



Some thoughts on the evolution of the solar photovoltaic technology based on previous research

Matthieu Glachant, Cerna, MINES ParisTech

AMPERE-ADVANCE-PIAMDDI workshop

Sevilla, 27-28 May 2013

Sources

- What cost for photovoltaic modules in 2020? Lessons from experience curve models, De La Tour, Glachant, Ménière, 2013, *i3 working paper*
- How do solar photovoltaic feed-in tariffs interact with solar panel and silicon prices? An empirical study, *De La Tour, Glachant, 2013, i3 working paper*



Outline

To use the papers to illustrate two points:

- The learning curve paper
 - To **explain** and to **forecast** may be two contradictory research goals
- The FIT paper
 - **Observationally-small differences** in policy behavior may have strong impacts on the dynamics of the PV technology price

The learning curve paper

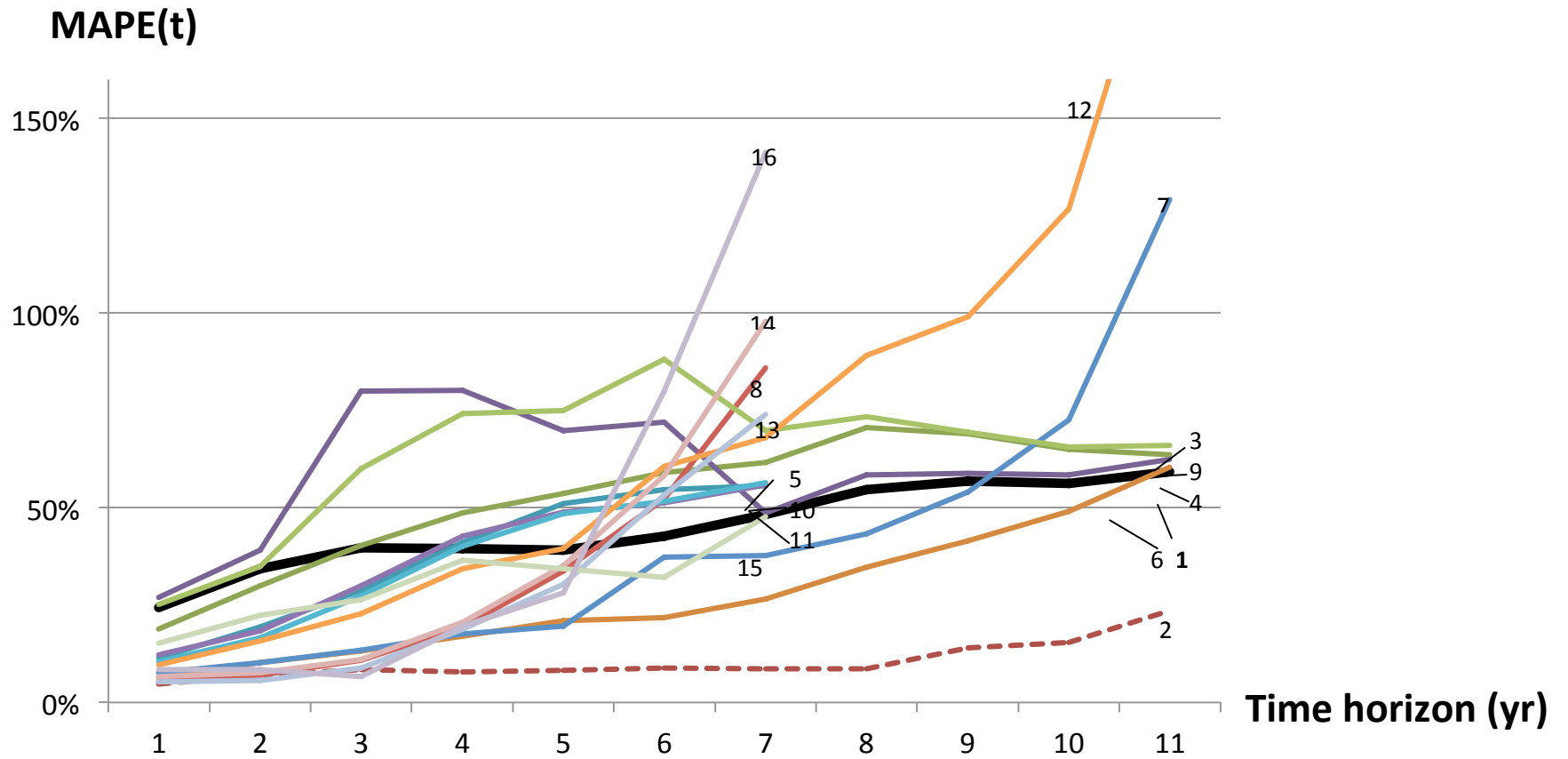
- The question
 - To predict the cost of PV modules out to 2020
- The method
 - We estimate multifactor learning curves with annual data on photovoltaic module prices over the period 1990 – 2011
 - We test different sets of explanatory variables including cumulative production, scale, R&D knowledge stock and input prices for silicon and silver
 - We select the model which minimizes the difference between predicted and actual module prices.

