



JOINT GLOBAL CHANGE RESEARCH INSTITUTE

# Overview of GCAM

December 5, 2019

**The GCAM Team**  
(PNNL-SA-148929)



PNNL is operated by Battelle for the U.S. Department of Energy





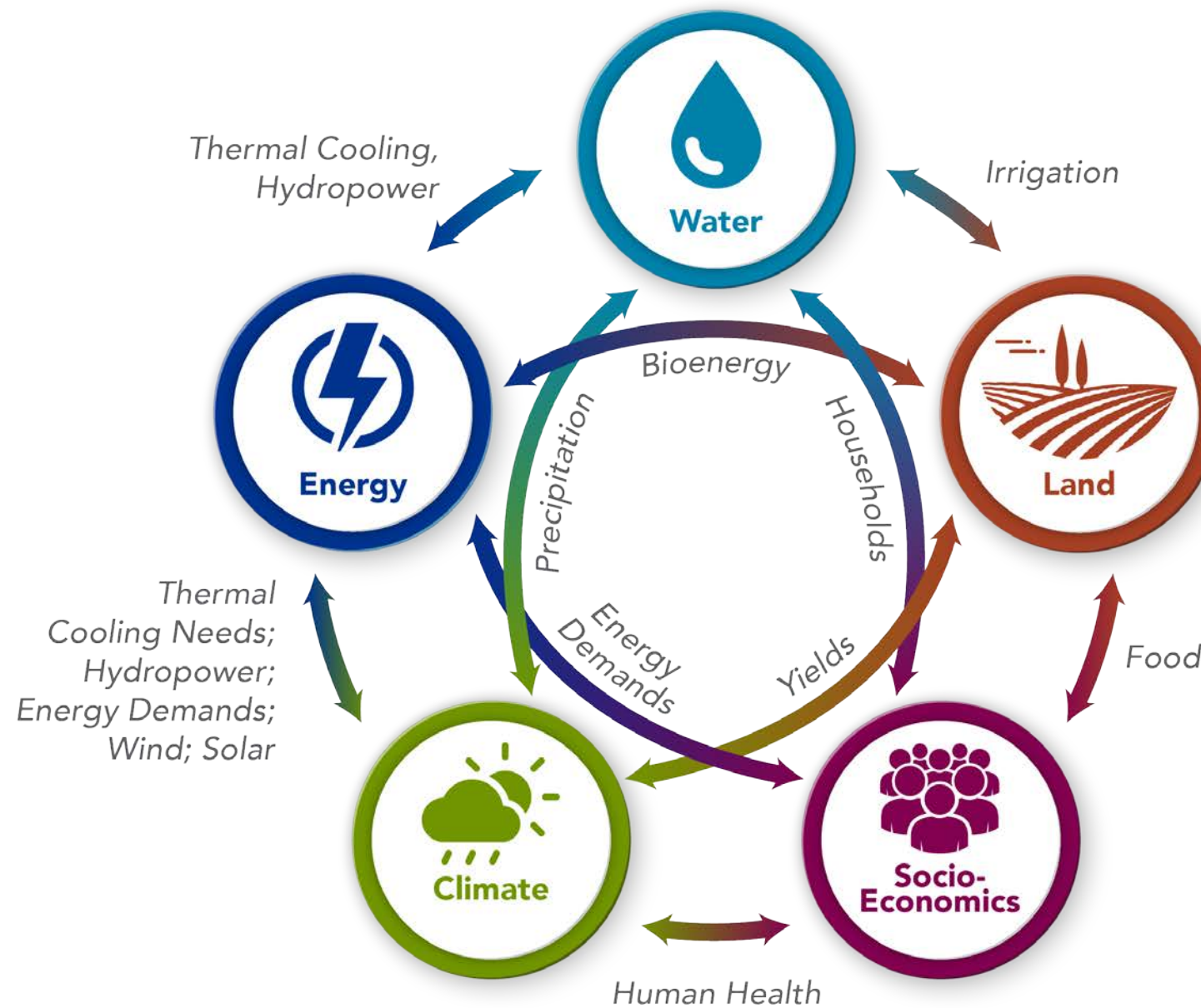
# Outline

- Introduction and Overview
- Energy
- Land
- Water
- Emissions
- Earth System
- Socioeconomics and Trade
- Policies



