Technology R&D as Greenhouse Insurance

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Uncertainty and Emissions Control

- Uncertainty about how emissions today will cause damages tomorrow.
- But, we are learning more and more.
- Uncertainty, learning, and adaptation impact current decisions
- General conclusion: Uncertainty + Learning = less control of emissions.
 - Kolstad
 - Ulph & Ulph
 - Manne & Richels
 - Baker



What about R&D?

- R&D planning is complicated by different programs
 - Solar PVs, windpower
 - Efficiency of coal-fired electricity
 - Gas turbines
 - Sequestration

How does optimal R&D change with
Increasing risk and learning about climate damages
choice of R&D program



Overview

- Explore in a top-down framework the response of optimal R&D to increasing risk
- Theoretical results indicate that there is no single directionality:
 - How R&D is modeled matters, and
 - How increasing risk is modeled matters.
- Confirm this in a IAM.
- Along the way, discuss approaches for representing R&D effects in top-down models.



Agenda

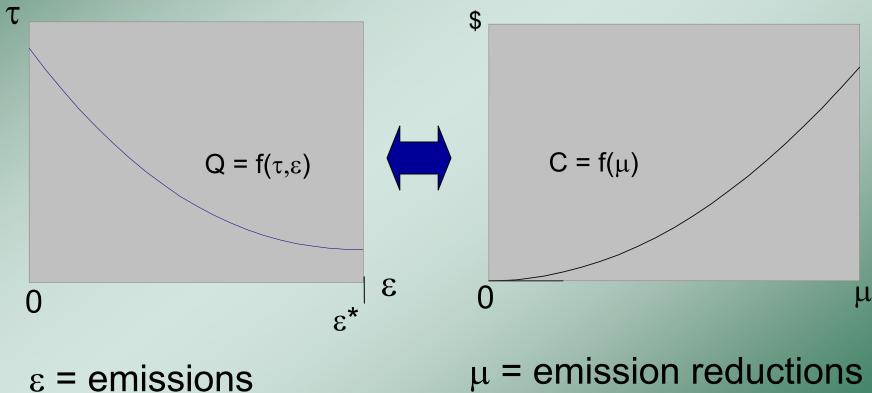
Introduce Technological Change
 Introduce Increasing Risk
 Discuss Theoretical Model and Results
 Discuss Implementation in DICE
 Conclusions



How Might R&D Change Technology

Production Function

Abatement Cost Curve



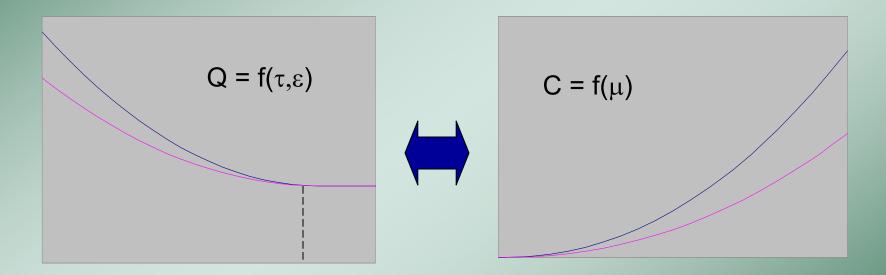
 τ = "standard" inputs



R&D and Technology: One Example

Production Function

Abatement Cost Curve



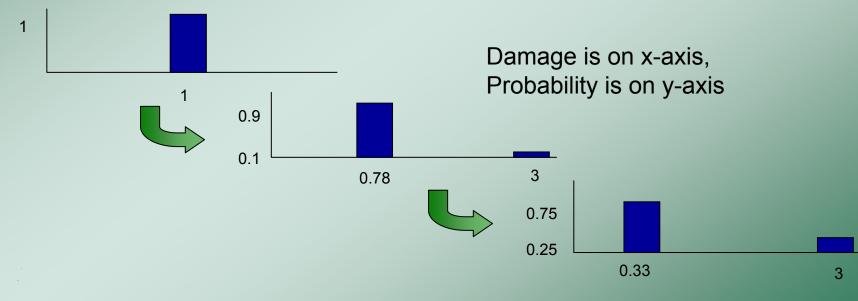
Many ways in which R&D might alter technology



What is Increasing Risk?