16 sets of independent variables

Cumulative capacity and...

- 1) No additional variable
- 2) Si (Silicon)
- 3) Ar (Silver)
- 4) Scale
- 5) R&D
- 6) Si and Ar
- 7) Si and Scale
- 8) Si and R&D
- 9) Ar and Scale
- 10) Ar and R&D
- 11) Scale and R&D
- 12) Si, Ar, and Scale
- 13) Si, Ar, and R&D
- 14) Si, Scale, and R&D
- 15) Ar, Scale, and RD
- 16) All (Si, Ar, Scale, and R&D)

Comparison of MAPE for each model,



MAPE(t) is the mean absolute percentage error according to the time horizon t:

$$MAPE(i,t) = \frac{1}{n} \sum_{i=1}^n \left| \frac{P_{i,t} - \hat{P}_{i,t}}{P_{i,t}} \right|$$

Results

- The best model only includes cumulative capacity and silicon price as explanatory variable
- It predicts a 67% decrease of module price from 2011 to 2020.
- Hence increasing the number of explanatory variables does not necessarily improves the predictive power
 - A compromise between omitted-variable bias and multicollinearity
- But necessary if the goal is to study precise policy scenarios
 - Increasing public R&D support
- The fundamental question when choosing the learning curve specification: **What policy questions will be addressed with the results?**

How do FITs and module prices interact with each other?

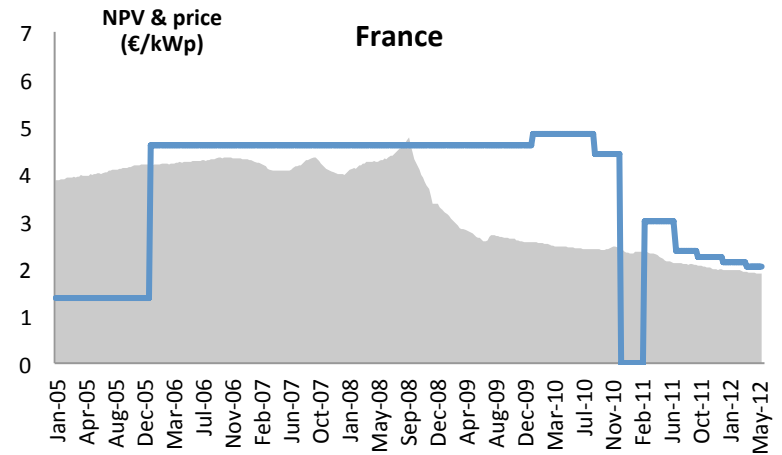
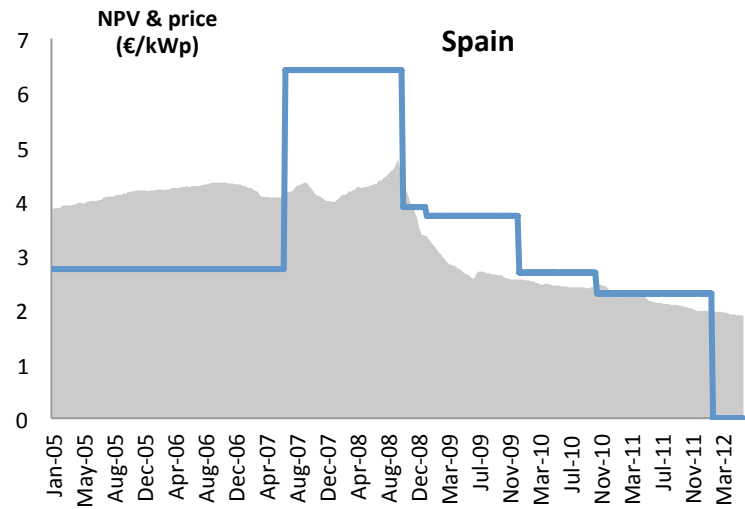
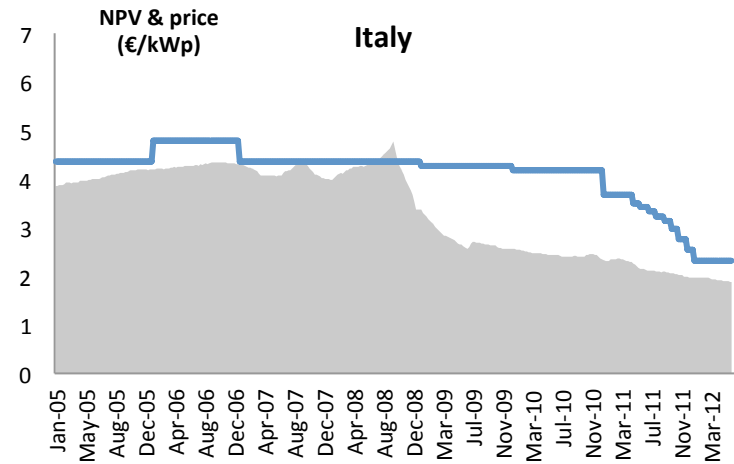
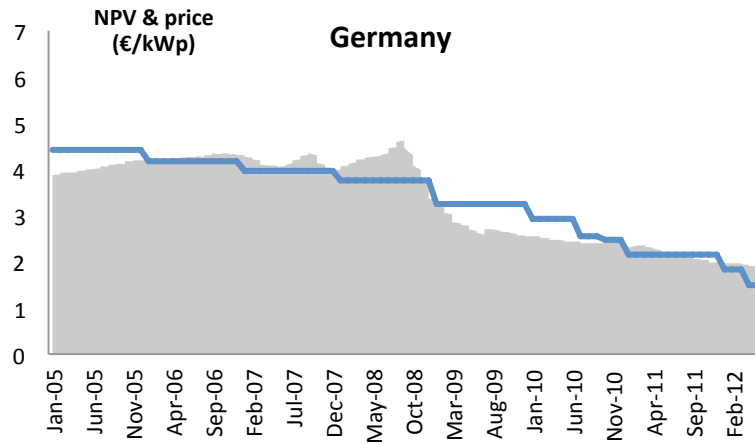
- A FIT subsidizes the demand for PV systems
- In particular, it increases their prices
- It allows PV system producers to collect rents
 - Including foreign manufacturers
- The regulator needs to adjust the FIT in order to minimize these rents while maintaining the incentives to install PV systems

