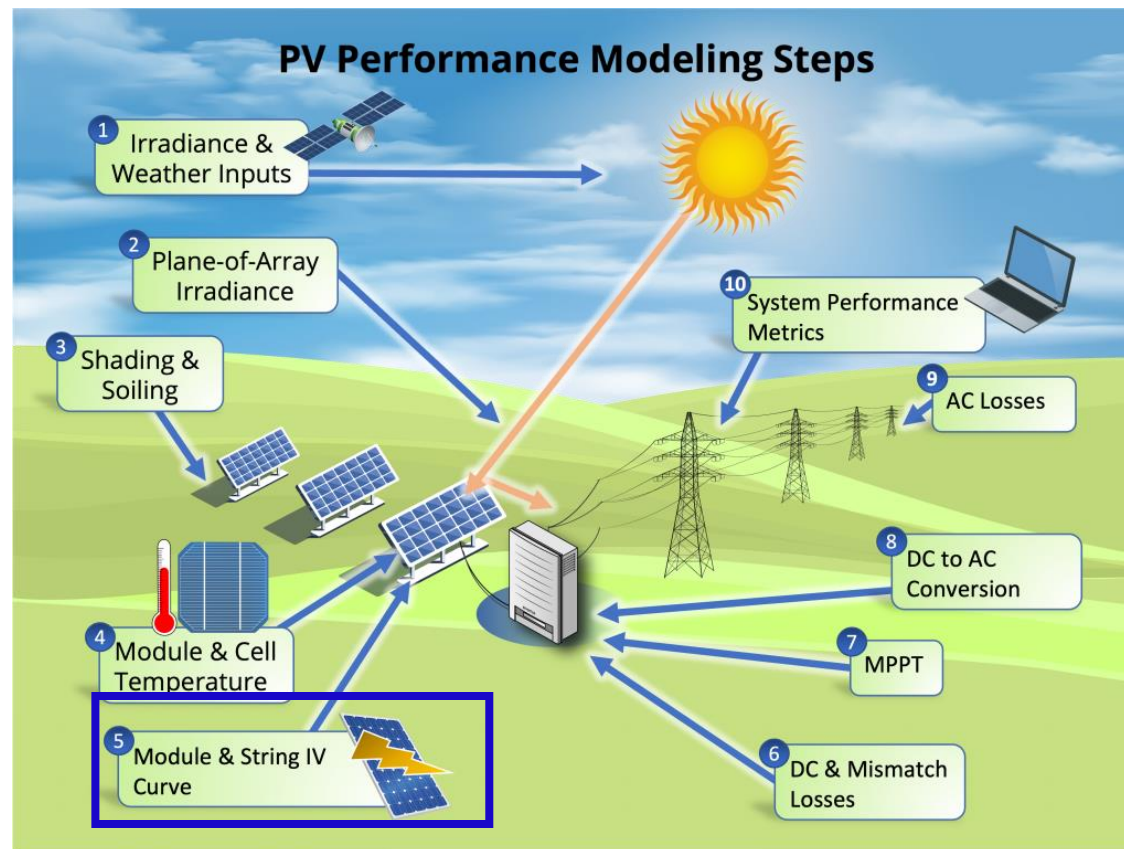
Follow the python tutorial and estimate the cell temperature for one year.

Time for some  
hands-on exercises !



# Modeling steps

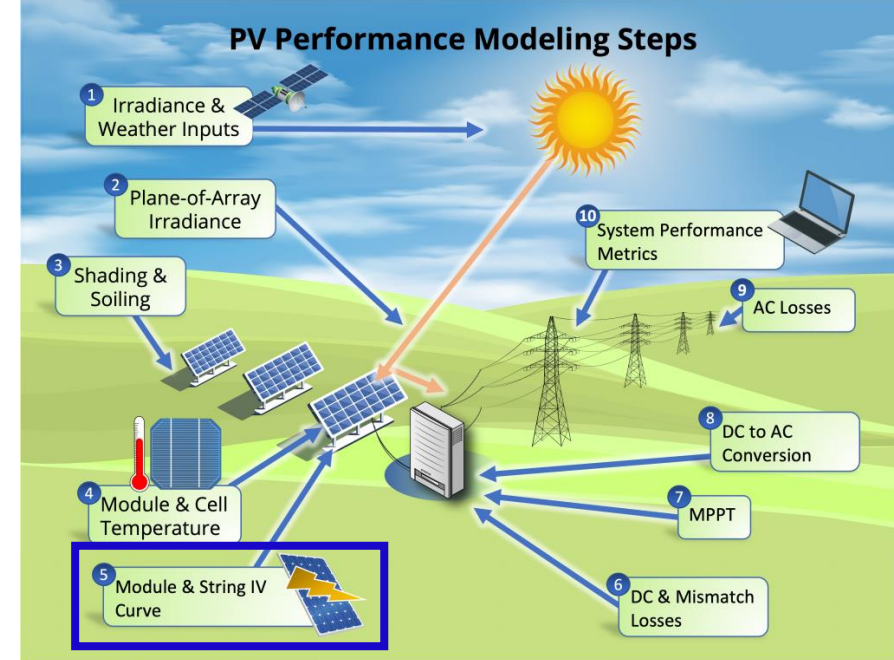
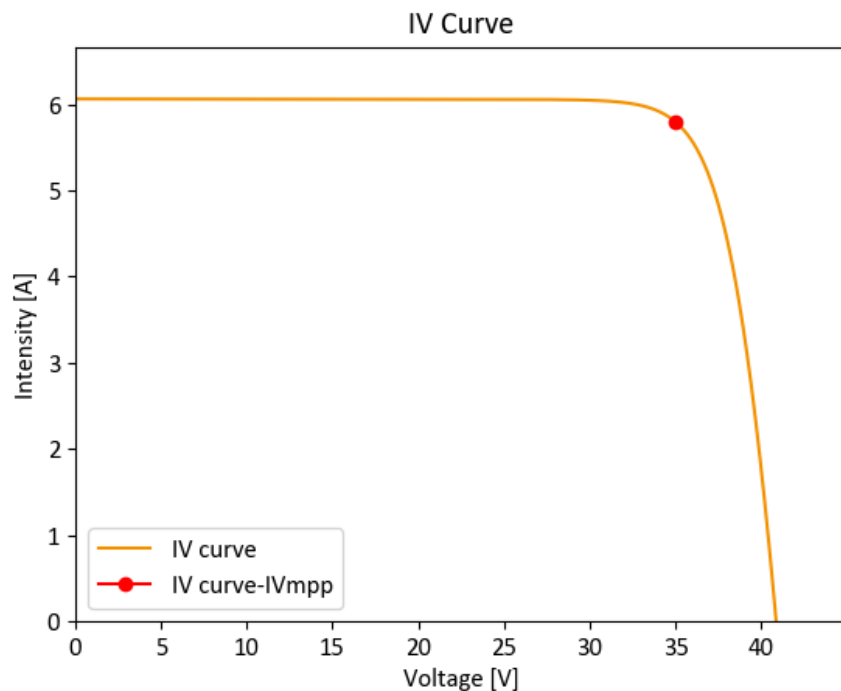
## 5. Module and String IV Curve



