

THE IMPORTANCE OF TECHNOLOGICAL INNOVATION IN POLICY ASSESSMENTS OF CARBON CAPTURE AND STORAGE

Edward S. Rubin,^{*} Keywan Riahi,[†] Sonia Yeh,^{*}
David Hounshell^{*} and Leo Schrattenholzer[†]

^{*} Carnegie Mellon University

[†] International Institute for Applied Systems Analysis (IIASA)

U.S. Department of Energy
Second Annual Conference on Carbon Sequestration
Alexandria, Virginia

May 6, 2003

Motivation

- Computer models used to analyze climate policy scenarios typically treat technological change as an exogenous factor, or ignore it altogether
- Assumptions about technology cost and performance can have a strong influence on model results
- This study examines the importance of technological “learning” on the role of carbon capture and storage (CCS) systems as a mitigation strategy, under alternative climate-related policy scenarios

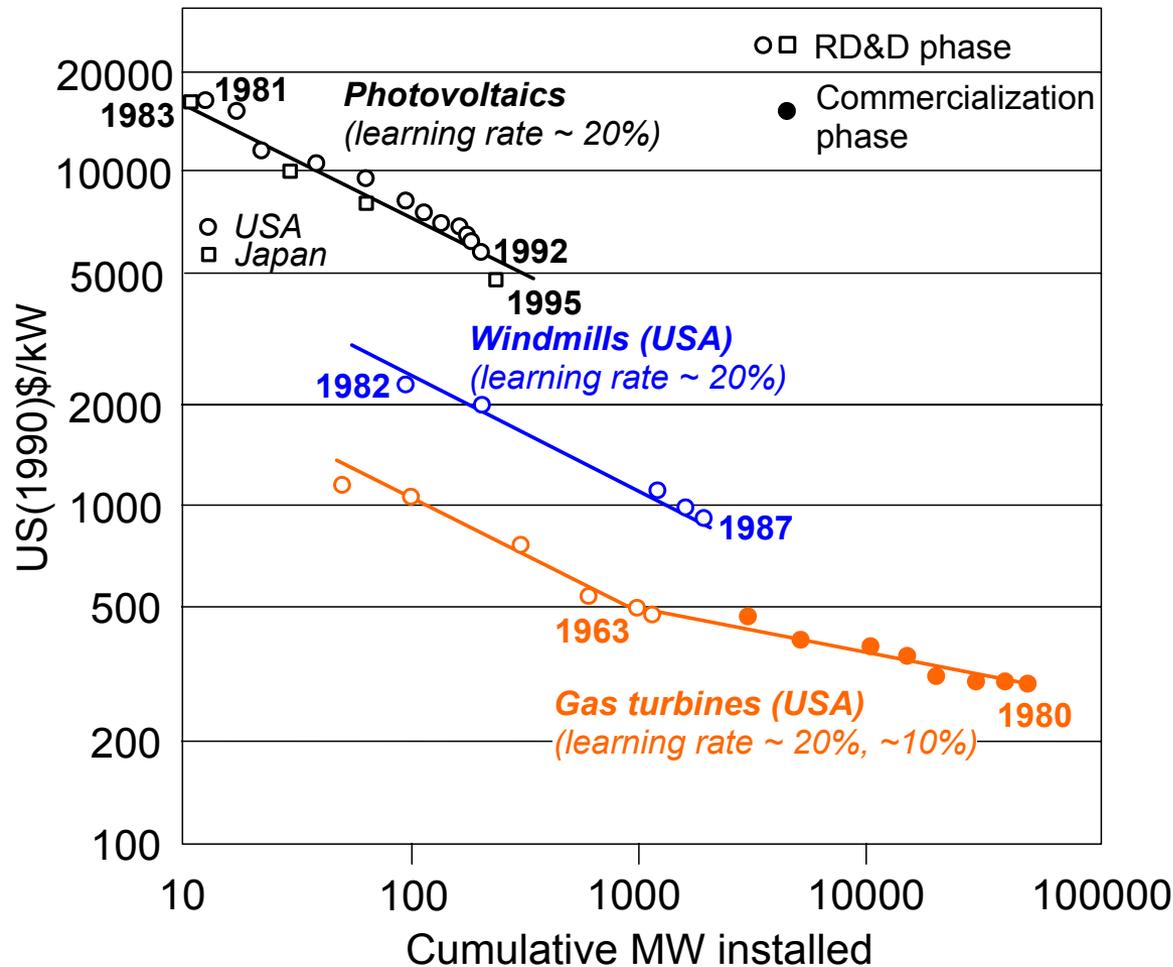
Approach

- Use historical studies to estimate rates of technological innovation for environmental technologies
- Incorporate these findings in a large-scale climate policy model to represent future cost trends for CCS systems
- Assess the policy implications of including technological “learning” for carbon sequestration technologies

Technological Innovation

- Future characteristics of a technology (e.g., costs) are not “autonomous;” they depend on intervening actions
- Improvements in technology are realized through investments in R&D, production, and deployment (resulting in learning-by-doing, and learning-by-using)
- For power generation technologies, cumulative installed capacity is the common proxy for accumulated knowledge

Learning Curves for Electricity Generation Technologies



Learning Curve Formulation

General equation:

$$y_i = ax_i^{-b}$$

where,

y_i = cost to produce i^{th} unit

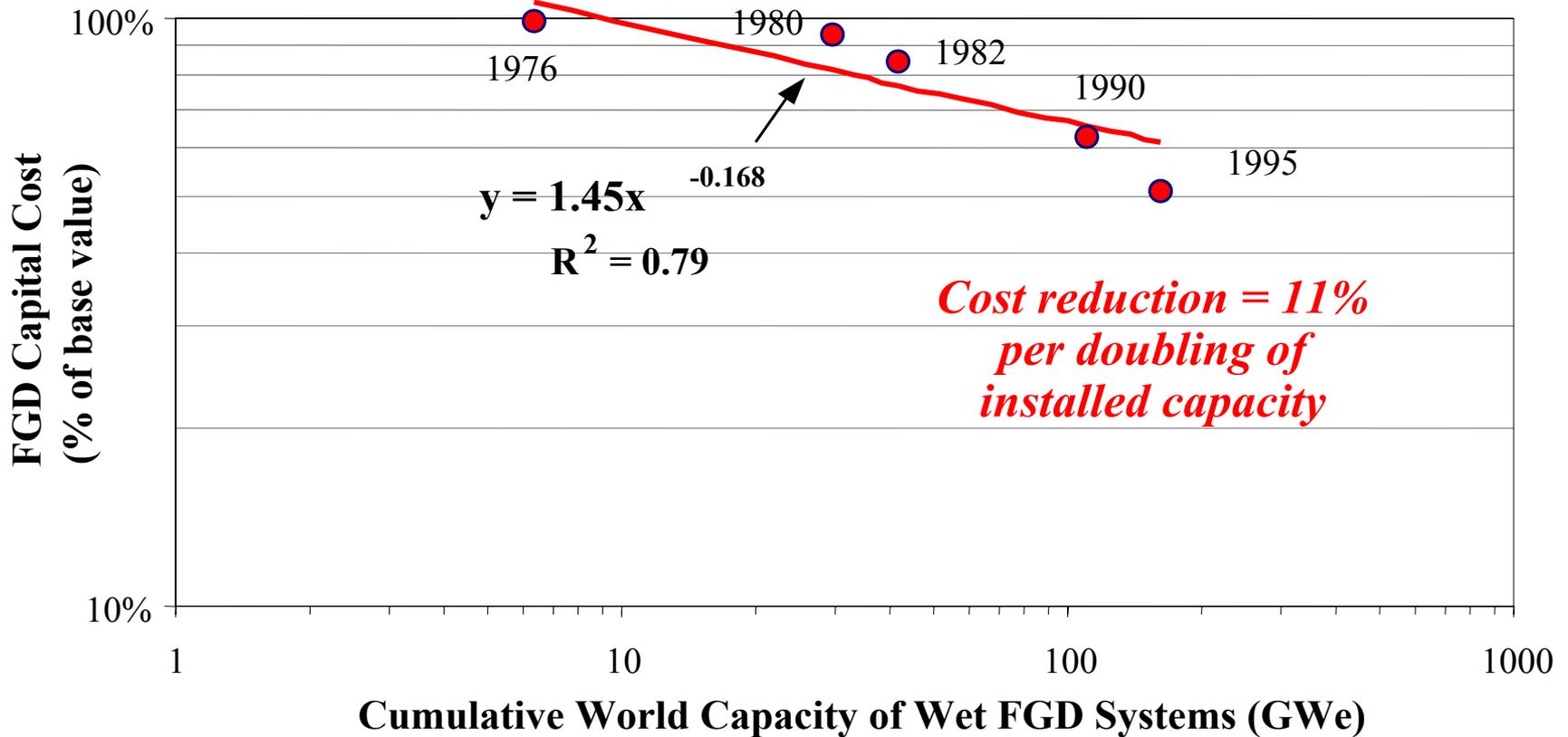
x_i = cumulative production thru period i

b = learning rate exponent

a = coefficient (constant)

