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Introduction to the Special Issue of *The Energy Journal*

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Introduction to the Special Issue of The Energy Journal

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Abstract

After nearly two decades of debate and fundamental disagreement, top-down and bottom-up energy-economy modelers, sometimes referred to as modeling ‘tribes’, began to engage in productive dialogue in the mid-1990s (IPCC 2001). From this methodological conversation have emerged modeling approaches that offer a hybrid of the two perspectives. Yet, while individual publications over the past decade have described efforts at hybrid modeling, there has not as yet been a systematic assessment of their prospects and challenges. To this end, several research teams that explore hybrid modeling held a workshop in Paris on April 20–21, 2005 to share and compare the strategies and techniques that each has applied to the development of hybrid modeling. This special issue provides the results of the workshop and of follow-up efforts between different researchers to exchange ideas.

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I. The original bottom-up / top-down division

Policy-makers are interested in a better understanding of the effectiveness and cost of policies whose purpose is to shift energy systems toward more environmentally desirable technology paths. What technologies would serve this purpose, and how could or would the economy adapt in response to policy to achieve this end? Two contrasting modeling types have developed to answer these questions.

Conventional bottom-up (BU) models have described the current and prospective competition of energy technologies in detail, both on the supply-side (the substitution possibilities between primary forms of energy) and on the demand-side (the potential for end-use energy efficiency and fuel substitution). These models were helpful in illustrating the possibility for radically different technology futures with significantly different environmental impacts. However, they have been criticized for not providing a realistic portrayal of either micro-economic decision-making by firms and consumers when selecting technologies, or the macro-economic feedbacks of different energy pathways and policies in terms of changes in economic structure, productivity and trade that would affect the rate, direction and distribution of economic growth.

Conventional top-down (TD) models, in contrast, have addressed the consequences of policies in terms of public finances, economic competitiveness and employment. Since the late 1980's TD energy-economy policy modeling has been dominated by computable general equilibrium (CGE) models, reflecting the decline in the influence of other macroeconomic paradigms, such as disequilibrium models. CGE models were assumed to represent real-world micro-economic responsiveness to policies, such as the substitutability of energy for other inputs or consumption goods. What CGE models tend to lack, however, as do TD models in general, is technological flexibility beyond current practice. If the input substitution elasticities critical to technological response in TD models are estimated from historical data, there is no guarantee that the values for these parameters would remain valid in a future with ambitious policies for environmental improvement, *i.e.* shaped by induced technical change. For example, until recently, there was no incentive to innovate and commercialize technologies with low greenhouse gas emissions. Today, such

technologies are under development worldwide, providing households and firms with new choices that may change elasticities at the level of inter-factor substitution and the level of structural change, with significant implications for total economic output. Furthermore, while energy production and use can be treated as a marginal component of the overall economy in the short run, in the long run large changes in energy supply and use can have significant macroeconomic growth and structure implications. At the extreme, for very long term scenarios and in case of large departures from baseline projections, TD models cannot guarantee that their economic projections are underpinned by a feasible technical system.

Another limitation of the conventional TD approach is that the constraints of policy design processes often push policy-makers towards technology- and building-specific policies in the form of technology or emission standards, regulations, information programs as well as tax credits or subsidies. Conventional TD models represent technological change as an abstract, aggregate phenomenon—implicit in their substitution elasticities— an approach well suited to helping policy-makers assess economy-wide price instruments such as taxes and tradable permits, but one that has difficulties in assessing the combined effect of these price-based policies with technology-specific policies.

The TD/BU debate first came to prominence during the efficiency-gap debate of the 1980s and '90s (Grubb et al., 1993). On the one hand, TD modelers (notably CGE modelers) generally work with model forms that assume that competitive markets automatically allocate all inputs and final goods efficiently. This economic perspective a priori denies the existence of an energy efficiency gap – that there could be a quantity of energy efficiency that society could profitably achieve. On the other hand, bottom-up models suggested that there were significant “no-regrets” possibilities for increasing energy efficiency in the economy; this divergence of views is still not completely resolved, with significant import for energy policy. In the opposite direction, the conventional TD models also underestimate the transition costs of policies due to inertia in the adaptation of markets and to imperfect foresight.