# Modeling steps

## 5. Module and String IV Curve

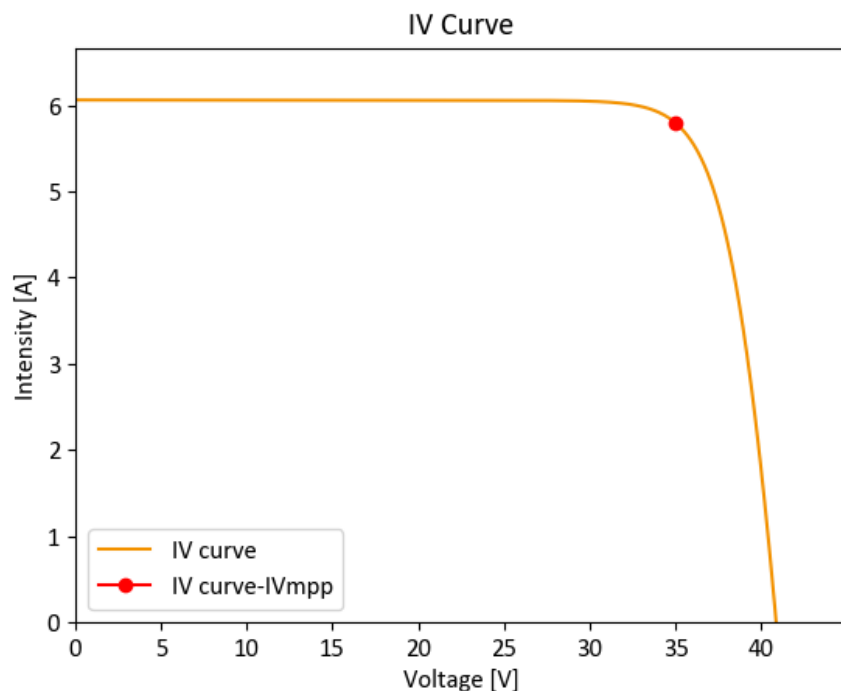
For a fixed irradiance and module temperature, the PV module has its  $I$ , current which depends on  $V$ , voltage and it can take many operating points.



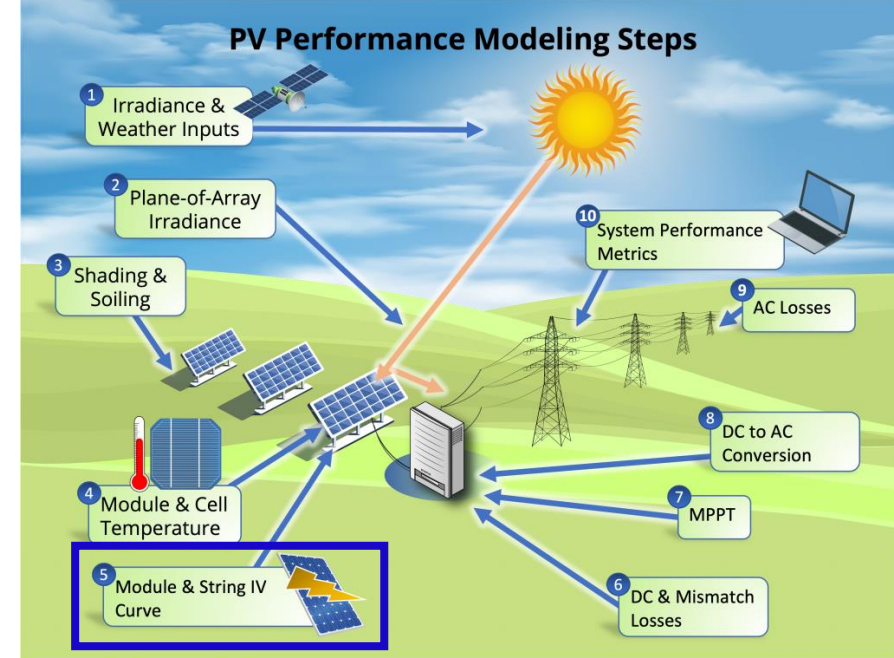
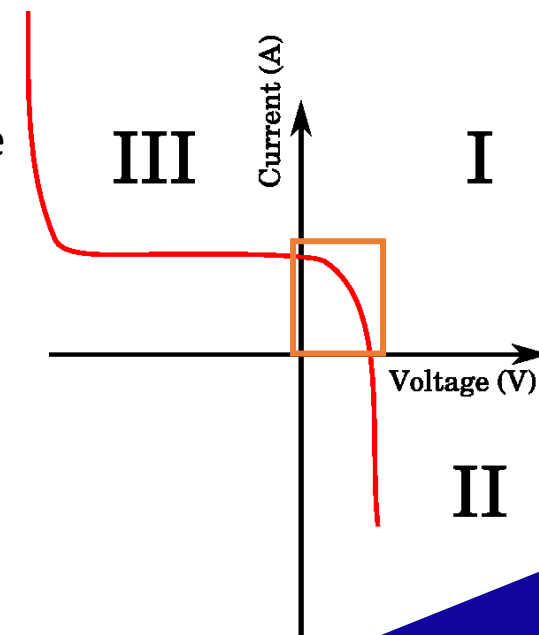
# Modeling steps

## 5. Module and String IV Curve

For a fixed irradiance and module temperature, the PV module has its  $I$ , current which depends on  $V$ , voltage and it can take many operating points.