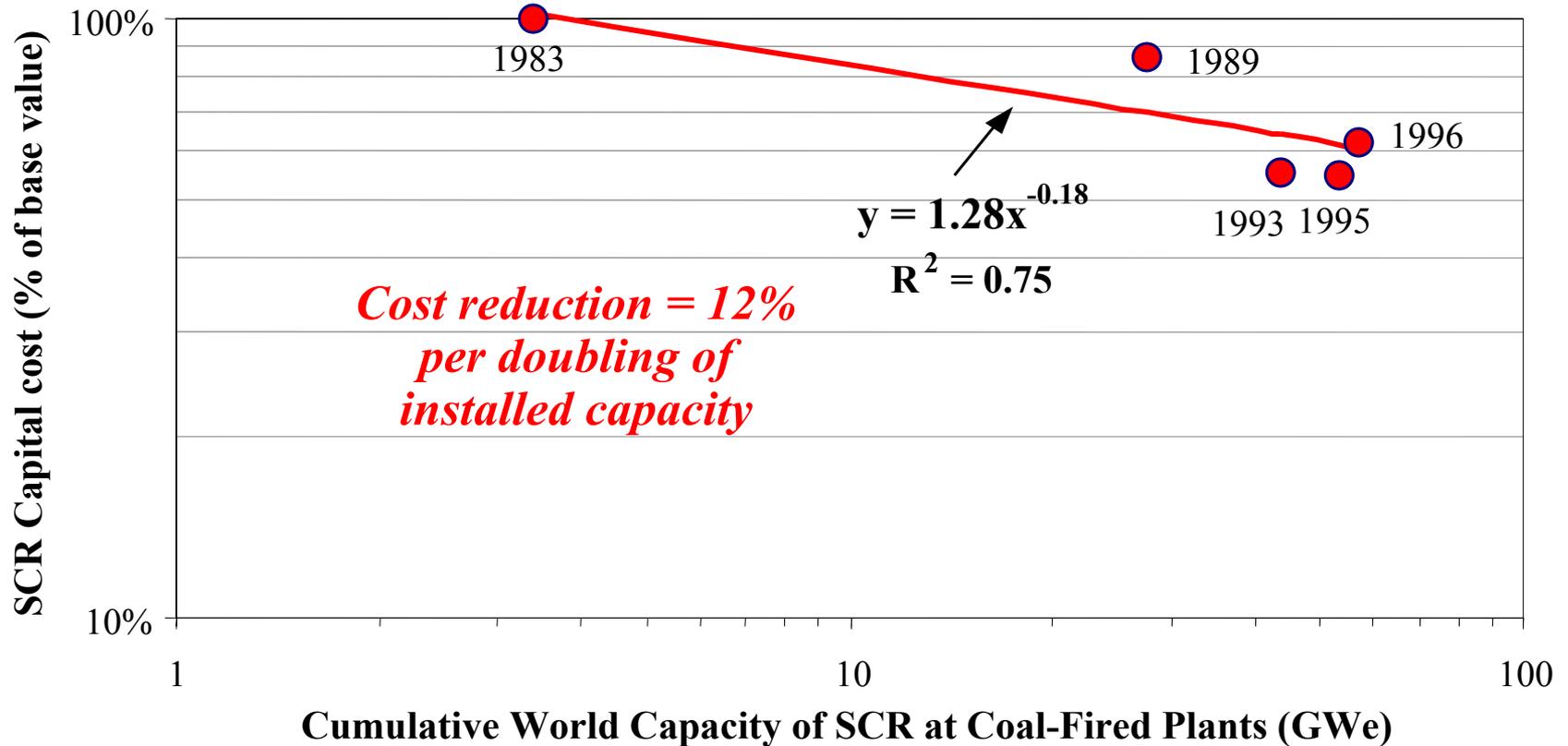
Percent cost reduction for a doubling of cumulative output is called the "learning rate" = $(1 - 2^{-b})$

Normalized Learning Curve for FGD Capital Cost



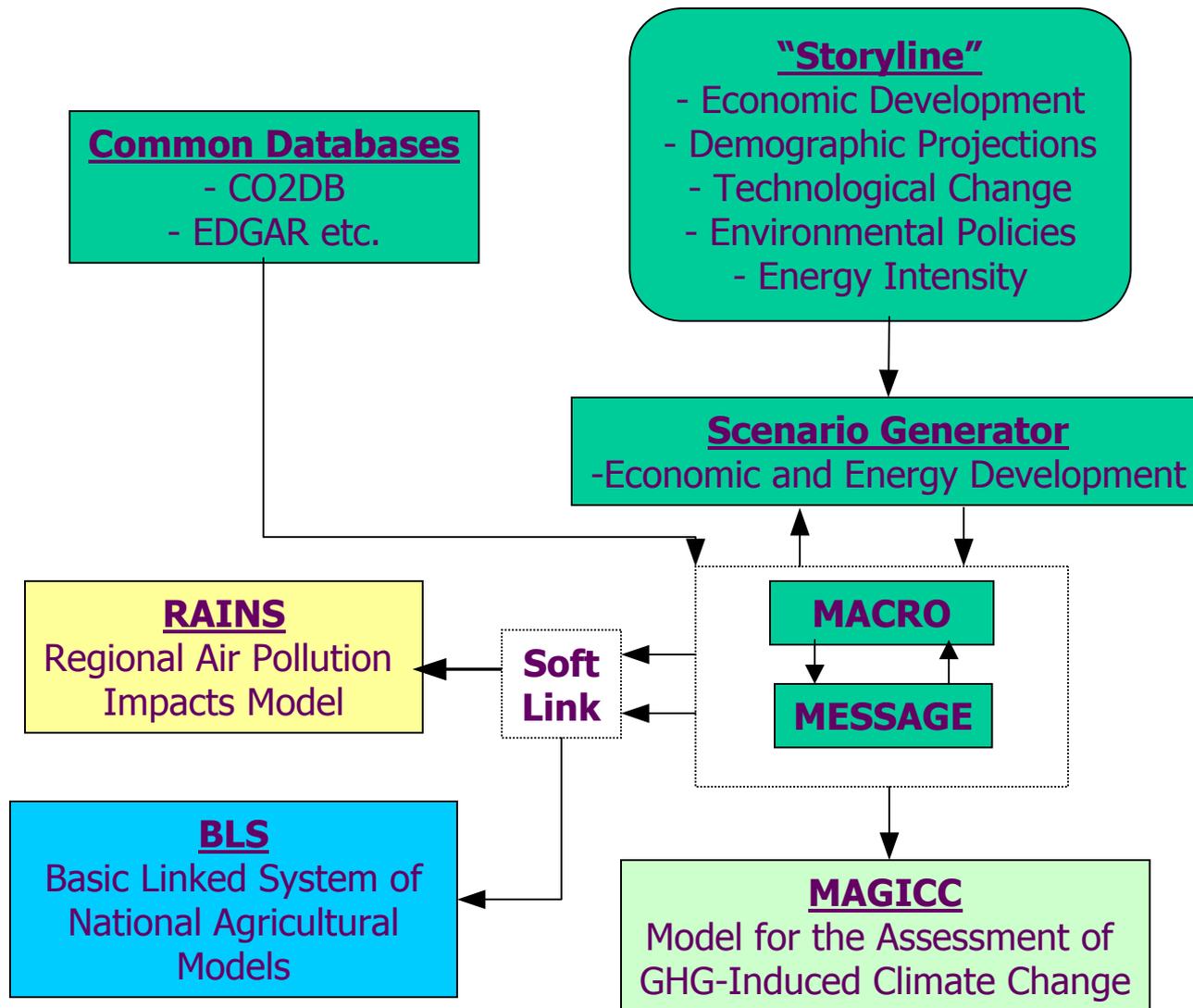
(Based on 90% SO₂ removal, 500 MW plant, 3.5% S coal)