"Risk" – "uncertainty" – "Mean-preserving-spread"

- See for example Rothschild & Stiglitz 1970,1971.
- NOT A CHANGE IN THE MEAN!



Many ways to create a mean-preserving spread.



Theoretical Model

Two period model

Cost of R&D

- R&D investments in first period;
- Abatement and improved technology in the second
- Initial uncertainty regarding the damages from climate change
 - Resolved at the start of the second period

Expected Costs of Abatement and Damages Assuming Optimal Abatement Behavior

 $g(\alpha) + E_z \min c(\mu, \alpha) + D(\mu, z)$ α



Theoretical Results

 $\min_{\alpha} g(\alpha) + E_z \min_{\alpha} c(\mu, \alpha) + D(\mu, z)$

- Proposition: For every R&D program, optimal R&D decreases with some increases in risk.
 - (Allowing for "Full abatement")
- The converse is not true some R&D programs will always decrease in risk.
- Individual R&D programs will react differently to an increase in risk.
- It is crucial to model the specific program.

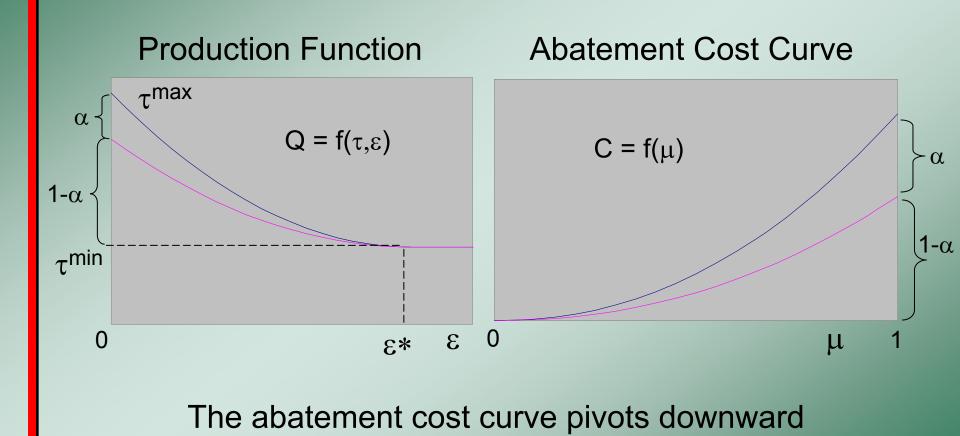


Integrated Assessment Model

- William Nordhaus's DICE
- Optimal Growth + Climate Model
- Added uncertainty, using stochastic programming
- Added R&D as a decision variable
 - One time decision in 1st period before learning
 - Cost reduction implemented in 50 years, after learning.



