

Analysis of Market Penetration of Renewable Energy Alternatives under Uncertain and Carbon Constrained World

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Abstract

• Future energy prices and supply, availability and costs of emerging technologies including CCS, wind, solar and bioenergy can have significant impact on how fast and cost effectively we could abate carbon emissions. Various carbon policies have been proposed and analyzed, however the adequacy of these policies is highly uncertain due to uncertain pace of the climate change. This study investigates how the uncertainty in future energy supply and prices, and various carbon abatement policies will affect the market penetration of CCS and renewable energy alternatives including bioenergy, wind, and solar. Two-staged decision making methods embedded in U.S. EPA's Markal modeling system will be utilized to find most robust mitigation strategy under carbon constraint world. A range of technology availability and cost scenarios are analyzed to capture the uncertainty in the R&D progress.





U.S. Energy Flow - 1950

All values in 1015 Btu (2.12 X 1015 Btu - 106 bbl/day oil)

Total energy consumption = 33.9×10^{15} Btu.

Total Energy Consumption = 35,765 PJ

Ref: Lawrence Livermore Laboratory UCRL-51487



Why Energy and the Environment?

In 2000:

- Air and Water Quality
 - Contributions to U.S. anthropogenic emissions
 - *NOx* 95%
 - SO₂ 89%
 - CO 95%
 - Hg 87%
- Climate Change
 - Contributes 94% of U.S. anthropogenic CO₂ emissions
- Water Supply and Use
 - 89% of U.S. electricity production uses water for steam or cooling
 - 39% of U.S. water withdrawals (agriculture ~ 41%; domestic ~ 12%)
 - 132 trillion gallons of water per day is required
 - 2 gallons/kWh evaporates from thermoelectric plants
 - 18 gallons/kWh evaporates from hydroelectric plants
 - One gallon of ethanol may require four or more gallons of water

Project Goals

- Developing and examining internally consistent scenarios of energy system evolution and anticipating environmental consequences
- Exploring the energy and environmental implications of current, proposed, and potential policies
 - -Climate policies
- Identifying important system interactions and potential unintended consequences
- Developing robust solutions while including consideration of uncertainties in fuel prices, technologies, and policy
 - -Carbon capture and sequestration



What is MARKAL?

- Developed in cooperative multinational International Energy Agency (IEA) project in late 70s
- Selects the optimal mix of technologies and fuels at each time step to minimize the net present value of energy system capital and O&M costs
- Subject to:
 - -Current and projected technology costs and efficiencies
 - -Resource supply costs and competition for fuel across sectors
 - -Resource supply constraints
 - -Trade costs and constraints
 - -Emission limits
 - -Other constraints (e.g., policies)
- Currently used by ~200 institutions and governments in 70 countries
 - Including Department of Energy and U.S. Environmental Protection Agency



Modeling Energy System Interactions



Agency

Forestry biomass

Modeling Technology Change with MARKAL

MARKAL Inputs:

- Future-year energy service demands
- Primary energy resource supplies
- Current & future technology
 characteristics
- Emissions and energy **policies**



 Through linear optimization MARKAL finds the least cost set of technologies

MARKAL Outputs:

- **Technology penetrations** for meeting industrial, residential, commercial, and transportation demands
- Fuel use by type and region
- Sectoral and system-wide emissions NOx, SO₂, PM₁₀ and CO₂
- Marginal fuel and emissions reduction prices



Soon will add Hg, CH₄, and N₂O as outputs

Factors Driving Energy System Evolution

- **Population** growth and migration
- Economic growth and transformation
- Land use change
- Technology innovation
- Climate change impacts on energy use and production
- Availability and cost of fuel resources
- Consumer and firm behavior



Many uncertainties when projecting to future



Scenario Analysis

- Examine hedging scenarios under an uncertain GHG policy considering different assumptions about the availability of advanced technologies such as CCS
 - -Follow a historical CO2 emission path until 2010
 - In 2010, Lieberman-Warner-Boxer CO2 abatement policy is implemented.
 - In 2020, more information could be attained regarding climate change which could change the CO2 emission limits



Results generated with U.S. EPA National Model database

Methodology

- Utilized two-stage stochastic optimization feature embedded in MARKAL framework
- Constructed the scenarios as multiple realizations of the future
- Run the stochastic model
- Run the deterministic (perfect foresight) models (9 runs)
- Compare the results from stochastic vs deterministic models

1st stage-year 2020 2nd stage-year 2025



Hypothesis

- Expect to see short-term differences between 2000 and 2020 to help prepare for the risk of change in 2020 for stochastic version
- Expect technologies to be chosen that are a reasonable price but have low CO2 output
 - -Coal is cheap, but is a large CO2 emitter
 - -Coal depends on CCS to be viable for CO2 reduction
 - Natural gas allows for reasonable CO2 emission levels without CCS → Could serve as a transition fuel
 - -Natural gas w/CCS is a good solution for extreme CO2 reduction
- The deterministic run knows what to expect in the future, but the stochastic version has to be more robust to minimize cost while having the ability to reduce CO2 emissions