Normalized Learning Curve for SCR Capital Cost

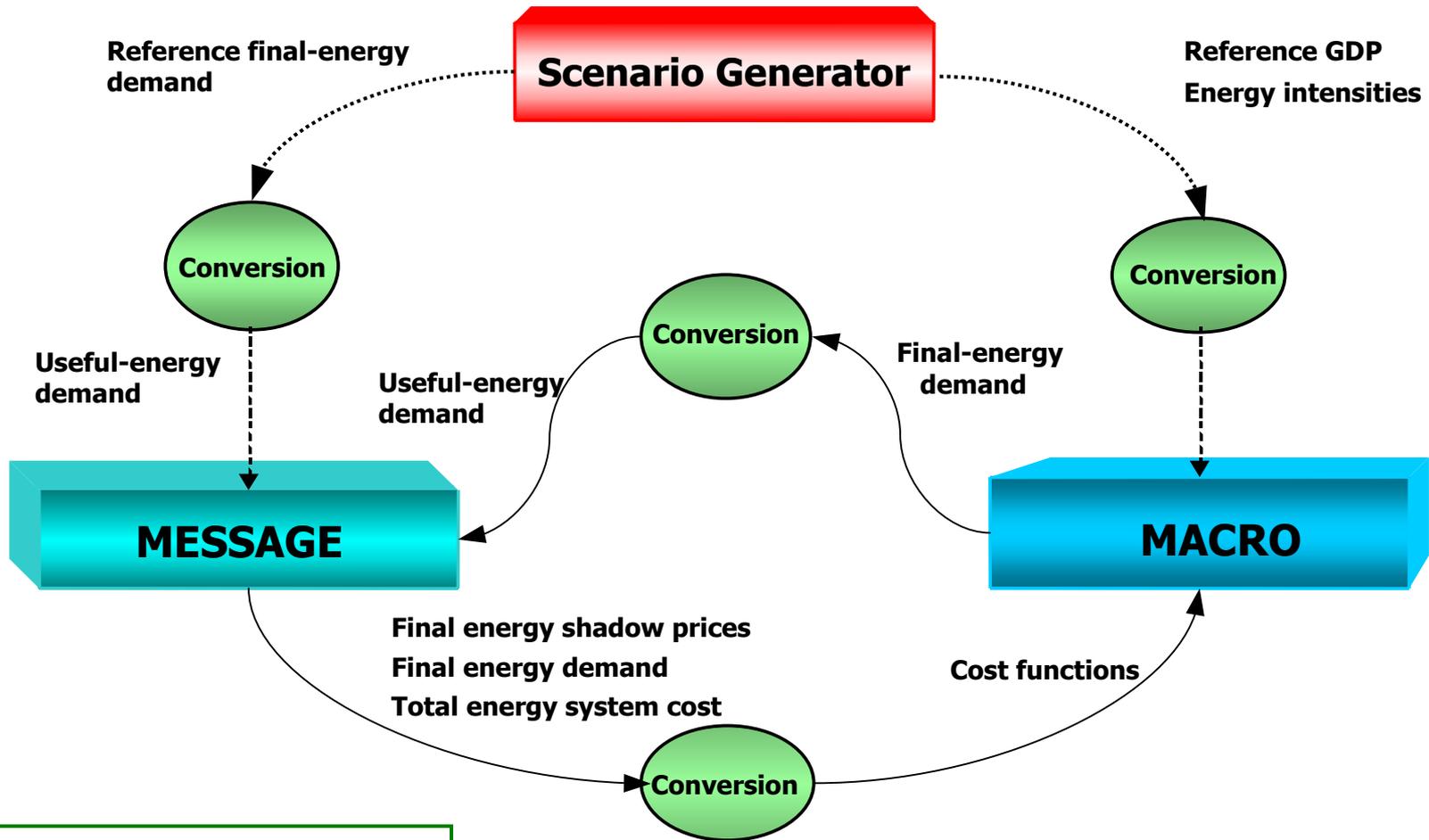


(Based on 80% NO_x removal, 500 MW plant, medium S coal)

The IIASA Modeling Framework

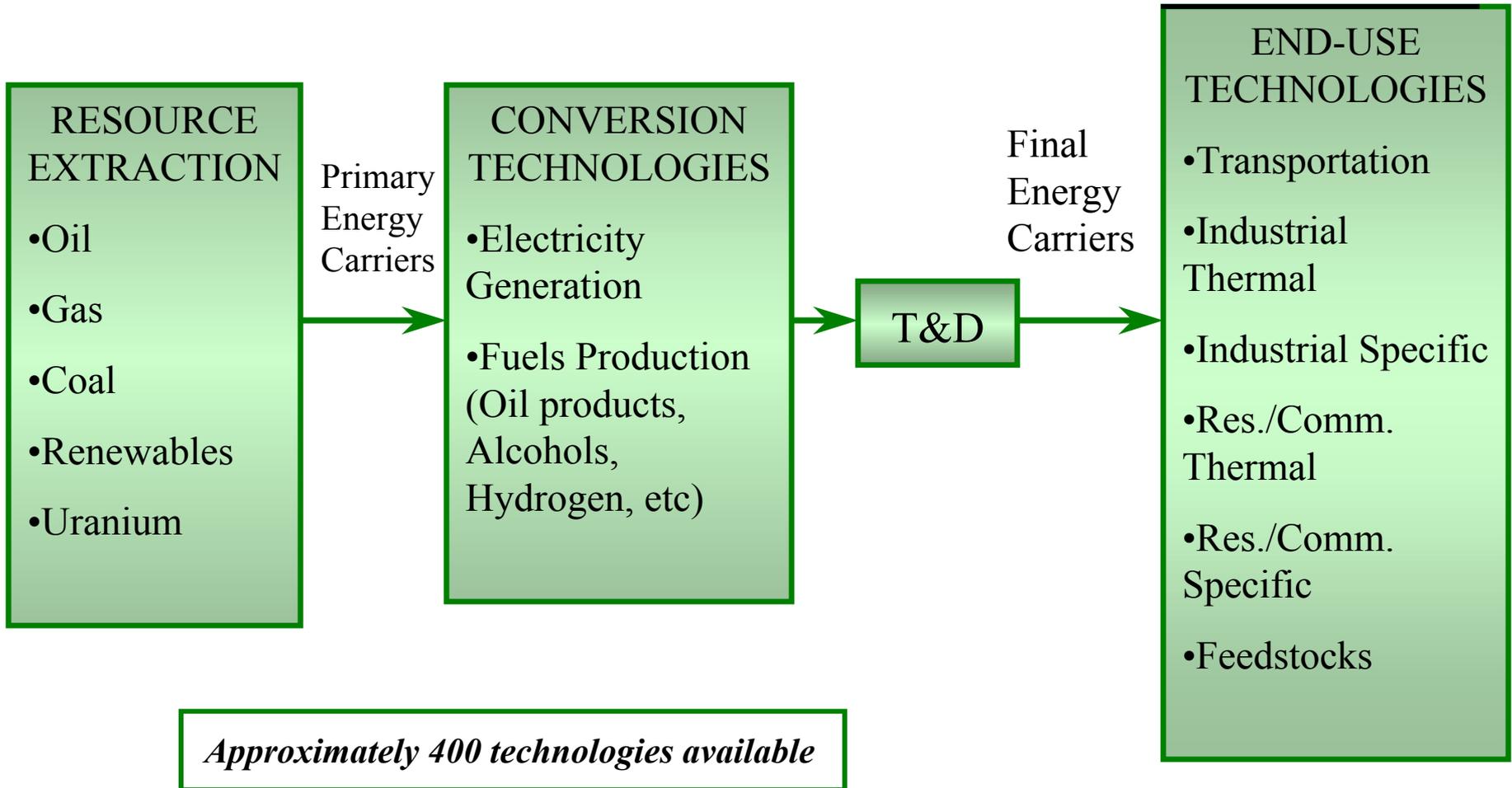


MESSAGE-MACRO



11 World Regions Modeled

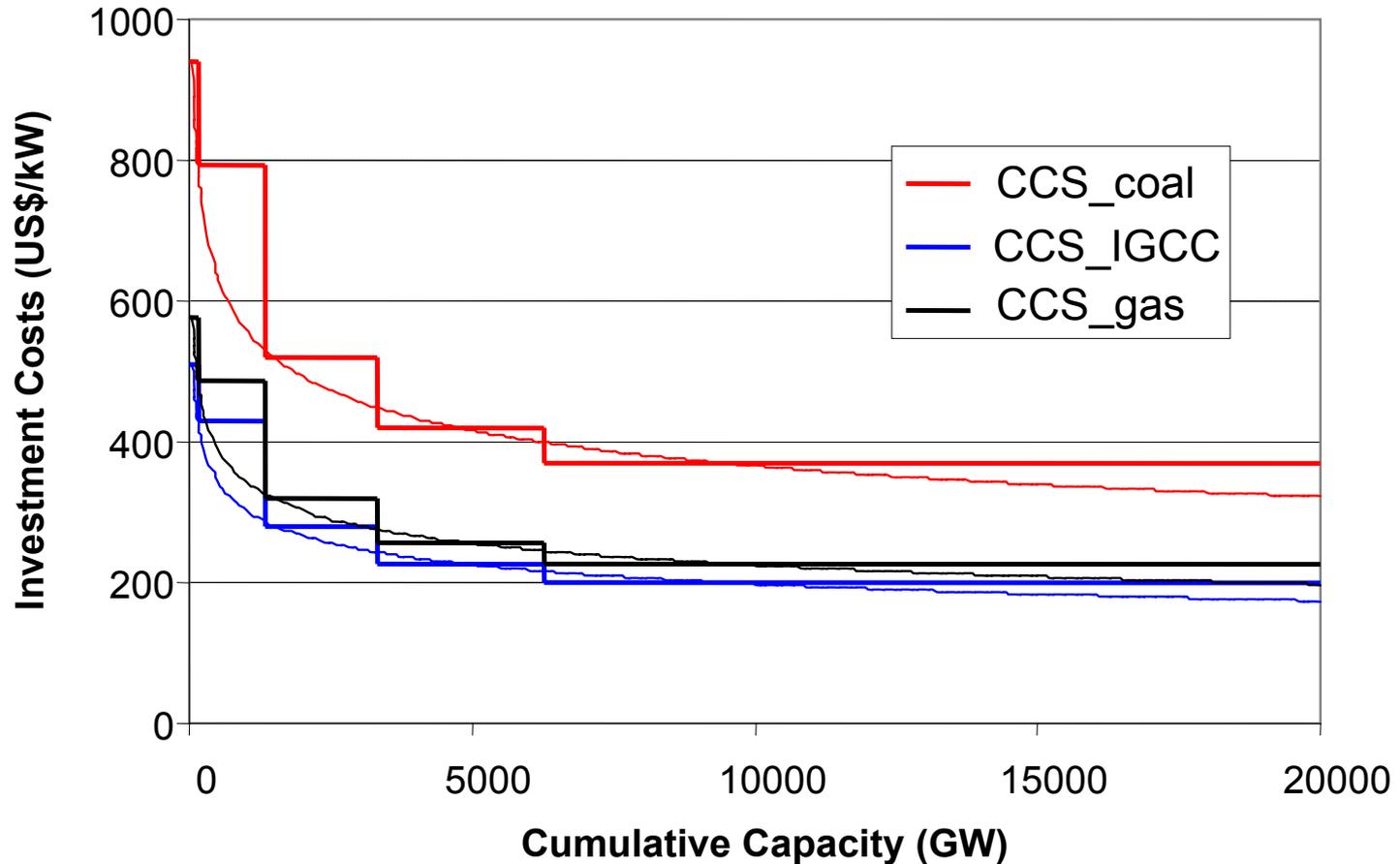
Reference Energy System



Carbon Capture Technologies in MESSAGE

- Electricity Sector
 - Natural gas (NGCC, GT)
 - Coal (PC, PFBC, etc.)
 - Coal IGCC
 - Synthetic fuels production
 - Fossil-based methanol
 - Fossil-based hydrogen
- Chemical separation processes
- Physical separation processes