# Overview of GCAM

# GCAM explores the interactions between multiple systems



# The GCAM ecosystem is a suite of models and tools

## Data Development

GCAM DS

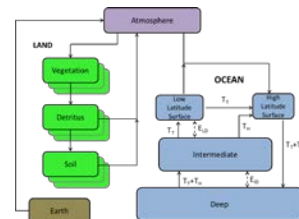


Moirai

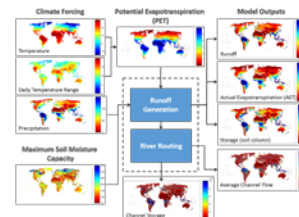


## Single System

Hector



Xanthos



## Dynamic Integration

GCAM-core



32 Energy Economy Regions



235 Water Basins



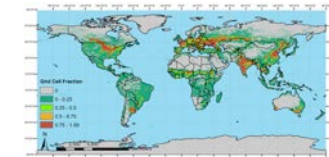
384 Land Regions



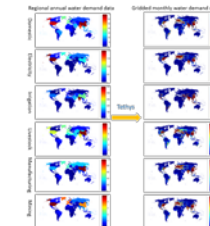
Province-Level Energy Economy Regions

## Disaggregation

Demeter

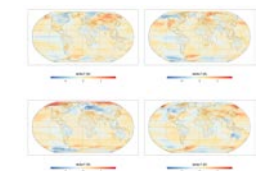


Tethys



## Dynamic Emulators

fIdgen



dnnclim



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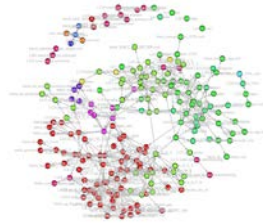
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# The GCAM Release a portion of these

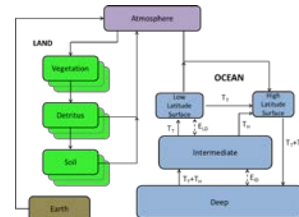
## Data Development

GCAM DS



## Single System

Hector



## Dynamic Integration

GCAM-core



32 Energy Economy Regions



235 Water Basins



403 Land Regions

# An Overview of GCAM

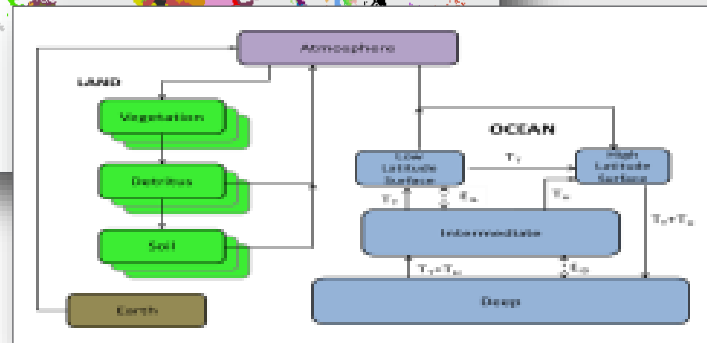
## Global Coverage

32 Energy  
& Economy  
Regions

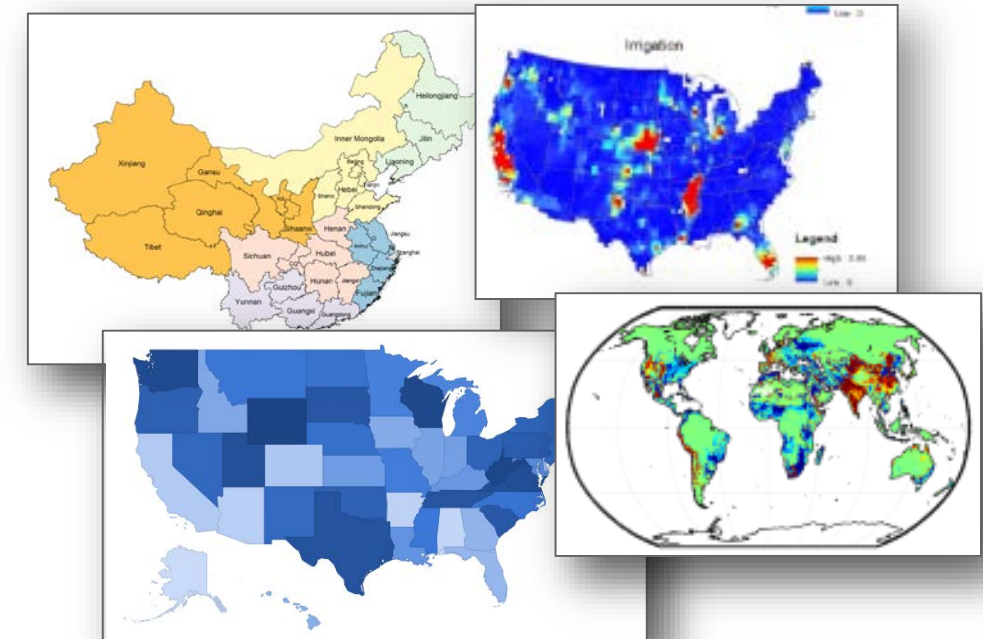
283 Land  
Regions

233  
Water  
Basins

Reduced-  
Form ESM  
(Hector)