The gap between the two representations of technology became very noticeable when the policy debate refocused on shifting the economy to a technology path with dramatically lower greenhouse gas (GHG) emissions. Policy-makers need to make decisions today about the magnitude and timing of energy-

environment targets, and about the specific policy package that would best achieve them in terms of the usual policy-making criteria — economic efficiency, environmental effectiveness, and administrative and political feasibility. To do so, they need to know the extent to which their policies might influence the characteristics and financial costs of future low or zero GHG emissions technologies, the likely willingness of consumers and businesses to adopt these, and the impact of policies on employment, competitiveness, and economic structure: neither modeling perspective is able to give completely defensible advice for these requirements.

To be particularly useful, an energy-environment policy model should perform fairly well in terms of all three dimensions of Figure 1. It should be technologically explicit, including an assessment of how policies to promote technology commercialization and diffusion might affect the future financial costs of acquiring new technologies. It should be behaviorally realistic, including an assessment of how policies to increase market share might affect the future intangible costs (specific consumer concerns and preferences) of acquiring new technologies. It should have macroeconomic feedbacks linking energy supply and demand to the evolution of the economy's structure and total output. This macroeconomic dimension should include trade and financial feedbacks between countries in cases where the environmental challenge is one that requires a global effort, such as with greenhouse gas abatement.

The characteristics of conventional TD and BU models are compared with respect to each other using the dimensions of Figure 1. Conventional BU models do well in terms of technological explicitness, but less well in terms of the other two attributes. Conventional TD models do well in terms of macroeconomic completeness and general micro-economic realism, but they fail to represent the potential for no-regret options over the short run and substantially different technological futures over the long run.

The three dimensions of the figure are useful not only for contrasting the two conventional modeling approaches, but for providing a framework for discussion. For example, because they lacked technological explicitness, conventional TD models have tended to suggest that efforts to substitute away from specific forms of energy for political or environmental objectives would be relatively costly (*i.e.* the economy's potential for technological transformation being somewhat limited as portrayed by historically-based elasticities). These models have therefore often produced high cost estimates for abatement of

energy-related greenhouse gas emissions. With their technological explicitness, but failure to include micro- and macro-economic realities (technology-specific risks and preferences, rebounds in demand resulting from greater efficiency), conventional BU models have tended to suggest that efforts to substitute away from specific forms of energy or to reduce greenhouse gas emissions would be relatively inexpensive and in some cases even profitable.