Endogenous Learning Curves for CCS Clusters



Climate Policy Scenarios

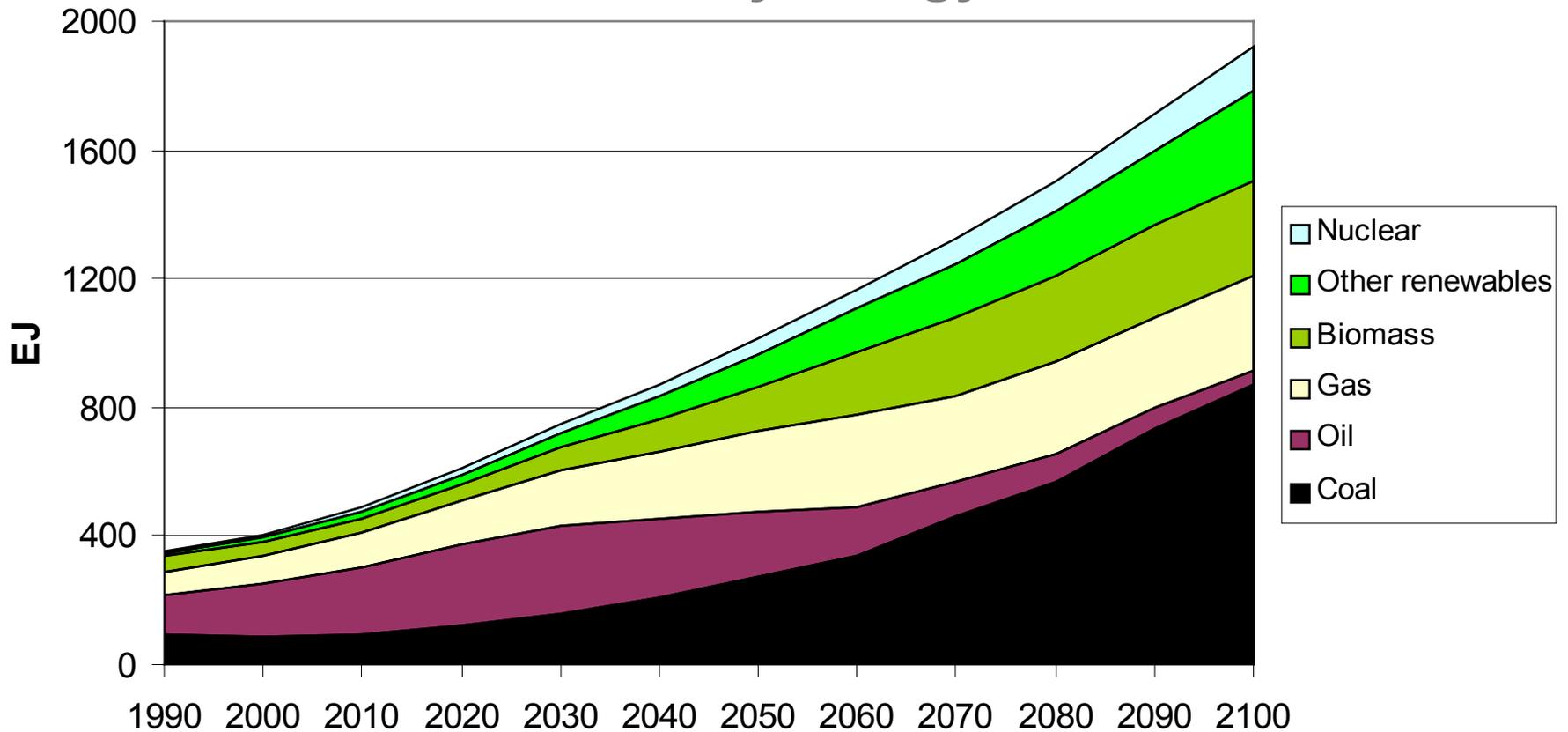
- Model the IPCC-SRES "A2" baseline scenario
 - with and without carbon constraints
 - with and without endogenous "learning" for CCS (assuming a 12% learning rate)
- Model two hypothetical policy scenarios: *
 - 550 ppmv CO₂ by 2100 (global optimization)
 - CCS required for power generation sector only (according to a scheduled phase-in)

The A2 Baseline Scenario

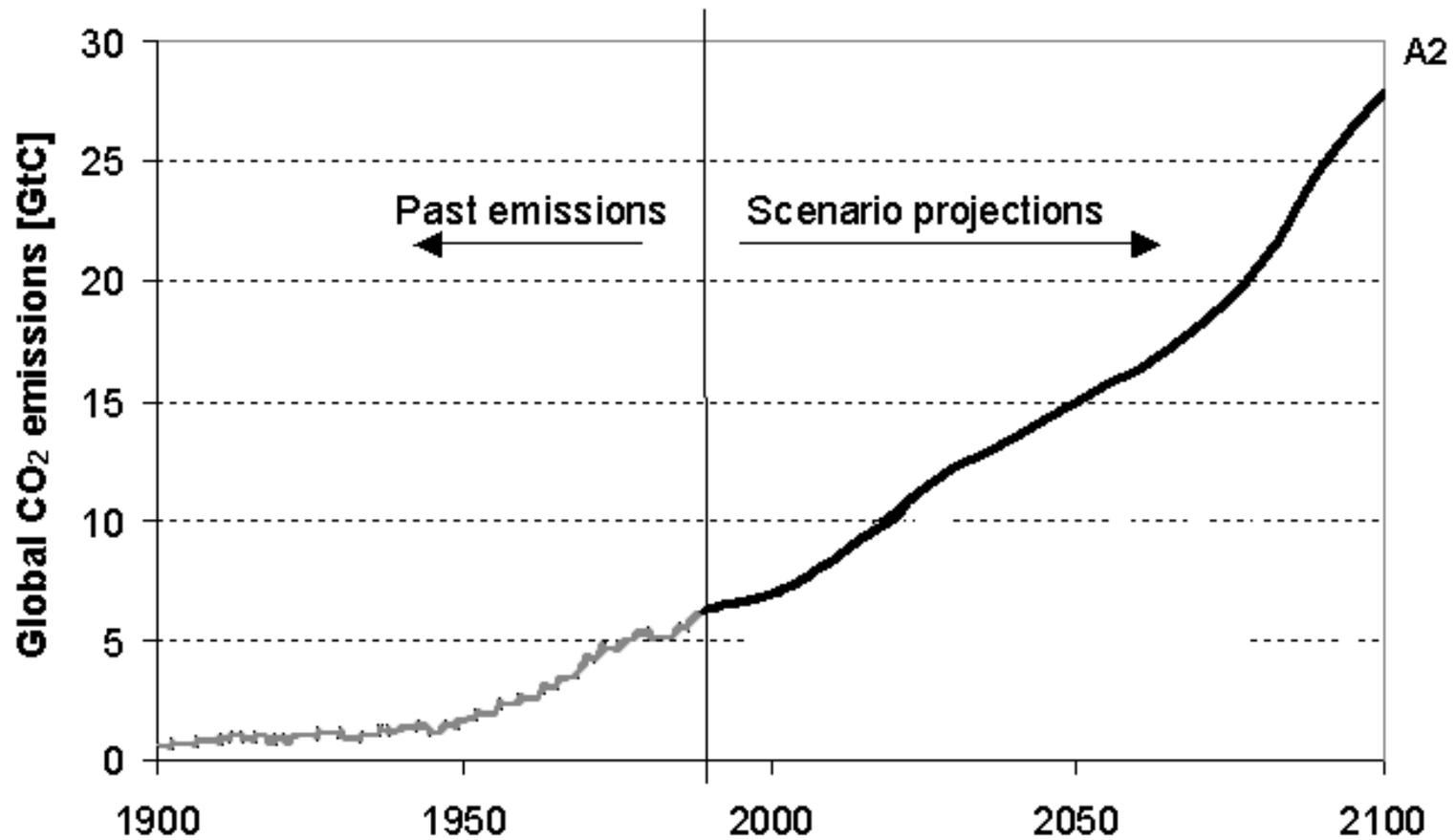
Scenario Parameter	1990	2100
World Population (billion)	5.3	15.1
World GDP (trillion \$1990)	20.9	243
Income Ratio (DEV/IND)	0.062	0.24
World Primary Energy (EJ)	352	1983
Cumulative CO ₂ (GtC)	6.2	1761
Atmos. CO ₂ Conc. (ppmv)	354	783