In reality, the IV characteristics go out of the 1st quadrant and the module can potentially consume power.

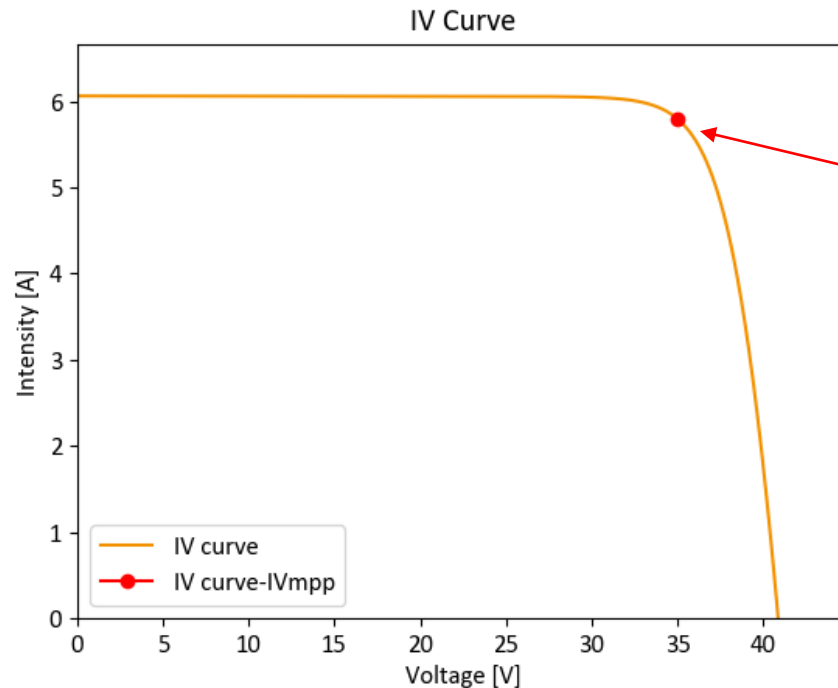




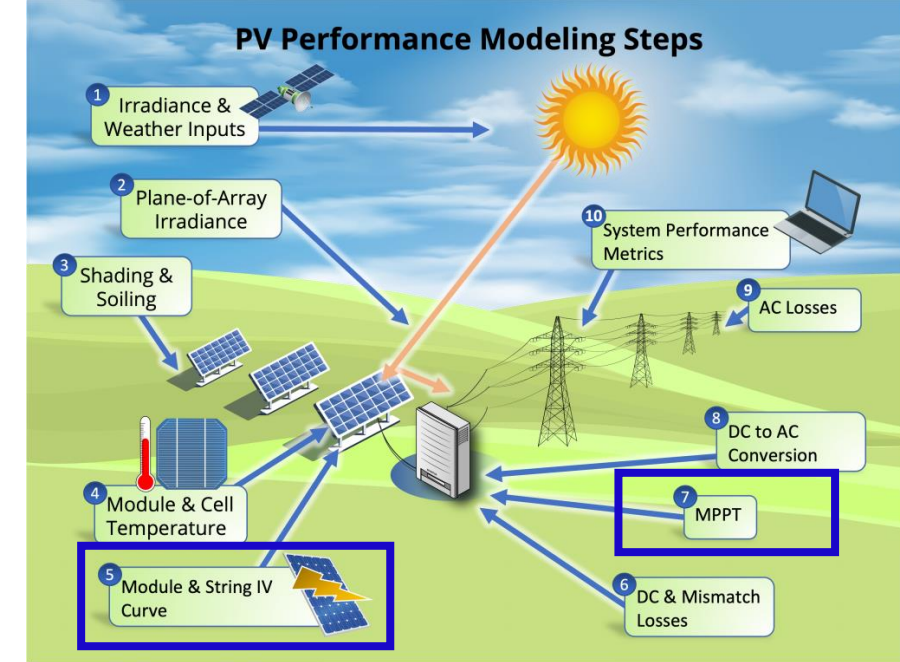
# Modeling steps

## 5. Module and String IV Curve

For a fixed irradiance and module temperature, the PV module has its  $I$ , current which depends on  $V$ , voltage and it can take many operating points



Then, the inverter is constantly searching for the operating point which maximizes the power MPP: Maximum Power Point.