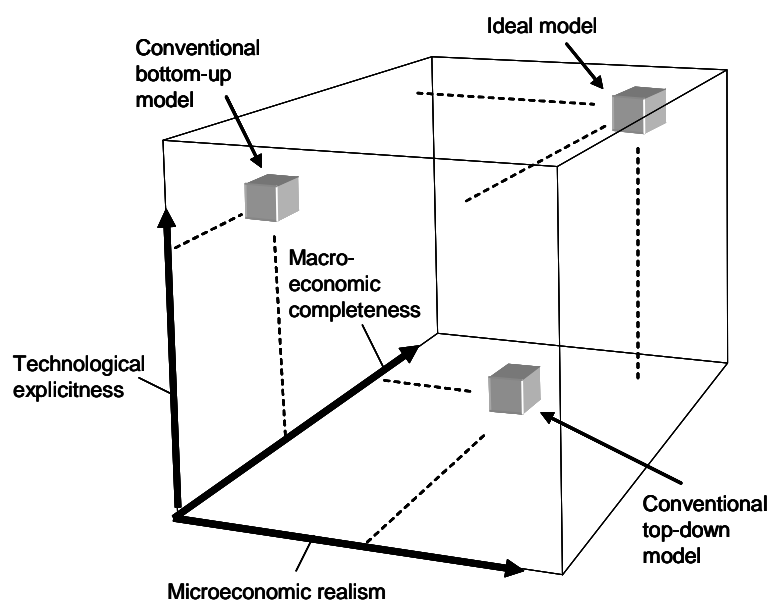


Figure 1. Three-dimensional assessment of energy-economy models

Figure 1 implies that there is the possibility of a better model, one that scores high on all three requirements, indicated by the “ideal” model in the back, right, and top corner of the cube. An increasing number of modeling teams have recognized the possibility of realizing something closer to this ideal model; we now turn to these approaches.

II. Toward hybrid models

Thus far, we have been careful to refer to “conventional” TD and BU models. While these two conventional modeling approaches are still used a great deal,

a number of researchers are developing “hybrid” models that seek to compensate for the limitations of one approach or the other. Some BU modelers have incorporated macro-economic feedbacks in their models, while others have estimated micro-economic behavioral parameters for technology choices in their models (see Bataille *et al.* in this issue for a survey). On their side, some TD modelers have incorporated technological explicitness for the energy supply or transportation sectors in their models. A few of them are now incorporating parameters for endogenous technological change, meaning that energy productivity or green-house gas intensity is somehow linked to policies that foster research and development or market penetration of low-emission technologies (Löscherl, 2002).

In terms of Figure 1, these developments imply that some BU models are shifting toward the right and back corner of the cube while some TD models are climbing vertically on the technology explicitness dimension of the cube. It can be misleading to refer to models that have undergone some of these developments using terms like BU or TD—hence our use of the term *hybrid* models. We define hybrid models, therefore, as those BU or TD energy-environment models that have made at least one modification that shifts them substantially away from their conventional placement in the cube of Figure 1. Some hybrids originated as BU models, some as TD models, but all hybrids have characteristics that differentiate them significantly from conventional TD and BU models.

The development of these models faces several challenges related to theoretical consistency, computational complexity, empirical validity and policy relevance. Each of the papers in this issue addresses these challenges to some degree. Thus, each paper explains the theoretical basis for its model design, the structure of the model and its key algorithms, the data requirements, empirical estimation of critical parameters, and illustrates all of this with a policy-relevant application.

III. Summary of the papers

Kim *et al.* argue for the necessity of a flexible and expandable model, developed in a modular form and capable of easily incorporating detailed, forward

looking technological information into a macroeconomic framework. To facilitate this they use an Object-oriented Energy Climate Technology Systems (OBJECTS) Framework, which they demonstrate by modeling transportation in the MiniCAM long-term, global integrated assessment model. This integration is made internally consistent through an ‘object hierarchy’ of the global economy, which links and unifies the top-down and bottom-up representations of technology and identifies the inter-relationship of each class of the hierarchy. This approach requires strong accounting discipline to ensure ‘objects’ do not double-count components of the economy. The authors’ priority is to facilitate interdisciplinary dialogue, and to incorporate sector information of differing types and quality, however it may be found. This is why MiniCAM operates as a recursive partial equilibrium model, where: only markets for energy and agricultural goods are equilibrated; the corresponding prices, including wages and capital costs, are exogenous; and a simple feedback loop of energy prices on GDP allows the assessment of macroeconomic costs.

At the other end of the spectrum of modeling approaches gathered in this issue, the WITCH model of Bosetti *et al.* incorporates an engineering-based energy supply specification into a Ramsey-type optimal growth model. This choice was made to further the ultimate objective of analyzing international carbon policies as optimal strategies in a game-theoretic framework: the interactions between world regions are modeled as a non-cooperative Nash Game in which a social planner in each region takes as given the behavior of other countries. WITCH consists of a very compact growth engine depicting the dynamics of a single final good, but this high level of sectoral aggregation is combined with a more detailed representation of the energy supply sector. Treatment of energy demand remains very aggregate. The challenges met by WITCH are: to have a systematic endogenous technical change framework, including R&D and learning by doing, both for the energy and the composite goods sectors; and to represent investment decisions in the energy sector through a dynamic open loop with perfect foresight. This mix of intertemporal optimization, a limited disaggregation of energy production technologies and of integration of many interdependency channels, including a game-theoretic approach to international policy design, makes WITCH an original endeavor.