Flexible Scale



Market Equilibrium  
Solution



Community Model

<http://jgcri.github.io/gcam-doc/toc.html>

Flexible Time Scale

GCAM Core runs at 5 years; capability to run at one year; ancillary models run at finer scale



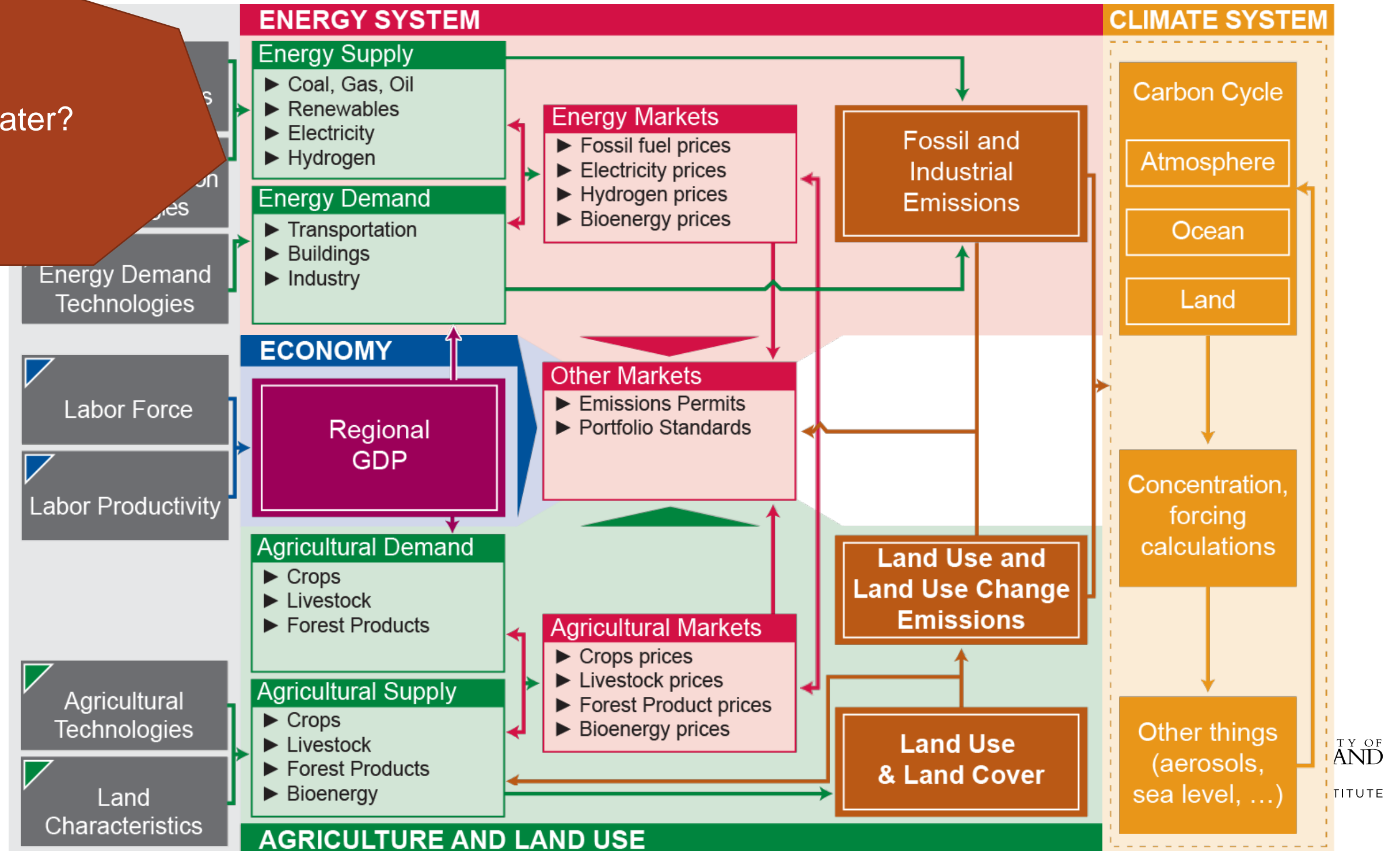
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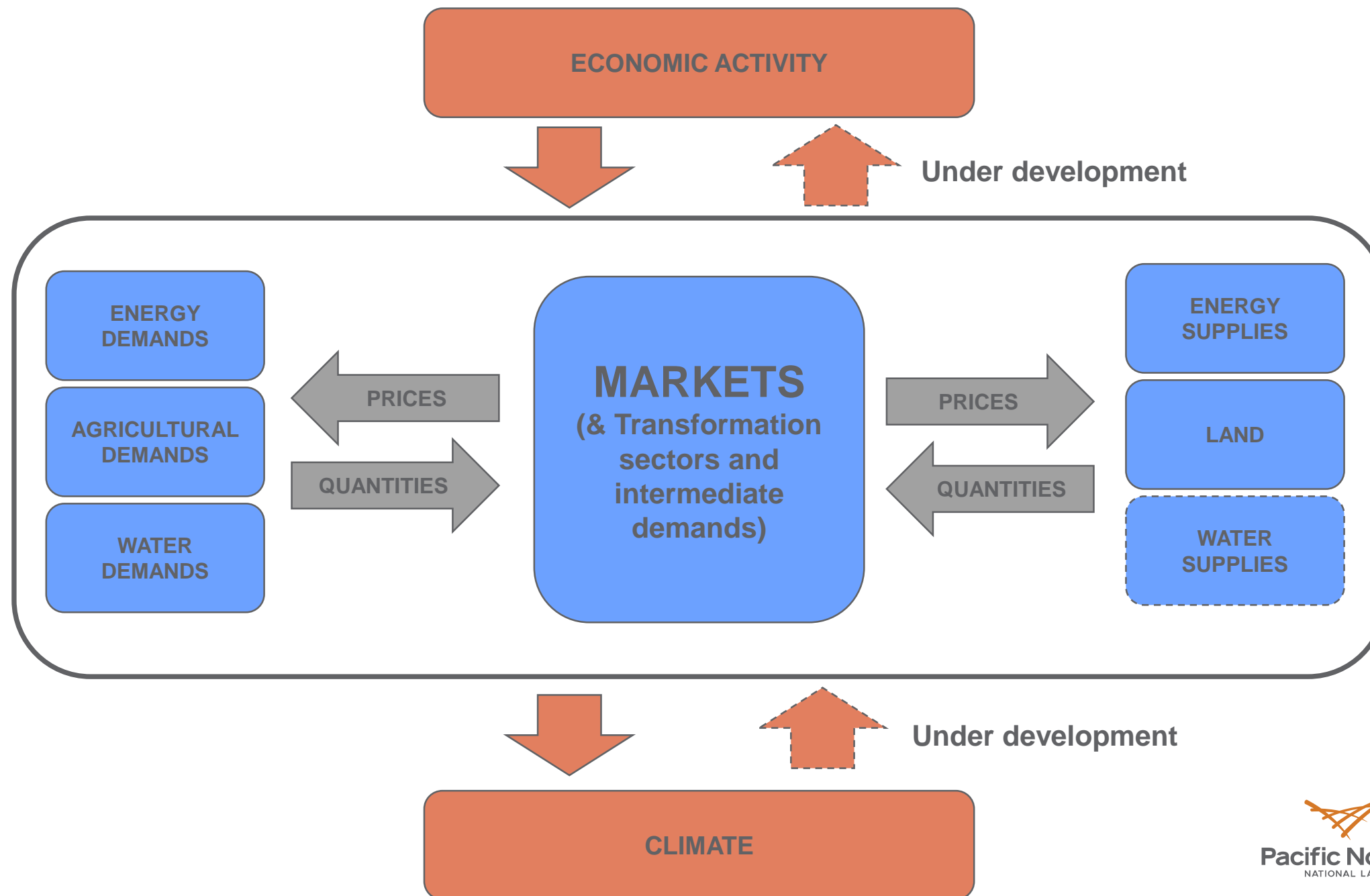
# What's inside the GCAM Core?