The question: Did the regulators succeed in this exercise? Were they able to limit rents?

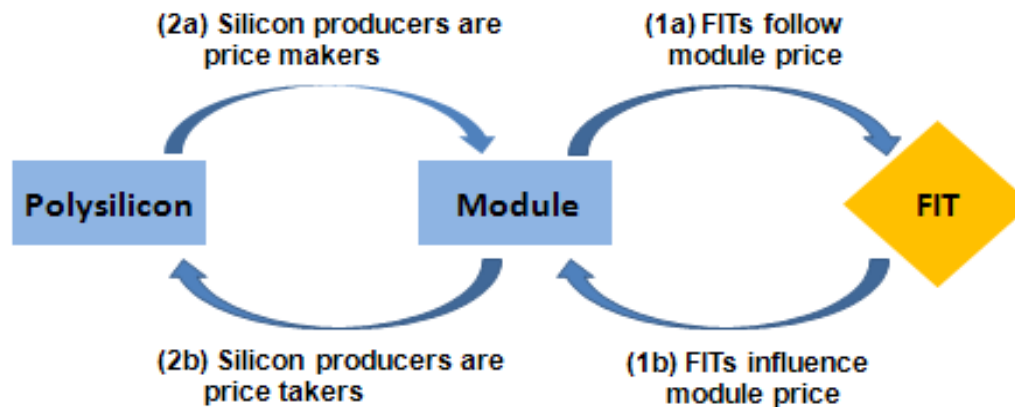
The approach

- The interactions between feed-in tariffs, silicon prices and module prices
- Weekly price data and FIT values in Germany, Italy, Spain, and France from January 2005 to May 2012
- Methodology based on Granger causality tests applied to vector autoregressive models (VAR)

Comparison of PV systems price (shaded area) with the value of the FIT (line)



Tested assumptions



- If **Assumption 1a** is true, it means that the regulator is able to adjust the FIT to changes in module price
- If **Assumption 1b** is true, it means that the opposite

Granger causality test results for the period after the silicon shortage

Dependent variable	Excluded	chi2	df	Prob > chi2
$\ln(\text{module}_{t+10})$	$\ln(\text{silicon}_t)$	3.090	3	0.378
	$\ln(\text{FIT}_t)$	2.722	3	0.436
	ALL	7.006	6	0.320
$\ln(\text{silicon}_t)$	$\ln(\text{module}_{t+10})$	17.47	3	0.001
	$\ln(\text{FIT}_t)$	0.567	3	0.904
	ALL	18.69	6	0.005
$\ln(\text{FIT}_t)$	$\ln(\text{module}_{t+10})$	1.518	3	0.678
	$\ln(\text{silicon}_t)$			

Results

- Since the end of the period of silicon shortage in 2009, module price variations cause changes in FITs, and not the reverse.
- This is good news as it suggests that the regulators have been able to prevent FITs to inflate module prices.
- General methodological point:
Observationally-small differences in policy behavior – whether regulators are able to timely adjust their FIT – may generate very different technological trajectories