Köhler *et al.* is the only article in this issue that describes a modeling system (E3MG) based on ‘post Keynesian’ dynamic macro-econometrics instead of a general equilibrium framework. Until now the use of Keynesian models has

been restricted to short and medium term time horizons, for which they are the favored tool. For longer term analysis, general equilibrium models have been the tool of choice. E3MG, however, challenges this worldview and uses its Keynesian foundation to address questions for which general equilibrium models are ill-suited, such as the effects of policy in a “second-best” environment. It explores the possibility of representing growth pathways under disequilibrium at a high level of disaggregation: econometrically-calibrated dynamic equations for 41 regional sectors link investment to historical demand and investment trends. It thus provides a direct contrast to the WITCH approach of using a high level of aggregation with a forward-looking decision framework. This representation of economic growth allows for a description of the evolution of the economic structure over the long run; it does not use production functions, and the evolution of the energy sector is described by a simplified description of the technological dynamics of the ETM model, including use of a learning curve. ETM receives information from E3MG in each period, and sends back the induced amount of investment and input and final output production costs.

The purpose of the AMIGA modeling system (Laitner and Hanson) is a better representation of the real decision parameters in technology choices, including market imperfections, risk aversion and pervasive principal-agent problems. It does so within a general equilibrium framework at a high level of disaggregation (200 sectors for the United States, 30 for the 20 other world regions). This sectoral disaggregation allows AMIGA to distinguish the behavioral responses to policy of different market segments, and just as importantly, between types of capital stocks, in terms of vintages and productive uses. Consumers maximize intertemporal utility but without perfect foresight. Imbedded in a nested structure of factor substitution, the substitution of capital to energy consumption is treated as movement along the isoquant of a conventional production function informed by BU analysis. This isoquant can be modified by technical change in the future, as in many other models, but a unique characteristic is a focus on the end-use level, based on explicit technologies from a vast data set, and use of a hurdle rate that represents the preferences of consumers for specific technologies, which can be modified by public policies or promotional efforts.

Böhringer and Löschel’s paper is methodological in nature, with an application to the penetration of renewable energy technologies in Europe. It ad-

addresses the difficulty of representing technology in top-down models using continuously differentiable functions that capture substitution possibilities through constant elasticities. While energy planning can be formulated as an optimization problem with inequality constraints on decision variables, such as capacity restrictions on production quantities, the shadow prices associated with programming constraints may not coincide systematically with market prices. This means that the conditions necessary for use of conventional production functions (the ‘integrability’ condition) are not respected. Böhringer and Löschel demonstrate how reformulating the issue as a Mixed Complementarity Problem allows for technological explicitness in the electricity sector (with discrete technologies), while aggregate technological options in other sectors can still be represented by means of constant elasticities of substitution—amounting to the assumption of a macro-economic growth engine independent from the energy systems, which is justified in the analysis of moderate energy policies as the one exemplified. Using this approach, the shadow prices associated with any non-price constraint, such as physical restrictions on capacity for a given technology, can be introduced.