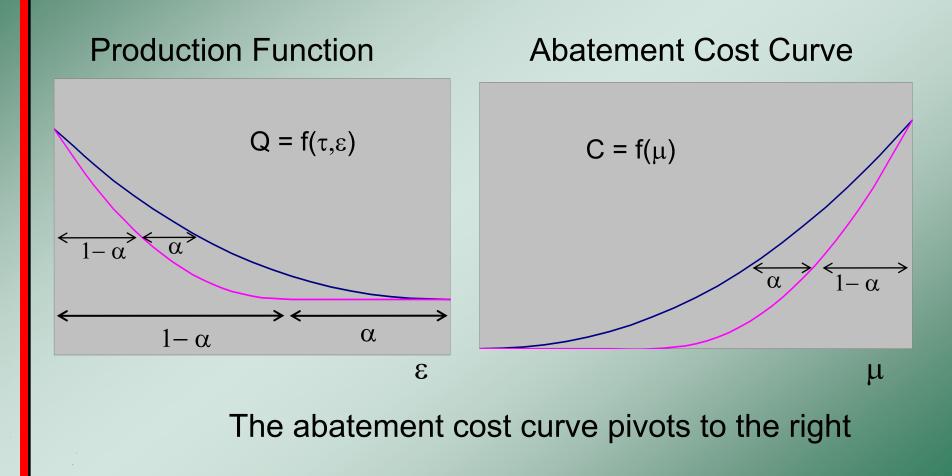
Two R&D Programs: (1) Cost Reduction



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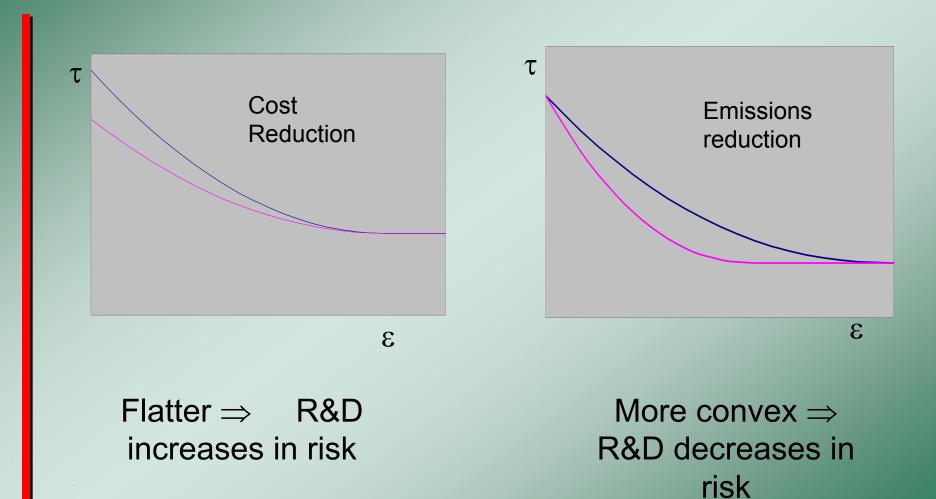


Two R&D Programs: (2) Emissions Reduction





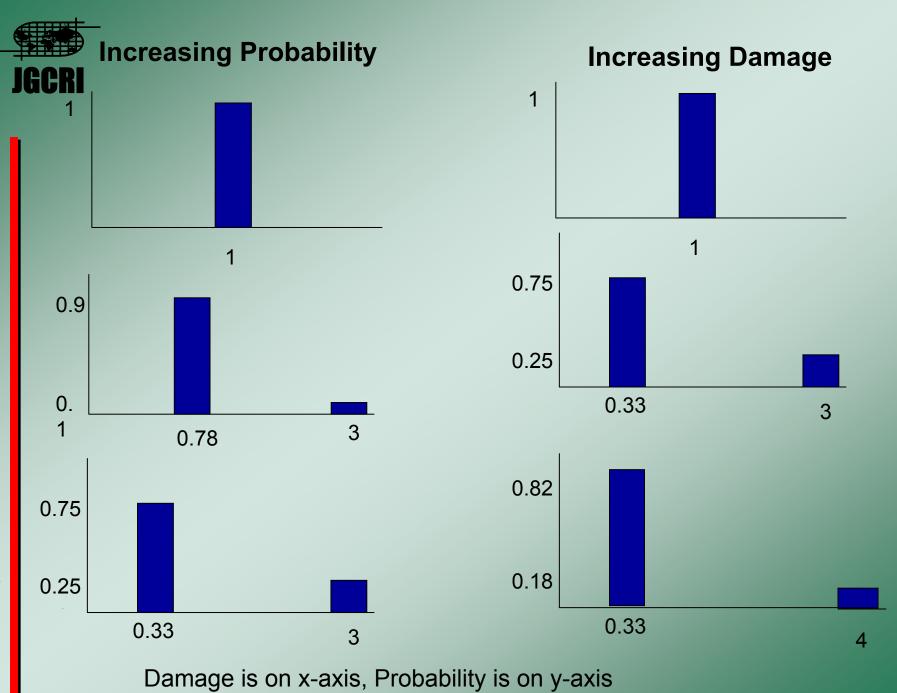
R&D impacts convexity of cost curve / production function