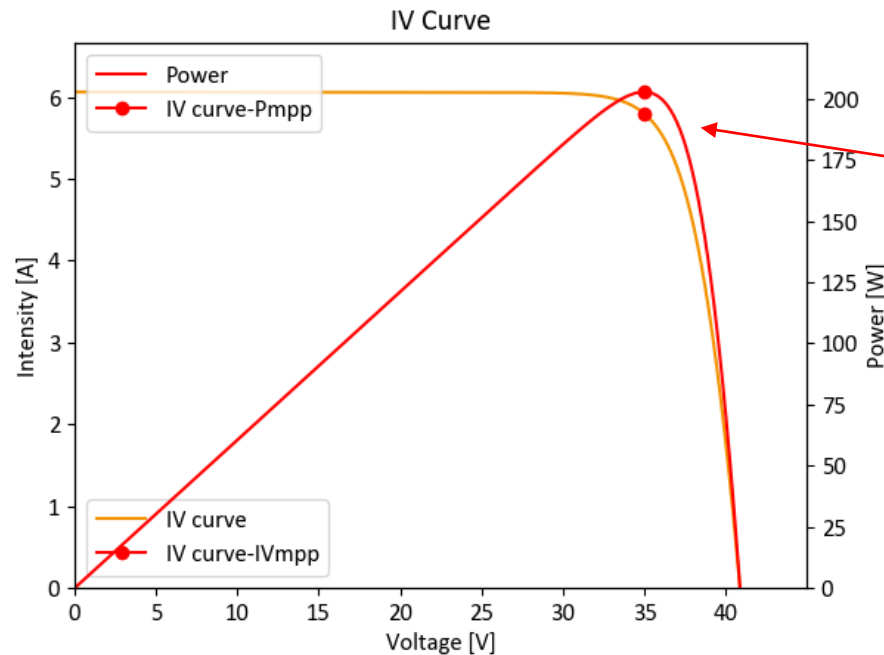




# Modeling steps

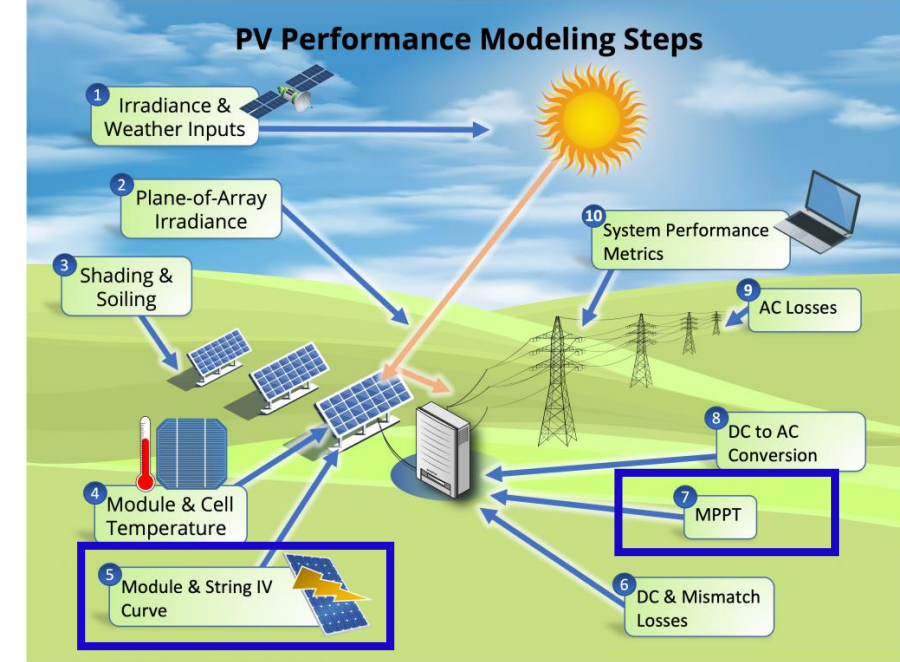
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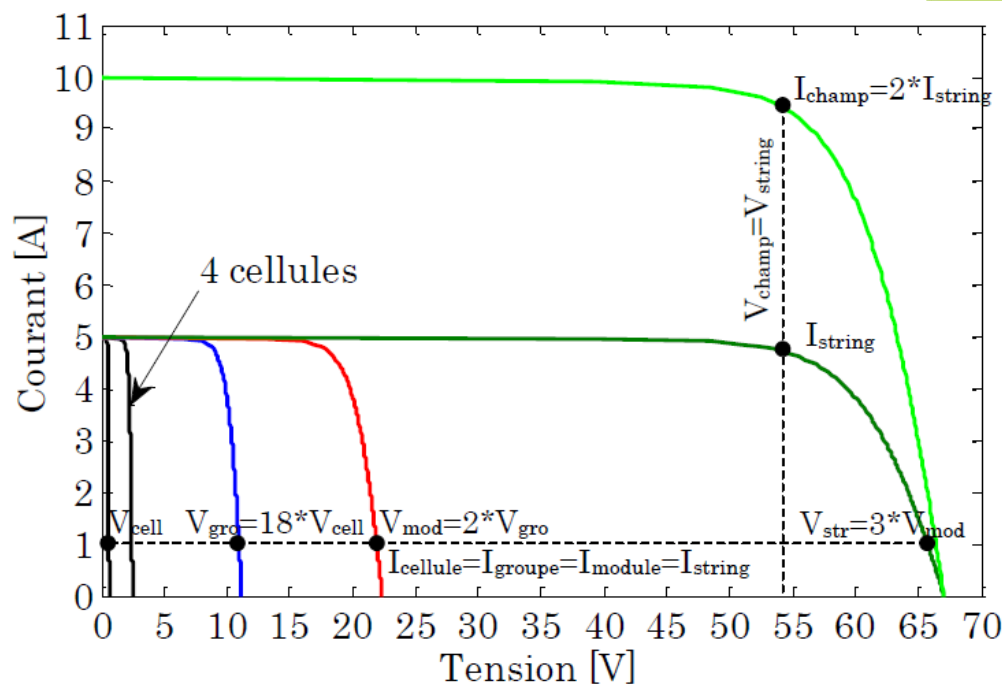
Especially, it changes the voltage with the MPP-Tracker (MPPT) to maximize power.



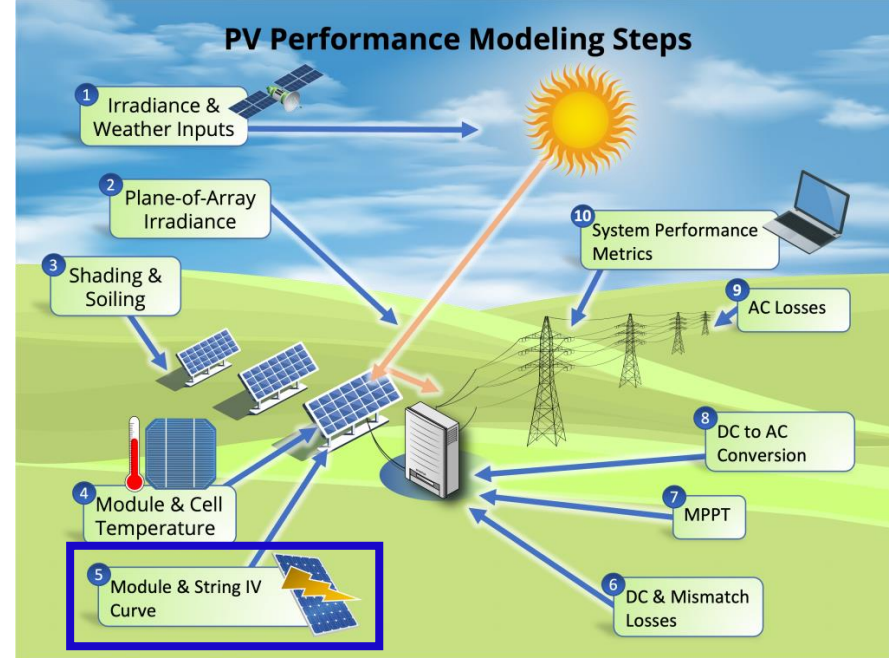
# Modeling steps

## 5. Module and String IV Curve

By the way... the IV curves can be summed up when the modules are connected in series or parallel! The inverter, then, maximizes the power of the PV array IV curve.