The IMACLIM model (Ghersi and Hourcade) addresses the consistency problem arising from using BU information to describe the transformation of energy supply and demand, which may indicate non-constant input or demand substitution, in the context of the standard general equilibrium practice of using constant elasticities of substitution in production and demand functions. This consistency issue is most pronounced under strong policy, such as in the case of very high carbon prices. The intuition is that large departures from reference energy trends (supply and demand) cannot but impact on the economic growth engine: there are inherent interdependencies between the long run macroeconomic possibility frontier and engineering based information. IMACLIM uses an innovation possibility curve to describe the envelope of production possibilities generated by various sets of price signals applied to BU data. The same method is used for demand functions. The value of IMACLIM’s method is illustrated by a general equilibrium analysis of the effect of a large set of carbon prices in 2030 on the world economy. Results are compared to those obtained with conventional CES functions calibrated on the same BU data. The latter functions are shown to significantly bias macroeconomic costs estimates, especially for BU expertise showing important convexities (i.e. a large set of low-cost mitigation options eventually tending to satura-

tion of a given end-use). The sign of the bias is not systematic, although it generally causes an underestimate of costs for moderate-to-high carbon prices.

Schäfer and Jacoby propose yet another approach to hybridizing, based on the joint use of three models. An optimization model (MARKAL) of transportation technologies is loosely coupled to EPPA, a recursive dynamic CGE model. The interface between EPPA's aggregated transportation demands and the wide range of technologies represented in MARKAL is made by resorting to a modal choice model. This loose coupling addresses two of the main limitations of the conventional CGE approach: the substitution between energy and capital in the production of transportation services, together with the substitution between transportation services when available, are either exogenously forced or governed by an elasticity that is made to vary over time according to BU results; and the energy efficiency improvements of households' self-produced and purchased transportation services vary according to BU results.

The CIMS model (Bataille *et al.*) provides an alternative to the predominant TD approach to hybridization. Instead of adding technological explicitness to an existing TD model, CIMS was built by bringing together a collection of BU models that provide complete coverage of energy use in the economy, models that compete vintaged end-use and energy supply technologies against each other using behaviorally realistic choice algorithms with empirically estimated parameters, and link them in an integrated framework using a combination of CGE and macro-econometric approaches. CIMS clears the markets for energy commodities and final goods and services in each time period using CGE methodologies, with demand for traded goods being represented by Armington elasticities, and that for non-traded goods by macroeconomic functions linked to overall activity. Bataille *et al.* also describes how CIMS can be used to calculate better long run estimates of sectoral capital-for-energy and inter-fuel elasticities of substitution, as well as sectoral autonomous energy efficiency indices (AEEI), both highly debated parameters that are central to the functioning of TD models, and CGE models in particular. These parameters are difficult to estimate using standard econometric methods because future long-run technological development is not necessarily related to past development, and is also endogenous to policy influence. To estimate long run estimates of elasticities of substitution, CIMS was "shocked" with a series of different long run input prices; the resulting 2035 input shares were regressed

using a standard econometric production function, whose parameters were used to calculate elasticities that are sensitive to the future evolution of the technology stock. To estimate sectoral estimates of AEEI CIMS was run in two modes from 2000 to 2035: in the first, technology was allowed develop and turn over normally, while in the second firms and consumers were constrained to using the technology mix used in year 2000. The difference in energy consumption between the two modes in 2035 was used to calculate the long run AEEI.

IV. Lessons and prospects

The eight papers of this special issue exemplify the diversity of options and issues in the design and application of hybrid models. Many other approaches are possible and will be explored in the future, but in terms of their fundamental design these attempts will likely fit into one of the following categories.

- A TD model that partly renounces the conventional macroeconomist's toolkit (constant elasticities of substitution (CES), and the autonomous energy efficiency index (AEEI)), and relies on innovative ways to represent not only energy supply but also energy end-use technologies as described by BU analysis, and technology adoption as described by microeconomic studies, especially regarding households.
- A TD model that increases its disaggregation level and resorts to Leontief fixed-input ratios to include a reduced-form BU module of some part of the energy system (e.g. in energy supply or the transport sector).
- A BU model that includes: empirically estimated micro-economic parameters related to technology choice; functions to clear markets for energy, other intermediate inputs, and final goods and services based on changes in the cost of production, using either price elasticities or more advanced CGE techniques that utilize consumer utility and firm profit functions; and functions to balance government budgets, exchange rates, and capital and labor markets.
- A composite hybrid model that includes all of the major theoretical and structural characteristics of the most advanced TD models along with