World Energy Use: Scenario A2

Primary Energy



Energy-Related CO₂ Emissions

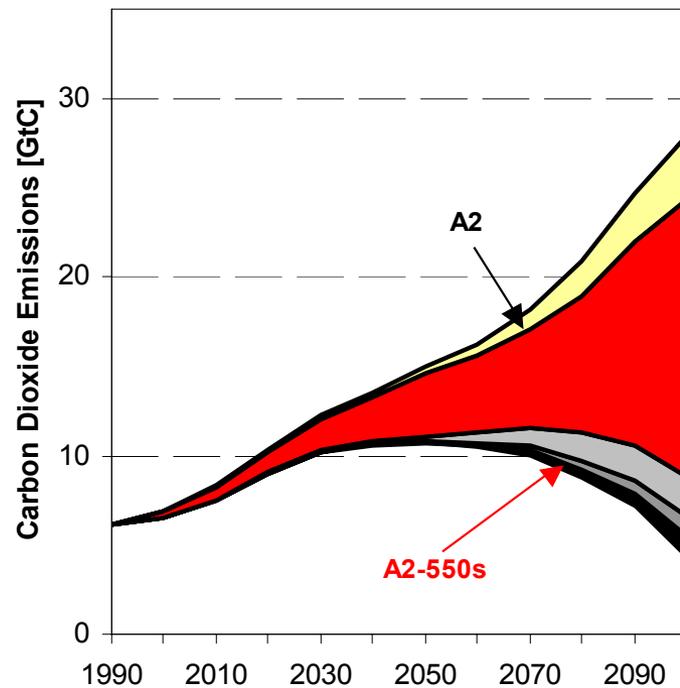


*Results for the
550 ppm stabilization
scenario
(global optimization)*

Main Mitigation Measures

(stabilize atmospheric CO₂ at 550 ppmv by 2100)

No learning for CCS

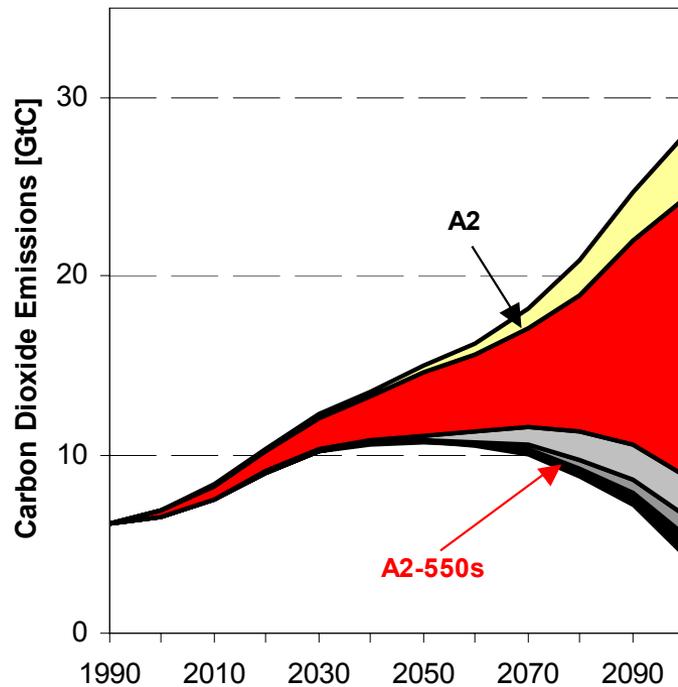


- demand reduction
- fuel switching (mainly shifts away from coal)
- scrubbing and removal - synthetic fuels production
- scrubbing and removal - power sector (natural gas)
- scrubbing and removal - power sector (coal)

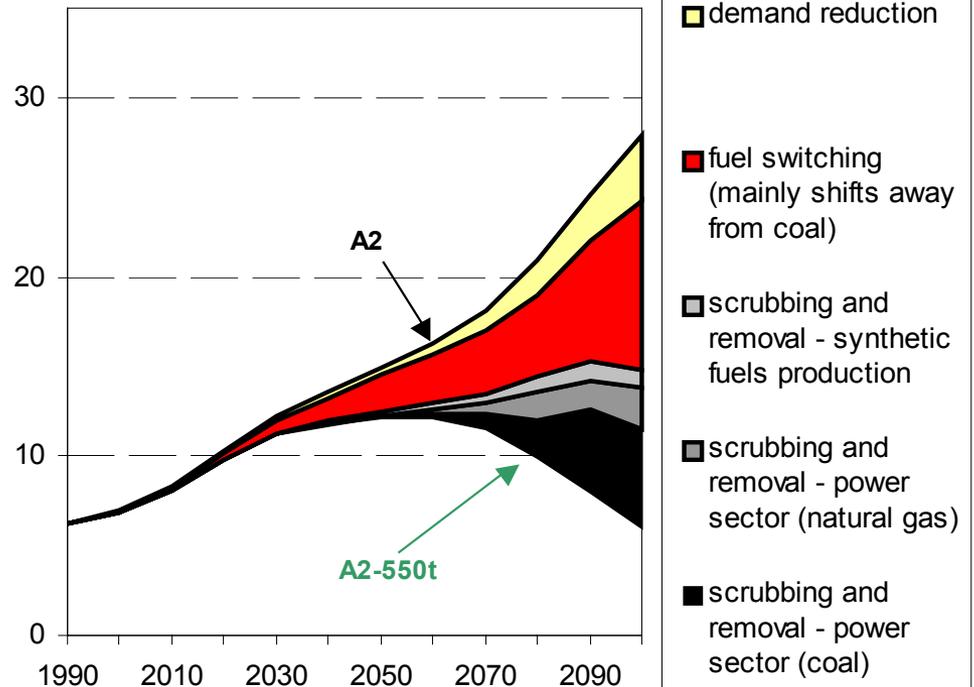
Main Mitigation Measures

(stabilize atmospheric CO₂ at 550 ppmv by 2100)