Bun, L.. "Détection et localisation de défauts pour un système PV." (2011).



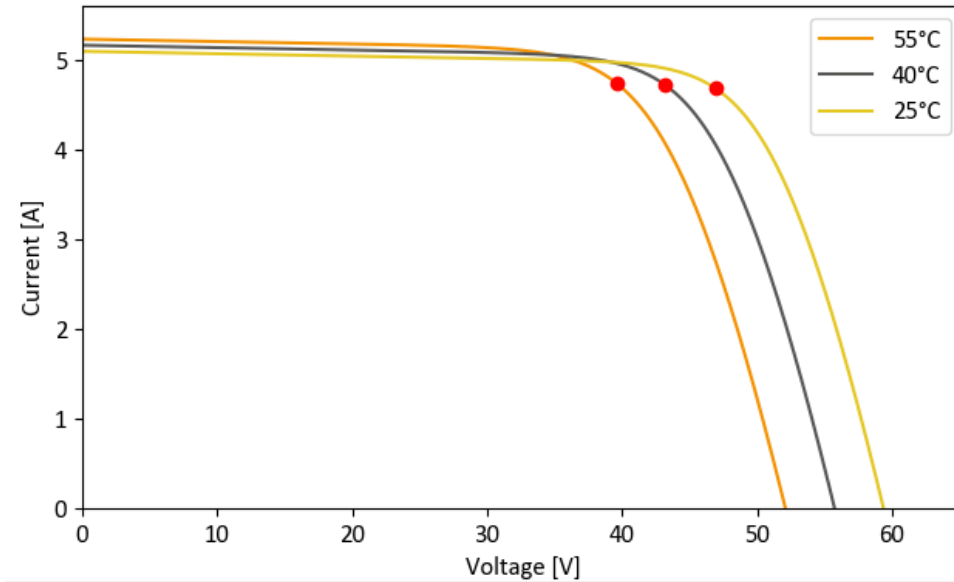
## Modeling steps

### 5. Module and String IV Curve

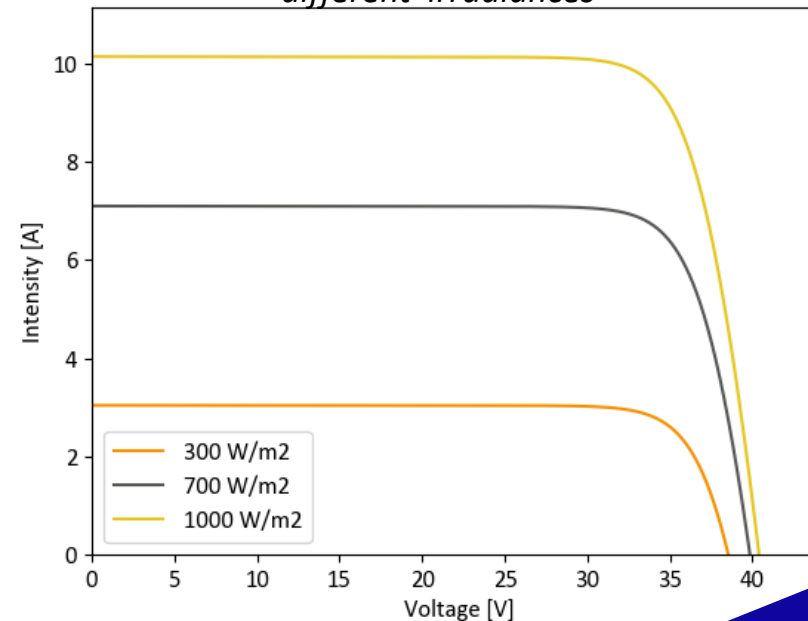
The IV curves' dependencies:

- Higher cell temperatures mostly decrease the voltage
- Higher irradiance level mostly increase the current

*Example of I-V curves as function of different module temperature*



*Example of I-V curves as function of different irradiances*

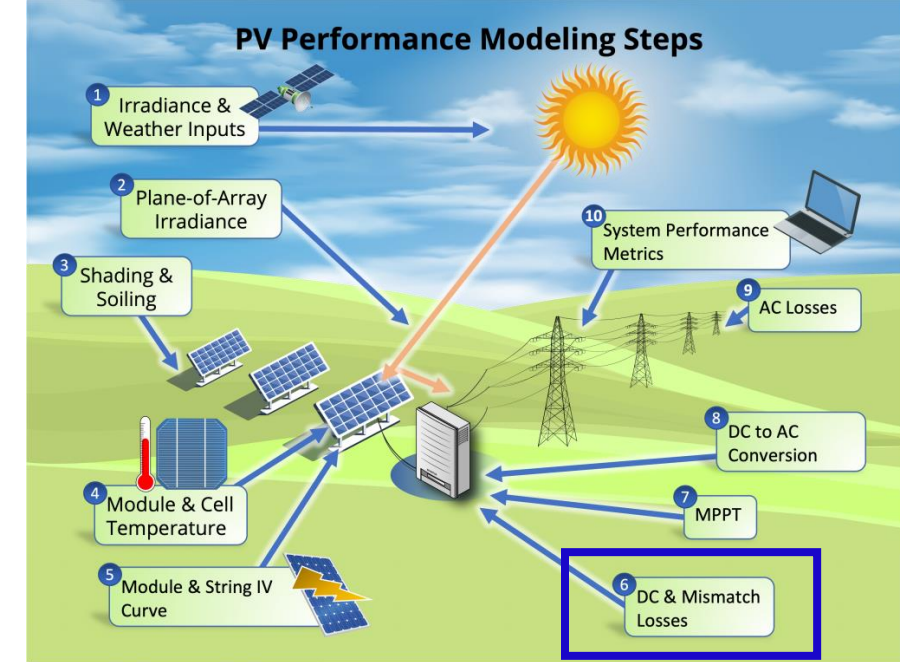


## Modeling steps

### 6. DC & Mismatch Losses

Not the focus of this class. However, keep in mind that:

- **DC wiring losses** are around 0.5%-2%.
- **Mismatch losses** refer to the fact that PV modules have different IV curves and this can entail significant losses.