Add Water?



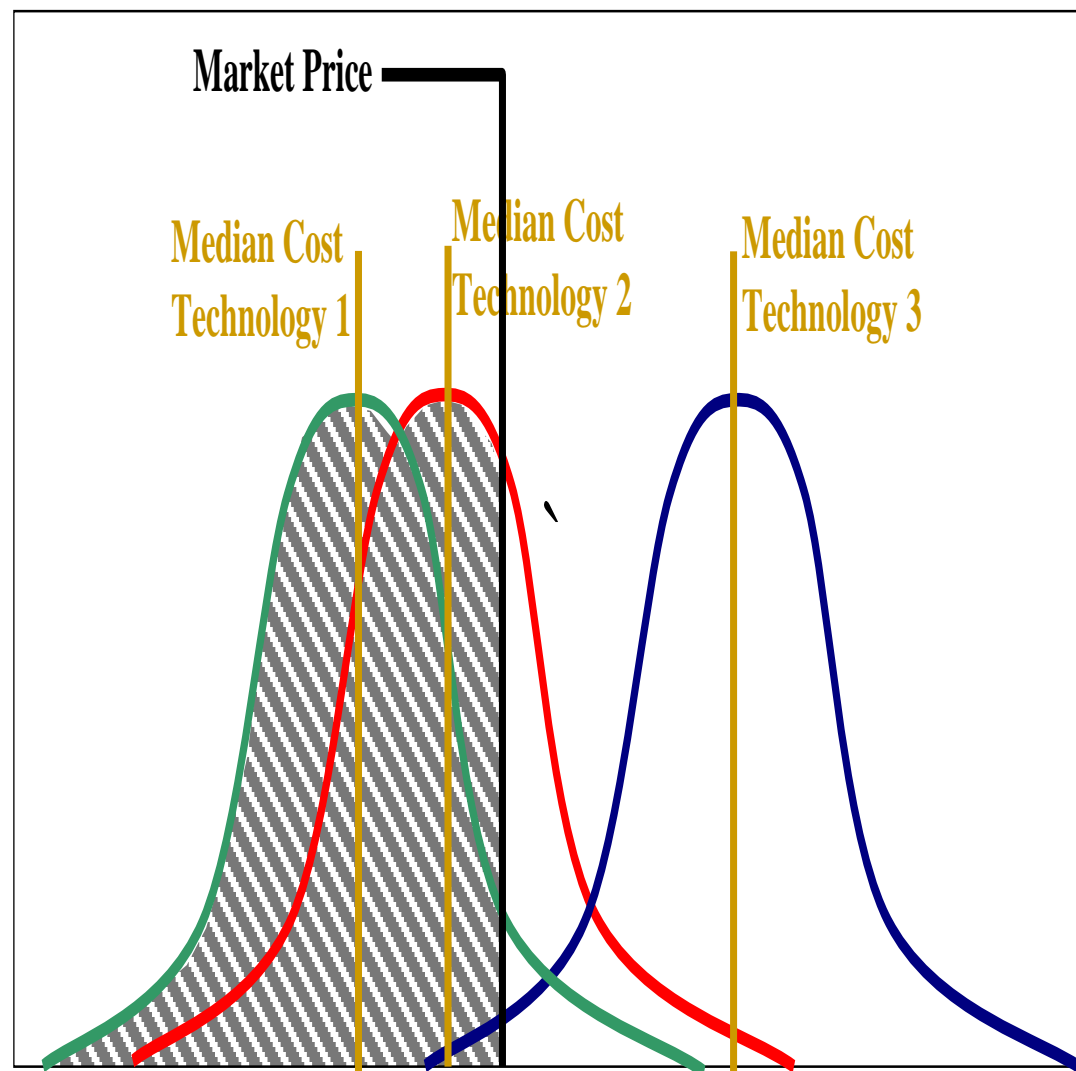


# Market Equilibrium in GCAM



# Logit Choice Mechanism for Decision Making

## *A Probabilistic Approach*

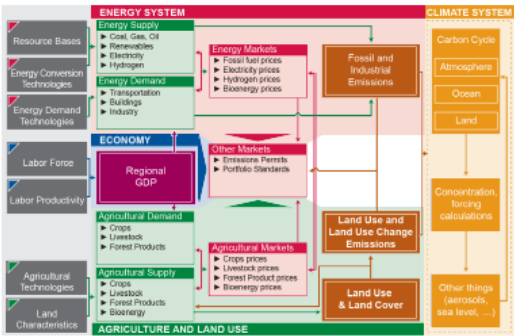
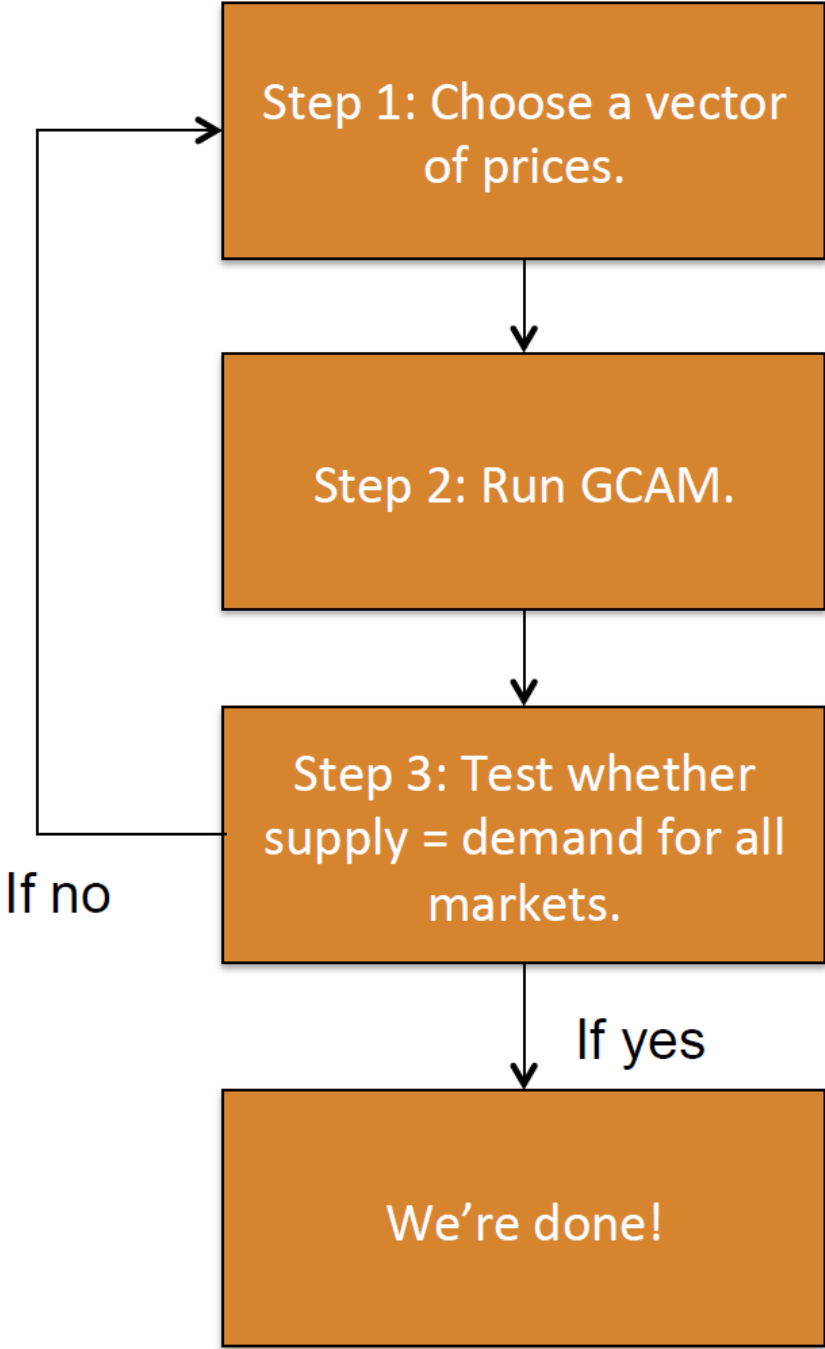


- Calibrated logit approach assumes a distribution of realized costs due to heterogeneous conditions.
- Market share based on probability that a technology has the least cost for an application.
  - Avoids a “winner take all” result.
- Historical calibration influences future competition through the “share-weight” ( $\alpha$ )