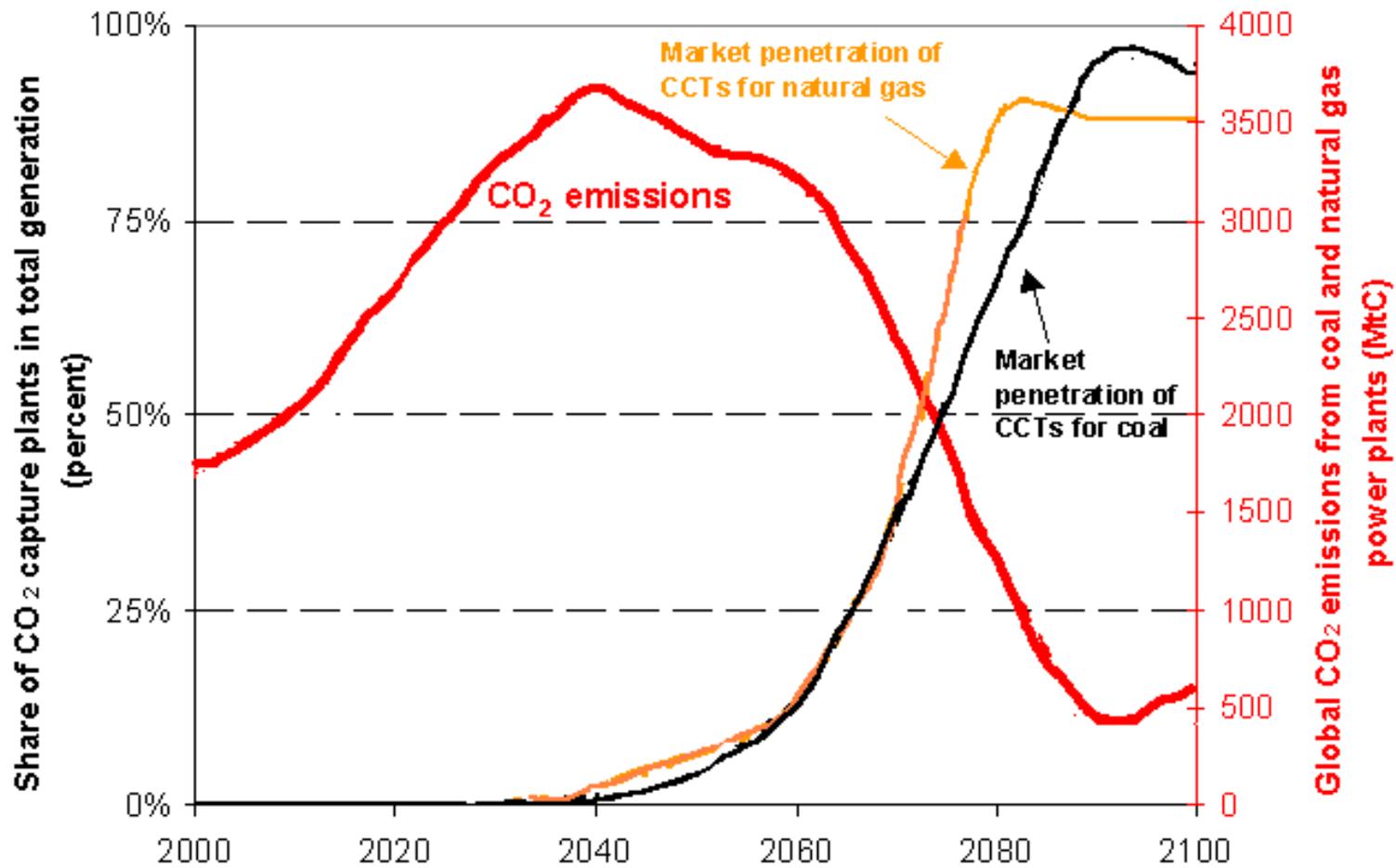
No learning for CCS



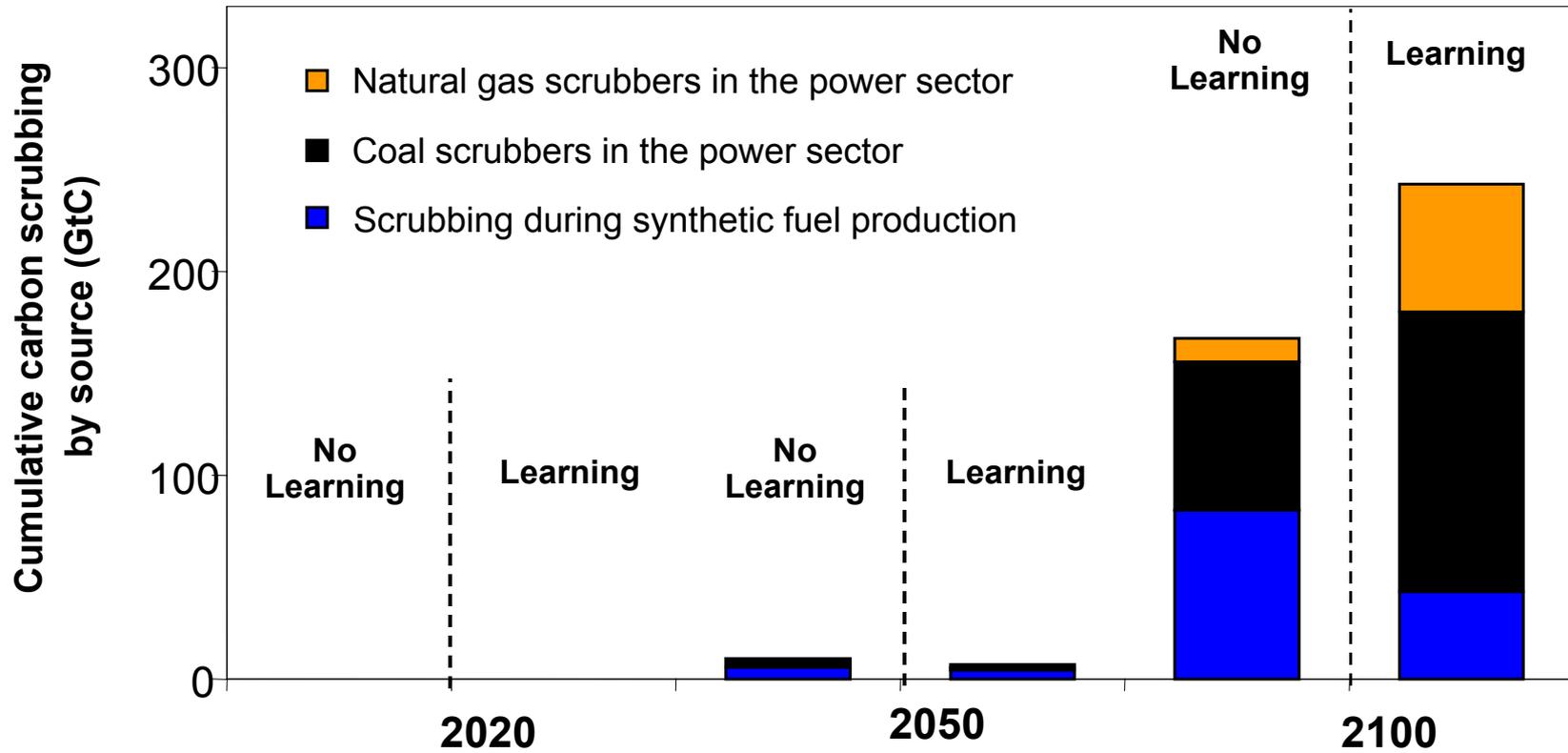
Learning for CCS



Market Penetration of CCS (no learning case)



Cumulative Carbon Sequestration



Average Carbon Tax (1990 US\$/t)

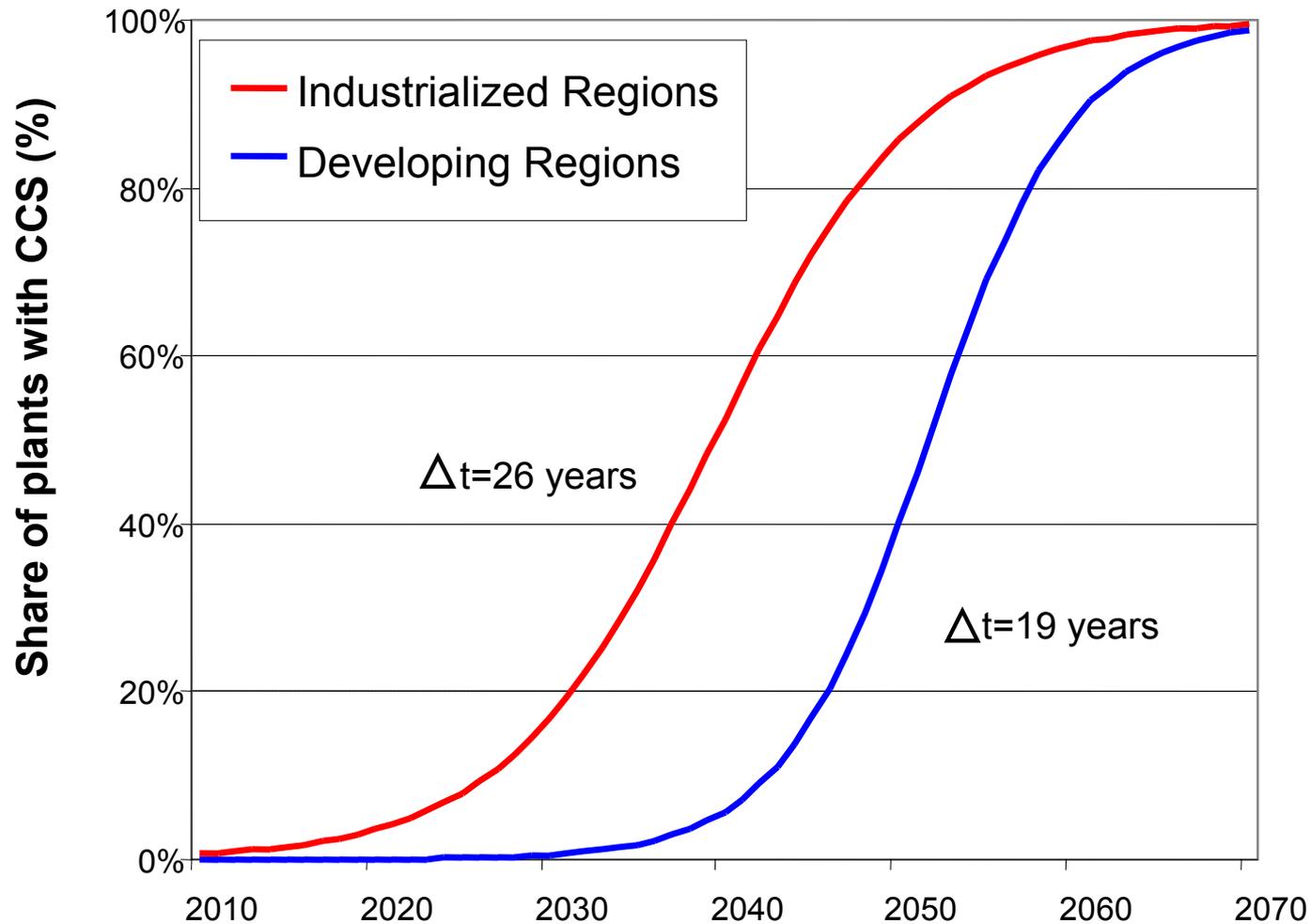
Scenario	2020	2050	2100
A2-550 - No Learning	25	82	496
A2-550 - Learning	19	27	490