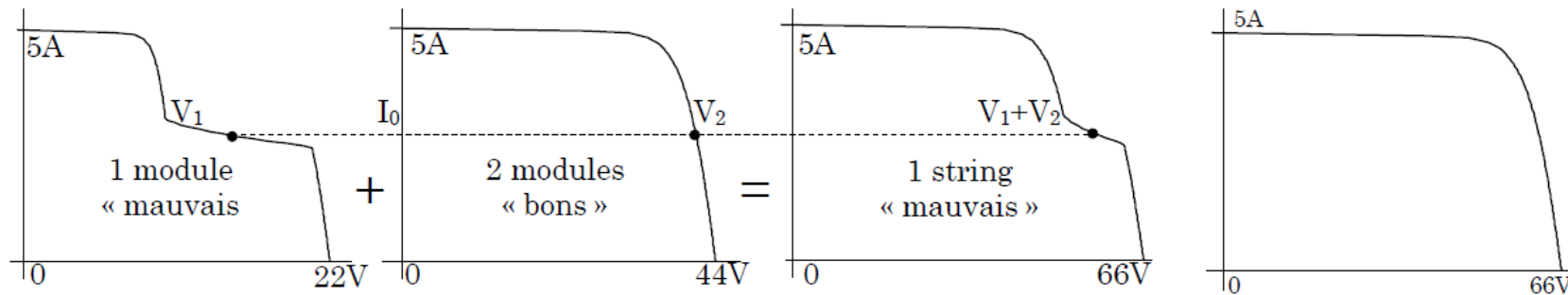
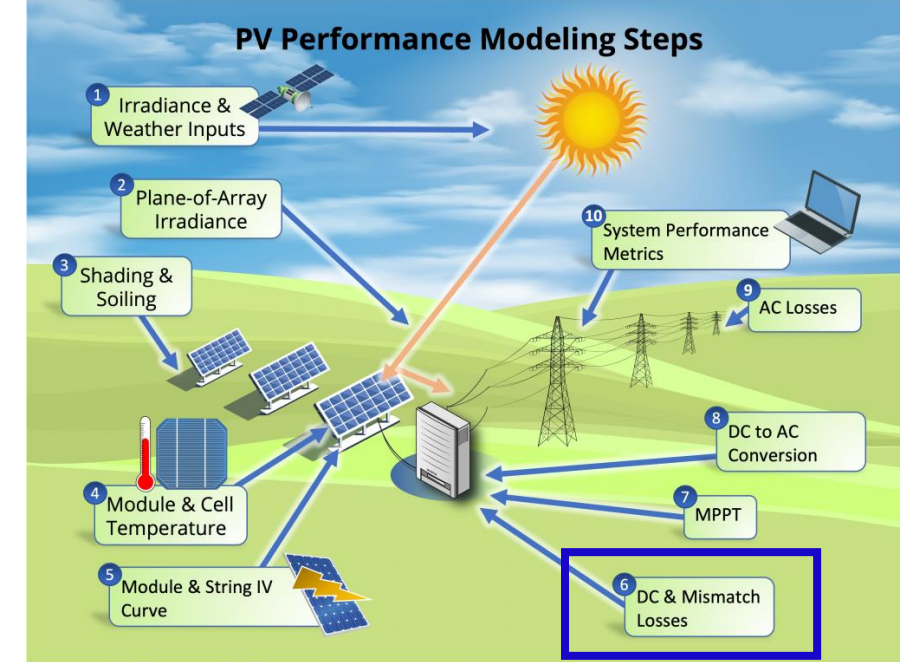


## Modeling steps

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Not the focus of this class. However, keep in mind that:

- **DC wiring losses** are around between 0.5% and 2%.
- **Mismatch losses** refers to the fact that PV modules have different IV curves and this can entail significant losses.  
For instance, if one of them has a very degraded IV curve (shading or other), it can significantly degrade the IV curve at the array level.



(a) string « mauvais »

(b) string « bon »



# Modeling steps

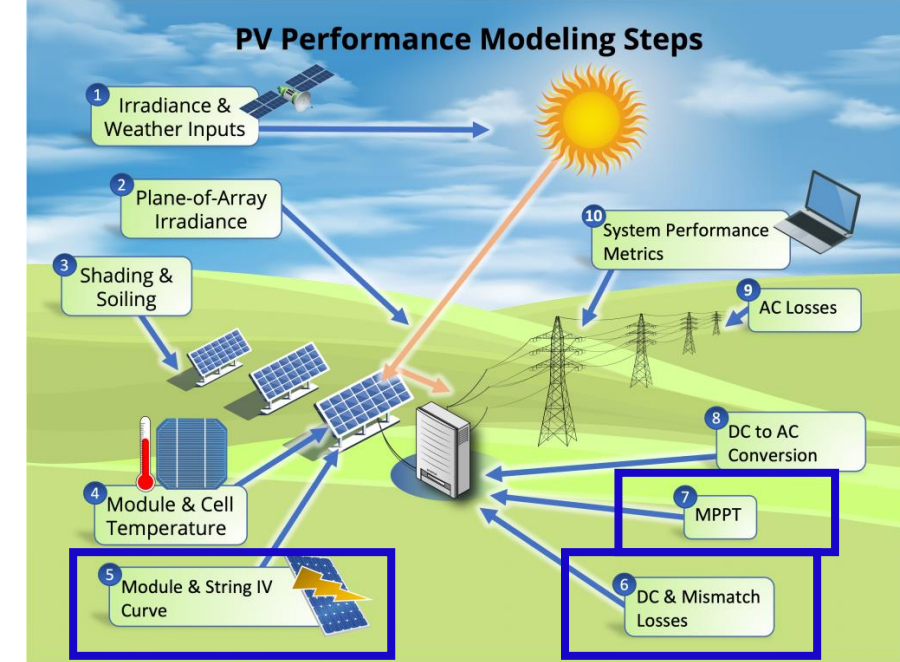
## 5./6./7. Power model

Constant efficiency model:

$$P_{dc} = \eta \cdot G_{POA} \cdot A$$

With:

- $P_{dc}$ , DC power in [W]
- $\eta$  efficiency around 20% (from datasheet)
- $G_{POA}$  the irradiance in the plane of array [ $\text{W}/\text{m}^2$ ]
- $A$ , the PV installation area [ $\text{m}^2$ ]





# Modeling steps

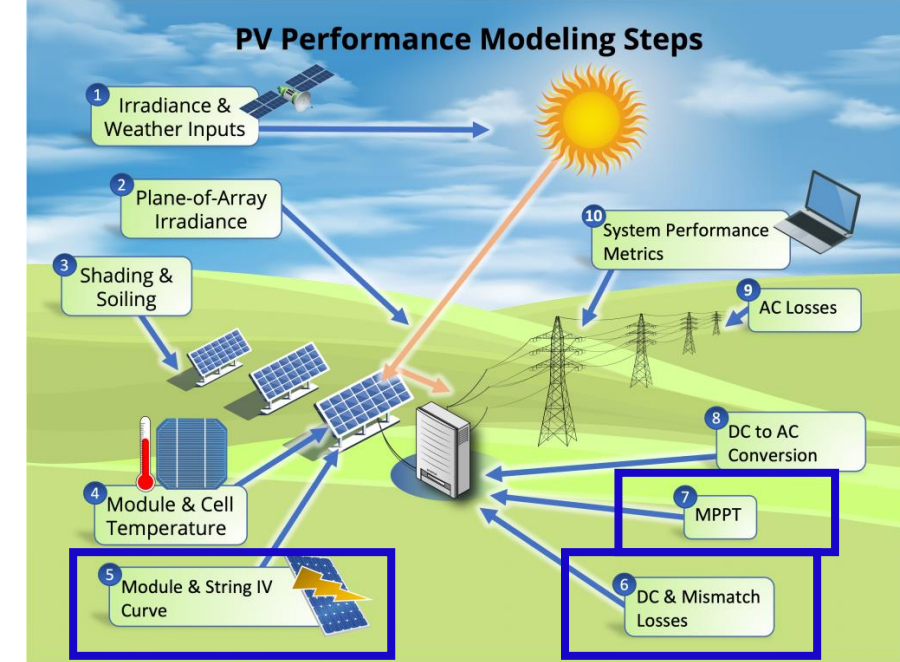
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Not really precise for  
instantaneous values

# Modeling steps

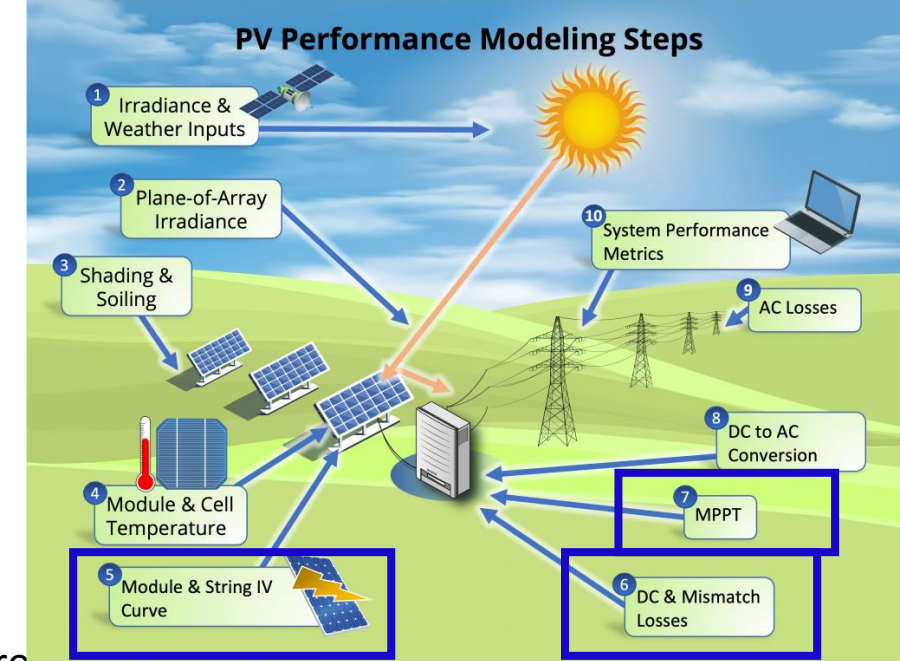
## 5./6./7. Power model

The PVWatts model enables to take into account the effect of the cell temperature

$$P_{dc} = \eta_0 \cdot G_{POA} \cdot A \cdot \left(1 + \gamma_{pdc} \cdot (T_{cell} - 25^{\circ}C)\right)$$

With:

- $\eta_0$ , the reference efficiency (around 20%) [%]
- $G_{POA}$  the irradiance in the plane of array [ $W/m^2$ ]
- $A$ , the PV installation area [ $m^2$ ]
- $\gamma_{pdc}$ , the temperature coefficient (negative, usually between -0.2 – -0.5 %  $W/m^2/^{\circ}C$ )
- $T_{cell}$ , the cell temperature [ $^{\circ}C$ ]



## Modeling steps

### 5./6./7. Power model

The Huld model (used in PVGIS) enables to take into account the module temperature and non-linearity with irradiance.

$$P_{dc} = \eta_{Huld}(G, T_m) \cdot G_{POA} \cdot A$$

$$\eta_{Huld}(G) = \eta_0 \cdot (1 + k_1 \cdot \ln(G') + k_2 \cdot \ln(G')^2 + k_3 \cdot T_m' + k_4 \cdot T_m' \cdot \ln(G') + k_5 \cdot T_m' \cdot \ln(G')^2 + k_6 \cdot T_m')$$