-the cost of having perfect information



Technology Assumptions

	Availability	Capital	Fixed	Variable	Heat Rate
Technology	Year	\$M/GW	\$M/GW	\$M/kWh	
Natural Gas - Advanced Combined-Cycle (Turbine)	2005	486	9.73	1.66	6333
Natural Gas - Advanced Combustion Turbine	2005	335	8.76	2.64	8550
Geothermal - Binary Cycle and Flashed Steam	2005	1919	68.52		35460
Biomass Integrated Gasification Combined-Cycle	2010	1516	44.37	13.74	8911
Pulverized Coal Steam - 2005	2005	1066	22.90	3.82	8600
Supercritical Coal Steam - 2010	2010	1100	22.90	3.82	7200
Integrated Coal Gasif. Combined Cycle	2015	1232	32.17	2.42	7200
Integrated Coal Gasif. Combined Cycle CO2 Capt.	2015	1873	41.44	4.03	8538
Oxyfuel Coal Steam CO2 Capture	2015	1873	41.44	4.03	8538
Natural Gas Combined Cycle CO2 Capture	2015	1021	18.12	2.66	7957
Nuclear LWRs in 2010	2010	1440		5.00	0.65*
Pebble-Bed Modular Reactor (PBMR)	2020	1250		2.50	0.36*
Gas Turbine - Modular Helium Reactor (GT-MHR)	2020	1122	26.48	0.64	0.22*
Solar PV Centralized Generation	2010	3436 - 3931			
Solar PV Distributed Residential Generation	2010	4512 - 6771			
Solar PV Distributed Commercial Generation	2010	3760 - 4870			
Solar Thermal Centralized Generation	2015	2325 - 2465			
Wind by Generation and Cost Category	2005	1381-4214			

* tons of U45/ PJ ELC

Most cost data is adopted from AEO 2006.



Illustrative Results

Stochastic vs. Deterministic?

Stochastic

Perfect Foresight



Results

- CCS Activity levels are nearly the same
- All available CCS is used in strict CO2 emission scenarios
 - An important technology in the face of CO2 abatement



Illustrative Results Total Costs and Value of Perfect Information

Variations of Scenarios	Nomenclature		D TOTCOST	D TOT FMIS TAX	Total Cost (M\$)
	PF is Perfect Foresight	PF1-1	60.026.452	38.236	60.064.688
2-CO2 relaxed	ST is Stochastic	PF1-2	53,098,505	60,680	53,159,185
3-CO2 same as 2020 levels	PF1-3	52,449,626	91,020	52,540,646	
1-CCS low CCS low			51,600,767	39,881	51,640,648
2-CCS medium	PF2-2	51,510,752	59,154	51,569,906	
3-CCS high			51,461,141	81,144	51,542,285
-		PF3-1	51,245,155	18,496	51,263,651
	PF3-2	51,245,155	18,496	51,263,651	
	PF3-3	51,245,155	18,496	51,263,651	
		ST1-1	60,394,425	39,881	60,434,306
Comparing costs	ST1-2	53,096,677	60,276	53,156,953	
 D.TOTCOST + D.TOT.EMIS.TAX = Total Cost 			52,514,519	91,020	52,605,539
 Value of perfect information = Stochastic – 			51,732,280	39,868	51,772,148
Deterministic			51,664,567	58,332	51,722,899
		ST2-3	51,620,980	80,785	51,701,765
Stachastia Datarmini	tion Cost of Porfact Information	ST3-1	51,415,262	15,214	51,430,476
SW/1.2 53 156 053 53 150		ST3-2	51,415,262	15,214	51,430,476
SW 1-2 53,150,955 539 52,540	646 64 803	ST3-3	51,415,262	15,214	51,430,476
SW 2-1 51 772 1/8 51 640	648 131 500				
SW 2-2 51,722,899 51 569	.906 152.993			.€F	ΡΔ



Illustrative Results ST1-2 (selected)

CO2 LWB with medium CCS

ST1-3 (baseline) CO2 LWB with high CCS



Less CCS (positive), natural gas is necessary to meet CO2 constraints.
With more CCS available (negative), Coal w/ CCS is cheapest, but as CO2 limits become tighter, Natural Gas w/CSS is necessary.





Illustrative Results ST1-2

CO2 LWB -CCS medium

PF1-2 CO2 LWB -CCS medium



 Wind and natural gas utilized more in ST1-2 than PF1-2



Electricity Production by Fuel & Type

Illustrative Results ST1-3 CO2 LWB-CCS high

PF1-3 CO2 LWB-CCS high



 Trade-off between coal and natural gas CCS



Illustrative Results

ST2-1 CO2 relaxed - CCS low





(2,000)



 Low CCS results in increased use of natural gas in PF2-1



Year

Electricity Production by Fuel & Type

Analysis of results

- Scales of differences were much smaller than expected between stochastic and perfect foresight
- Observed a trade-off between natural gas and coal
- New nuclear did not play a critical role



What's next?

- Updating the results with the latest AEO projections and recent technology cost projections
- Extend the analysis to include additional drivers
 - -Fossil fuel prices, especially natural gas prices
 - -Availability and costs of coal and natural gas technologies
 - -Availability and costs of renewable energy technologies
- Tinker with stochastic parameters to get different results
 - -Vary probabilities of branches
 - -Apply a risk aversion level to a parameter
 - Weights the cost of specified parameter, giving the modeler control of the uncertainty
 - λ = risk level, λ = 0 is Perfect foresight, λ > 0 is risk averse



THANK YOU

QUESTIONS?



Team and Contact Information

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