$$s_i = \frac{\alpha_i c_i^\sigma}{\sum_j \alpha_j c_j^\sigma}$$



# Solution Approach



$2 + 2 = 5?$

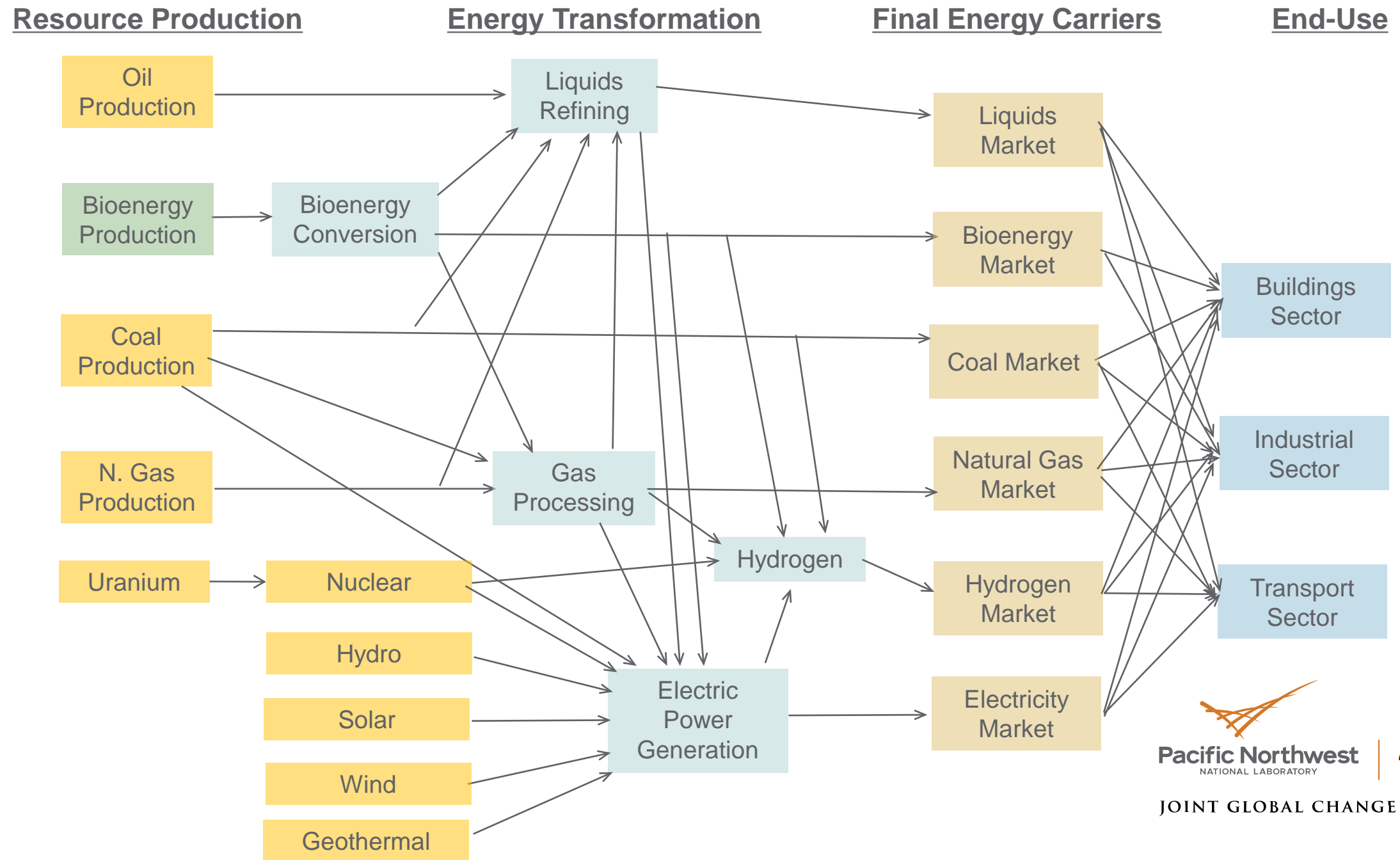




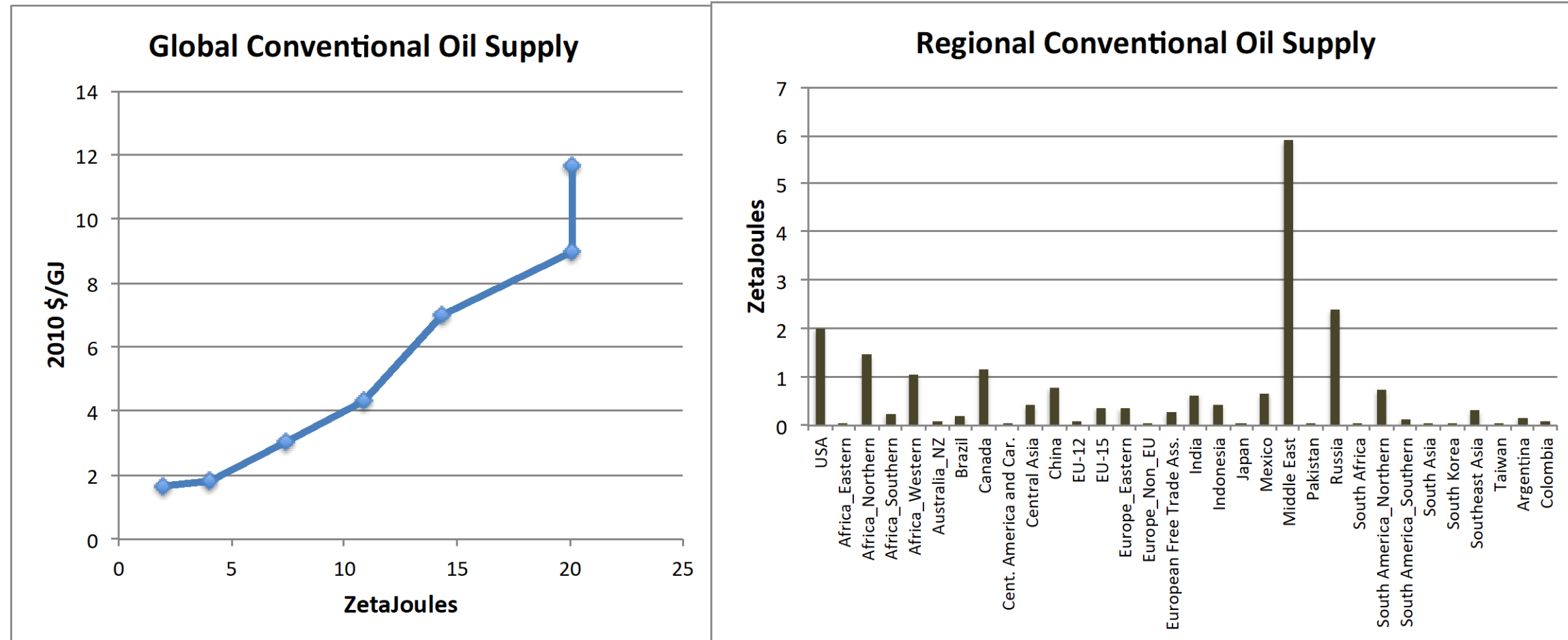
# **The Energy Sector in GCAM**



# The Energy System: Structure



# Depletable Resources: Conventional Oil



- ▶ Conventional oil, unconventional oil, natural