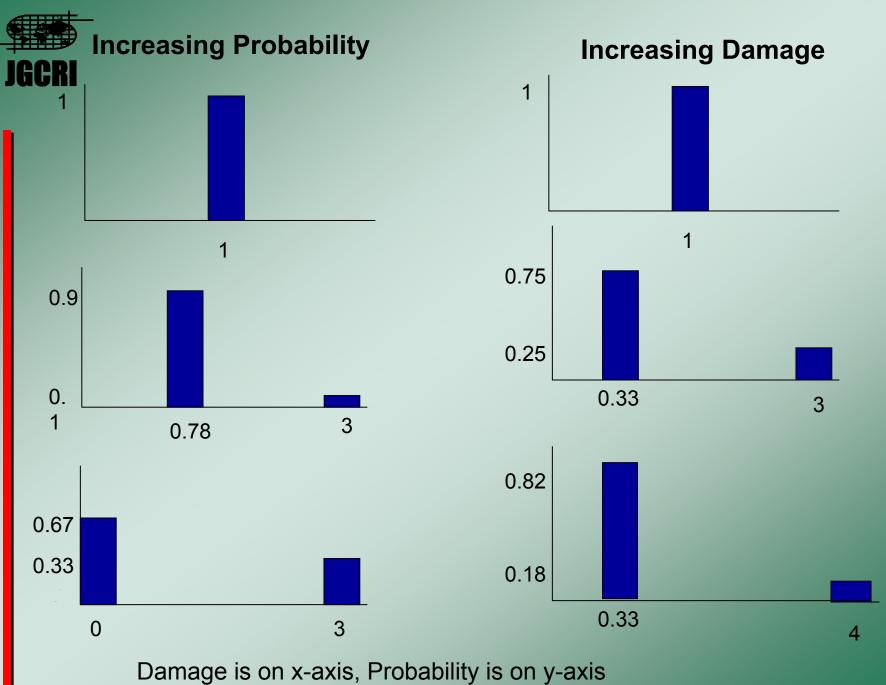


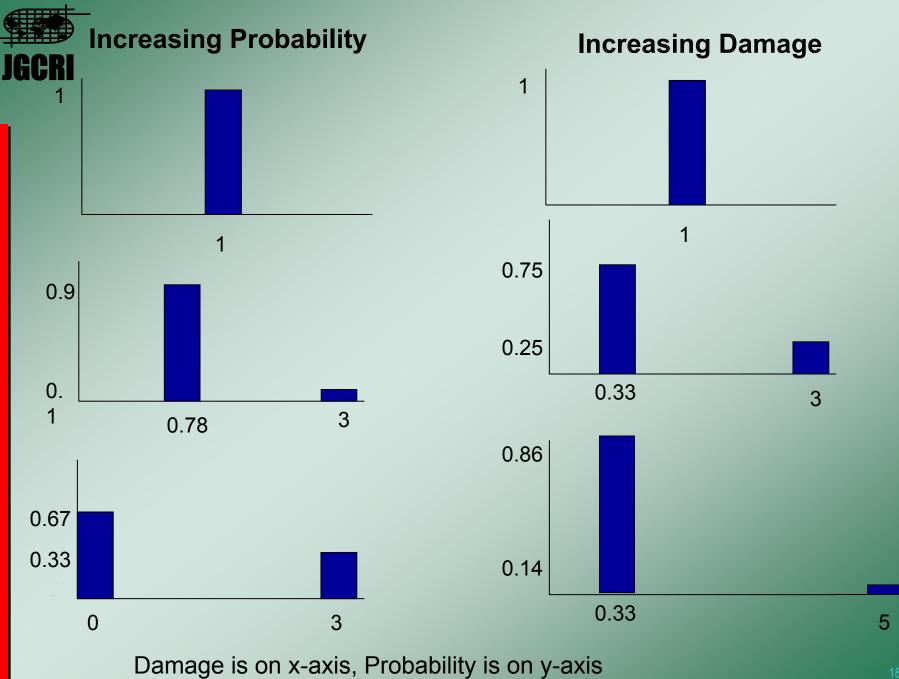


2 Types of increasing risk

Increasing Probability	certain	low	med	high
Probability of high damage	0	.018	.050	.083
Value of high damage	-	.042	.042	.042
Value of low damage	.0035	.0028	.0015	0
<i>Increasing Damage</i>	certain	low	med	high
Probability of high damage	0	.018	.013	.0024
Value of high damage	-	.042	.057	.30
Value of low damage	.0035	.0028	.0028	0028