the major characteristics of the most advanced BU models, with technological detail in all sectors and behavioral parameters that are empirically estimated from micro-economic and macro-economic research. While such a model would present the greatest challenge in terms of theoretical consistency, mathematical complexity and empirical estimation, it nonetheless represents an objective that some modelers might aspire to, and has been colloquially referred to as the “Holy Grail”.¹

Many factors influence which of these options might be pursued by a given group of energy-environment policy modelers. To some extent, the choice of model reflects the training and natural inclination of the modeler. Many, but not all, economists are more familiar with and attracted to models that begin from a TD perspective. Many, but not all, technology experts are more attracted to models that begin from a BU perspective.

But if a hybrid model is to be truly useful to the policy-maker, the model’s design should be governed by the objective it is meant to serve. Variations in that objective can have significant implications for the appropriateness of the different hybrid modeling strategies listed above. Here are some of the ways in which the modeling objective may differ.

- The specific energy-environment problem: a global problem like greenhouse gas abatement differs from a regional problem like acid rain or a local problem like urban air pollution. This affects the choice of spatial scale and resolution of the model.
- The policy-making reference point: one reference could be the setting of global energy-environment targets (like greenhouse gas abatement) and the required negotiation of international allocations to achieve them, as well as the design of international mechanisms for achieving targets in a cost-effective and politically feasible manner. Another reference point could be the setting of national energy-environment targets and the design of country-specific policies to achieve these. A third reference point could involve the setting of sub-national targets and policies.
- The policy timeframe: some greenhouse gas abatement models operate over a century—long-run stabilization of atmospheric greenhouse

¹ Models for simulating greenhouse gas abatement policies are still substantially less complex than most of the natural science models simulating the global climate system.

gases—while some are focused on a period as short as 10 years—the achievement of national Kyoto commitments or the operation of international trading mechanisms in the Kyoto timeframe.

- The policy options: a model for assessing the international implications of economy-wide energy or carbon taxes can be simpler in some respects than a model for assessing the national implications of technology- and sector-specific policies that involve a mix of regulations, taxes, subsidies, tradable permits and information provision. Also, a policy that provides future cost information to consumers and firms (such as a scheduled rise in greenhouse gas taxes) requires a model that can simulate foresight by decision-makers.

Whatever the ultimate objective, any effort at hybrid modeling will confront similar theoretical and methodological issues. Some of them are strictly a matter of modeling approach: accounting for capital stock inertia, aggregating economic sectors, guaranteeing the compatibility of technical change as described in endogenous growth models, and the shifts of energy systems as explicitly depicted by technology-rich models, etc. But real breakthroughs will not be made unless parallel progress is made in two partly intertwined directions.

The first concerns the need to achieve a better understanding and representation of business and consumer behavior in the face of uncertainty and diverse policy and market signals. The second concerns the constraints caused by data gaps. Any theoretical advance is bound to confront difficulties because of a lack of data or data mismatches. There may be gaps and inconsistencies of and between national accounts, input-output tables, energy balances and energy prices (type and level of aggregation, time series, etc.). There may also be difficulties in data collection; collected data may not translate easily into modeling structures. Data may be available, but only as case studies of real behaviors in various incentive contexts, which then need to be translated. Finally, there may be gaps in the record, and comparative studies may be required.

Beyond their fruitful direct application to policy analysis, hybrid models should ultimately be regarded as a major avenue of progress in the modeling discipline. They are communication tools between various fields of knowledge (engineering, macro- and micro-economic analysis, comparative institutional studies), and one of their most important contributions may be to help detect

missing information and dynamics, and to provide a structure for discussion and progressive consolidation of modeling methods.

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