With:

- $\eta_0$ , the reference efficiency (around 20%) [%]
- $G_{POA}$  the irradiance in the plane of array [ $\text{W}/\text{m}^2$ ]
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- $G' = \frac{G_{POA}}{1000 \text{ W}/\text{m}^2}$  the normalized irradiance
- $T_m' = T_m - 25^\circ\text{C}$ , the module temperature delta [ $^\circ\text{C}$ ]
- $k_1 \dots k_6$ , the model coefficients

## Modeling steps

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Reference values from PVGIS

Coefficient	c-Si	CIS	CdTe
$k_1$	-0.017237	-0.005554	-0.046689
$k_2$	-0.040465	-0.038724	-0.072844
$k_3$	-0.004702	-0.003723	-0.002262
$k_4$	0.000149	-0.000905	0.000276
$k_5$	0.000170	-0.001256	0.000159
$k_6$	0.000005	0.000001	-0.000006

## Modeling steps

### 5./6./7. Power model

Use the following notebook:

[https://github.com/AlexandreHugoMathieu/pvfault\\_detection\\_solar\\_academy/blob/master/notebooks/dc\\_power\\_estimation.ipynb](https://github.com/AlexandreHugoMathieu/pvfault_detection_solar_academy/blob/master/notebooks/dc_power_estimation.ipynb)

Follow the python tutorial and estimate the DC power for one year.

Time for some  
hands-on exercises !





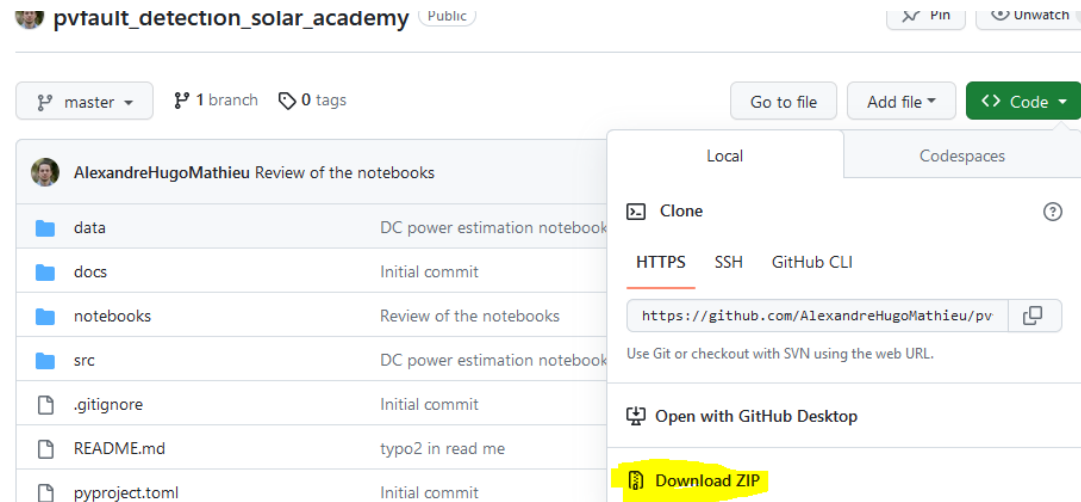
## Resources

- 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](https://iea-pvps.org/wp-content/uploads/2021/10/Final-Report-IEA-PVPS-T13-19_2021_PV-Failure-Monitoring.pdf)

## Appendix

How to install Python and import the course repository to use the notebooks on your local PC.

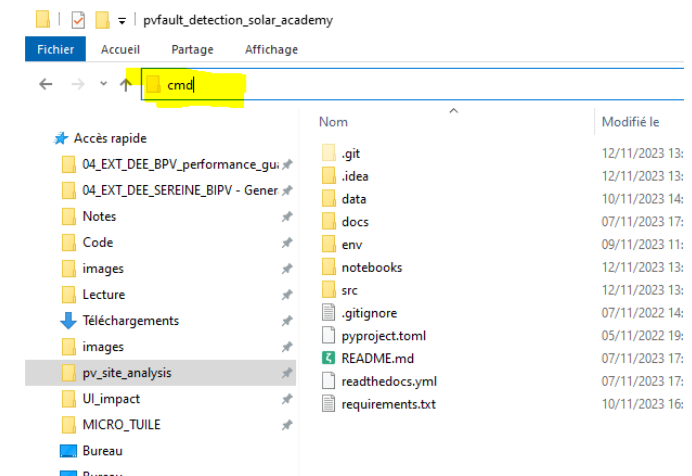
1. Install python: [www.python.org/downloads/](https://www.python.org/downloads/), download and install the 3.9.13 “release” (Add python to your Path)
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3. Unzip it and put it in adequate location in your PC.
4. Let’s create a virtual environment where you will find all the functions for this course:
  1. Go in the folder and open the command line from that same folder by writing “cmd” in the path bar (with Windows)



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3. Unzip it and put it in adequate location in your PC.
4. Let’s create a virtual environment where you will find all the functions for this course:
  1. Go in the folder and open the command line from that same folder by writing “cmd” in the path bar (with Windows)
  2. In the command bar: execute the following line to create the “solar\_env” environnement that you will use in your notebooks
    1. “pip install virtualenv”
    2. “python -m virtualenv solar\_env”
    3. “call solar\_env\Scripts\activate” (you should have a ‘solar\_env’ on the left of the command at this point)
    4. “pip install -r requirements.txt” (load all the libraries, take a little time, be patient)
    5. “python -m ipykernel install --name=solarkernel” (create a kernel for the notebooks)

# Appendix

## How to start a notebook

1. Go in the folder and open the command line from that same folder by writing “cmd” in the path bar (with Windows)
2. In the command bar, execute:
  1. “call solar\_env\Scripts\activate” (go in the virtual env)
  2. “jupyter notebook” (open the notebooks browser)
3. Browse to the notebooks folder, choose one and pick the solarkernel when asked.