*Results for the
Technology-Based
Policy Scenario*

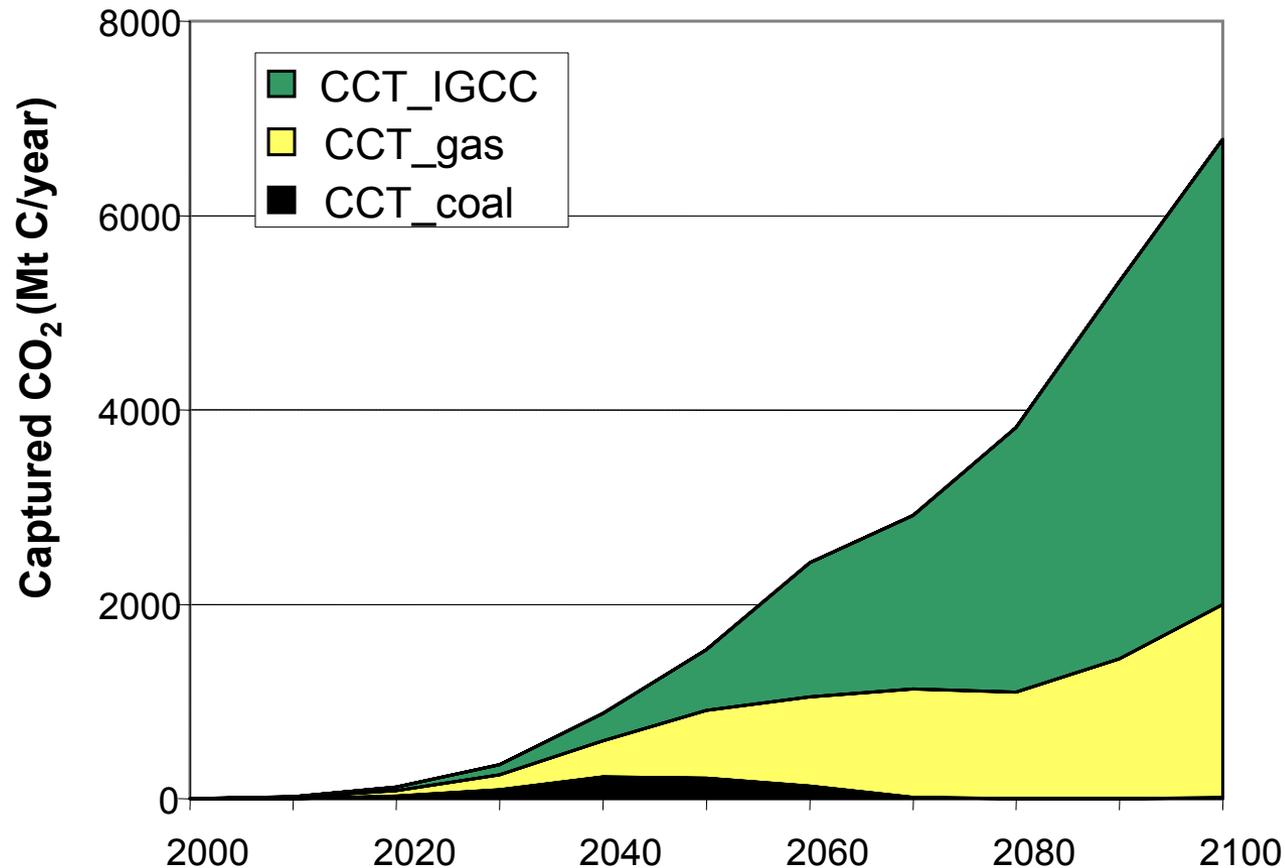
Technology-Based Policy Scenario

- An illustrative “technology forcing” policy is imposed on fossil-fueled power plants (only)
- A minimum time-increasing share of the total fossil-fuel capacity is required to capture and store 90% of potential CO₂ emissions
- Different schedules for industrialized and developing countries
- Cases are examined with and without endogenized learning for CCS technologies

Minimum Deployment of CCS for Fossil-Fuel Power Plants

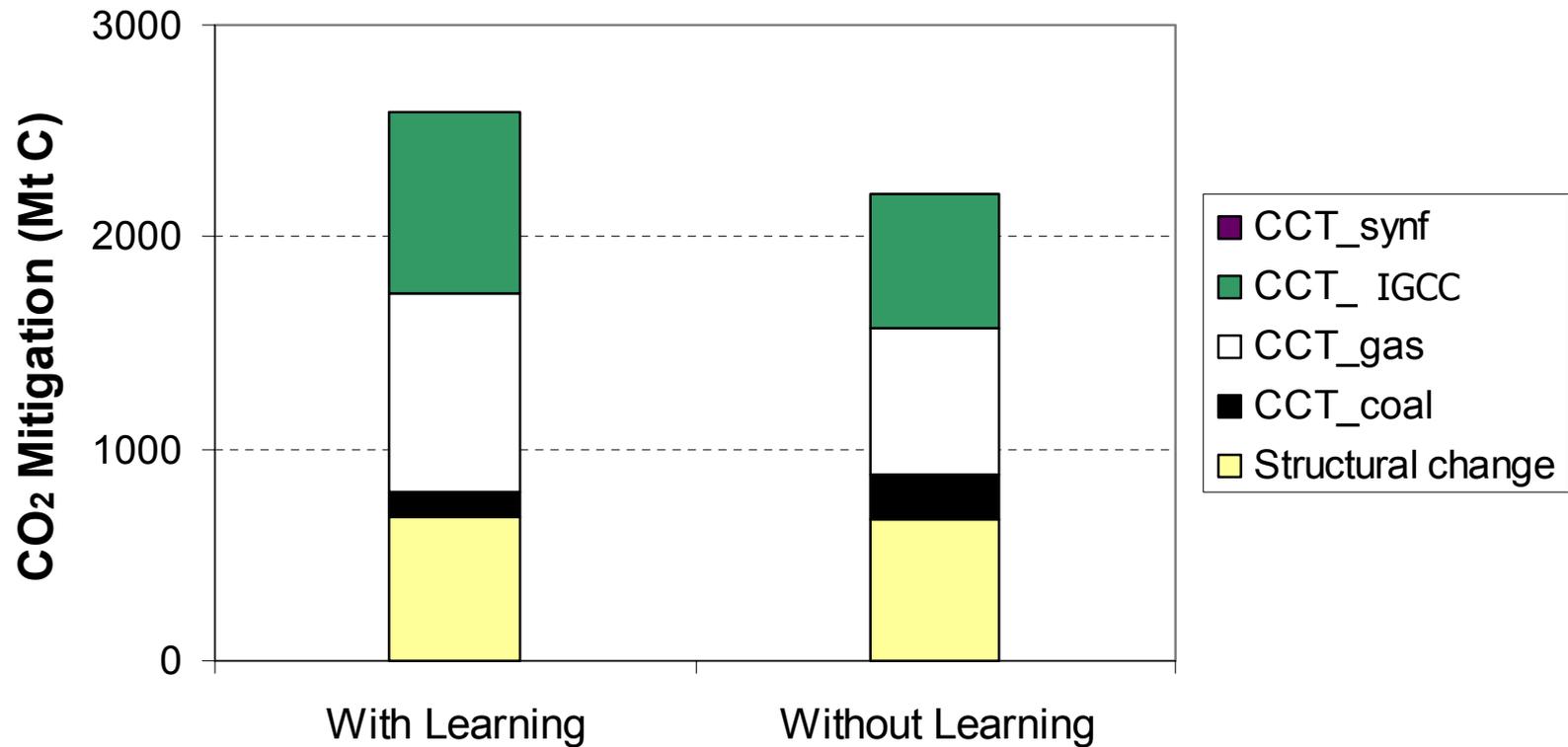


Captured CO₂ in Electricity Sector A2-CCT Scenario (no learning)