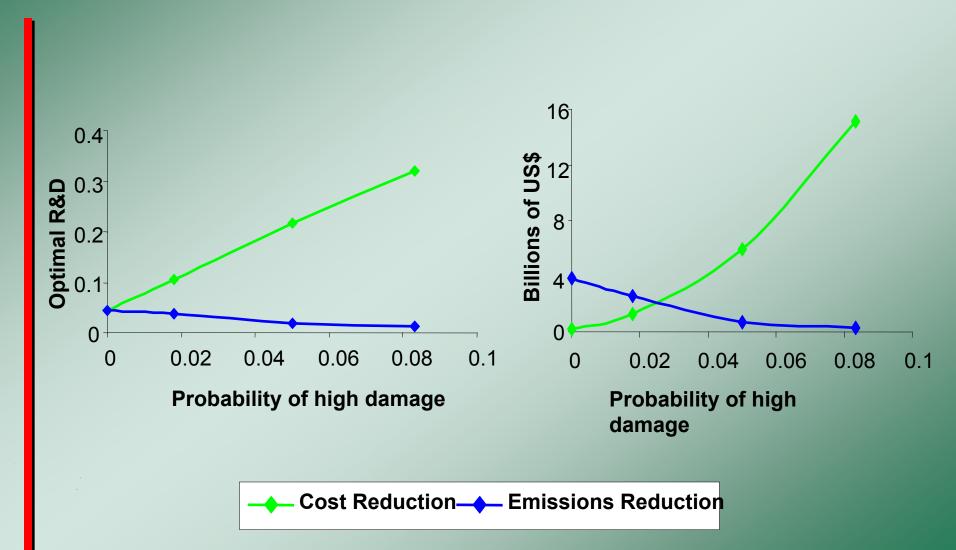






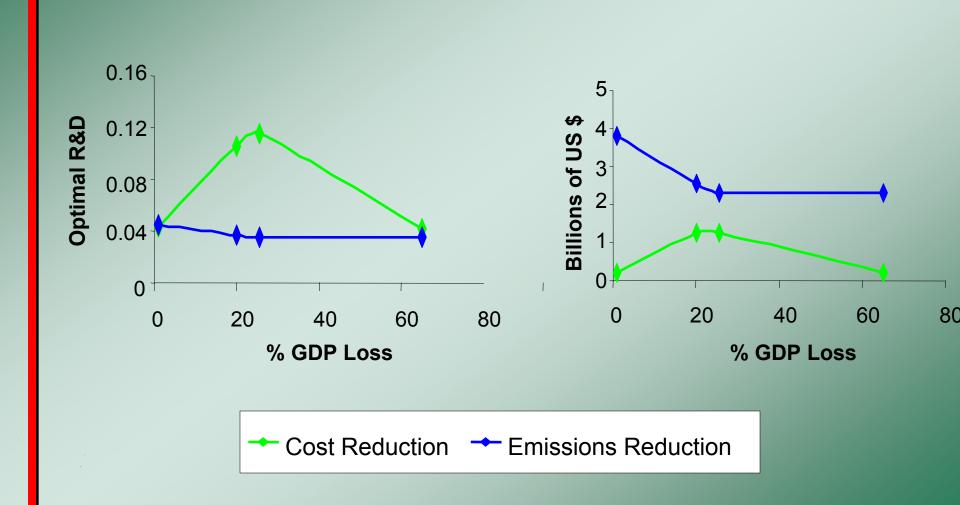


Results – Increasing Probability





Results – Increasing Damages





Conclusions

R&D can be a hedge against uncertainty.

▶ But, it depends on what kind of R&D.

 R&D into reducing the cost of low carbon alternatives

And what kind of risk.

 Increasing the probability of needing very low carbon technologies, rather than considering higher levels of damages.



DICE equations

$$Q_{t} = \frac{1}{1 + \theta_{1}T + \theta_{2}T^{2}} \left(1 - b_{1}\mu_{t}^{b_{2}}\right) A_{t}K_{t}^{\gamma}L_{t}^{(1-\gamma)}$$

 $E_t = (1 - \mu_t) \sigma A_t K_t^{\gamma} L_t^{1 - \gamma}$