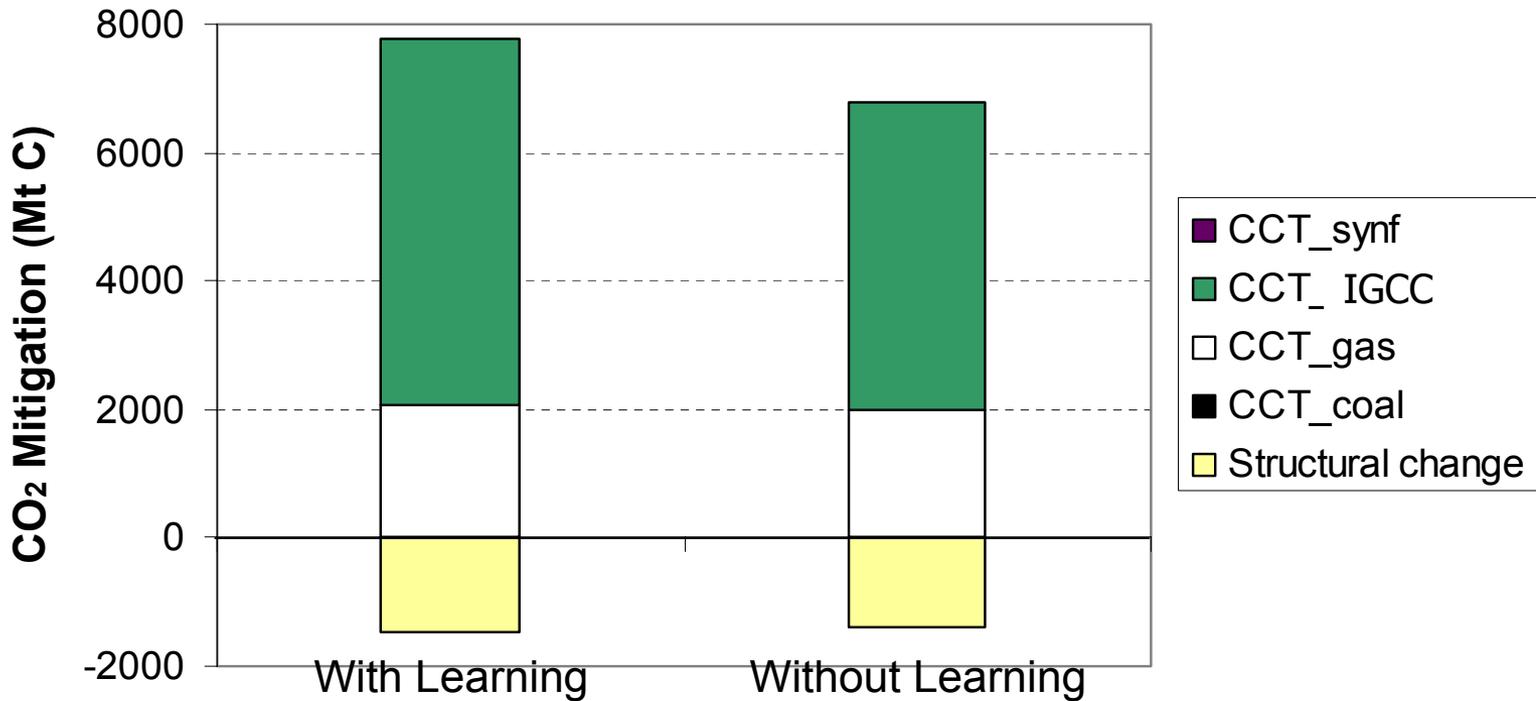
CO₂ Mitigation in 2050

year 2050

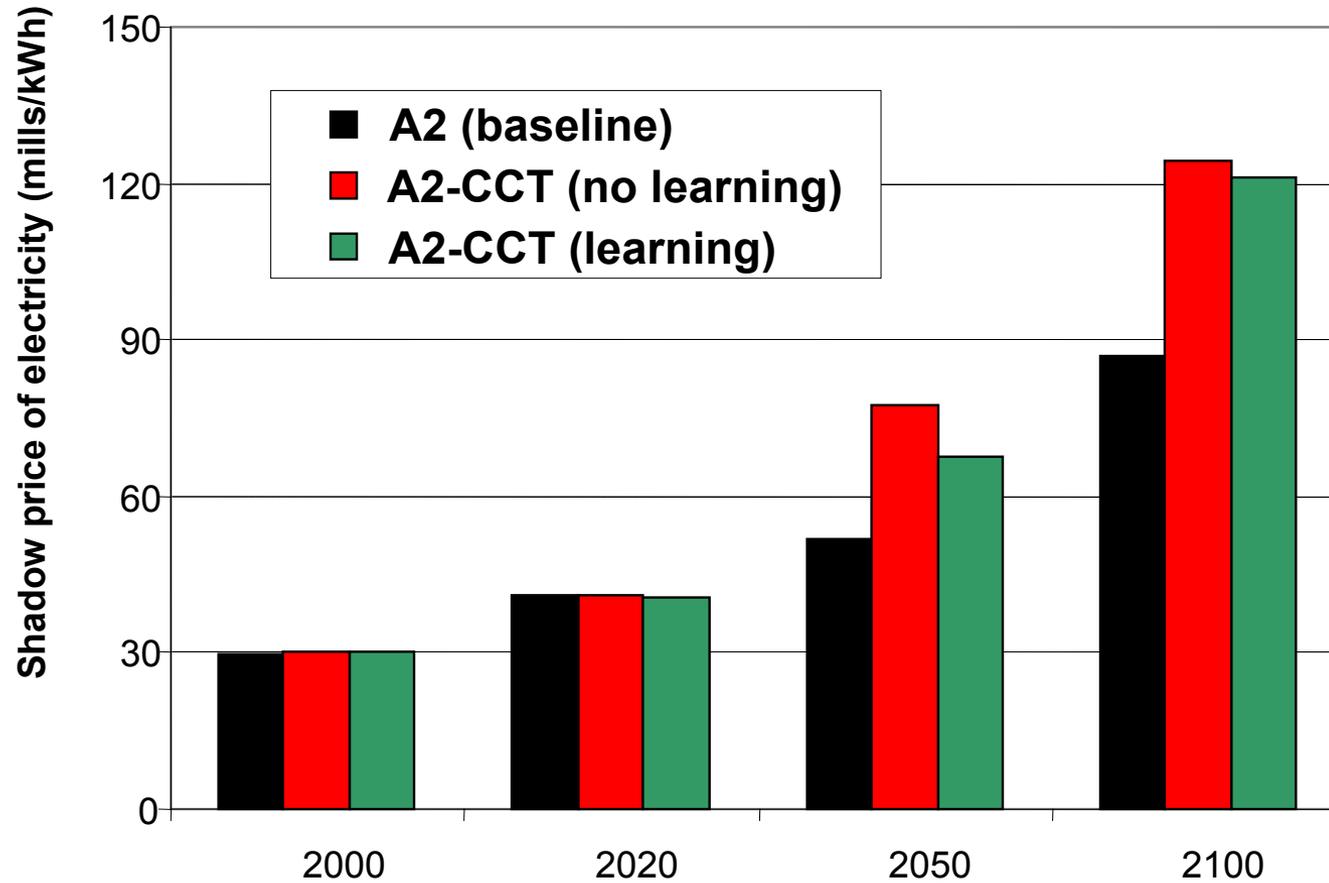


CO₂ Mitigation in 2100

year 2100



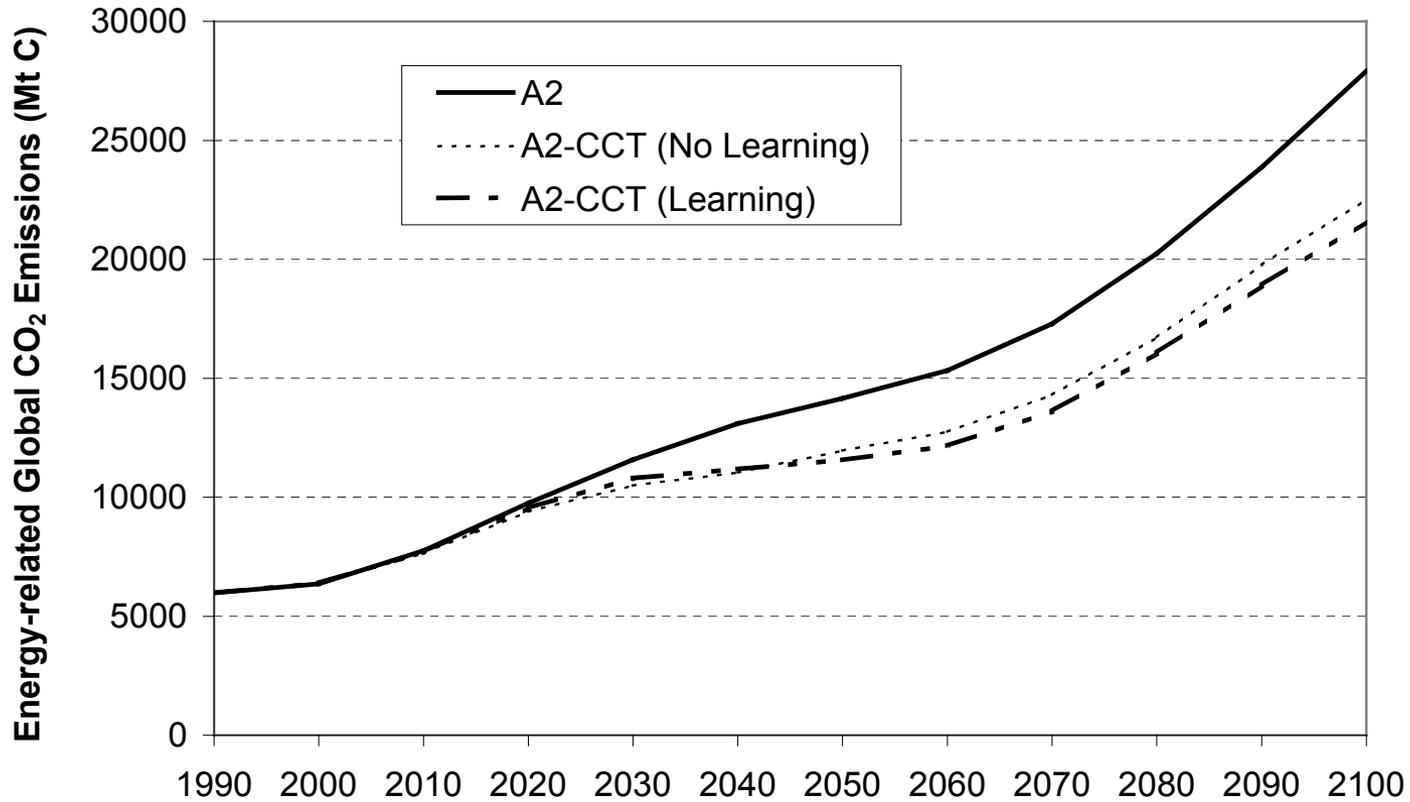
Electricity Prices



Average Cost of Carbon Abatement (1990 US\$/t)

Scenario	2020	2050	2100
A2-CCT - No Learning	103	98	84
A2-CCT - Learning	100	87	56

Global Energy-Related CO₂ Emissions



Conclusions

- Consideration of technological learning can have a significant influence on the expected role of CCS technologies, and the cost of alternative climate mitigation policies
- The magnitude and timing of impacts depends strongly on the policy scenario, and the reference case assumptions
- More work is needed to better understand and model the key factors that influence technology innovation, especially for environmental technologies like carbon capture and storage