gas, and coal resource supply curves derived from Rogner (1997) global assessment.
- ▶ Significant upgrade: Regional Reserve model to distinguish investment in production capacity from long-term resource bases.



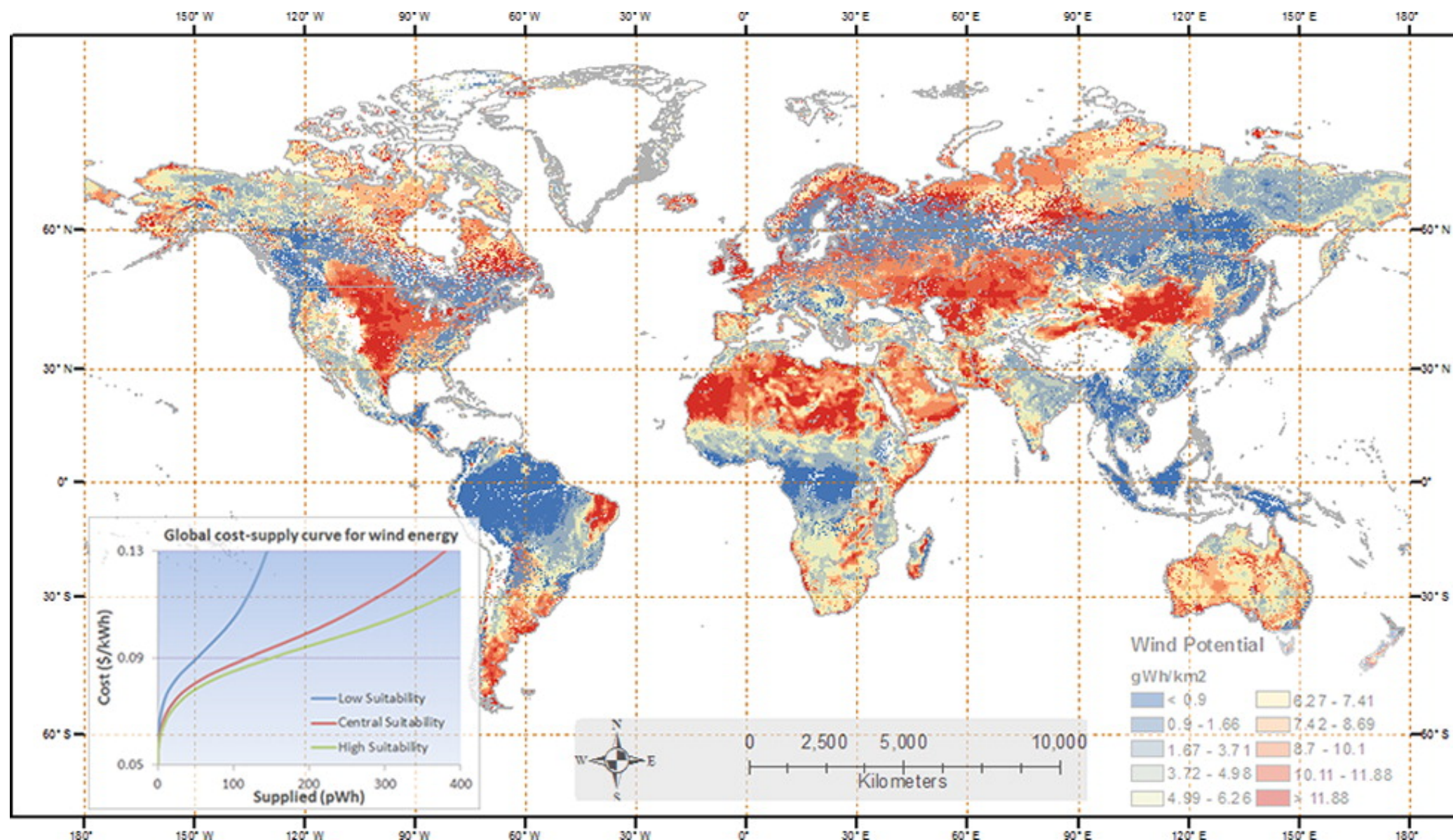
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# Renewable Resources: Wind



$$Q = \maxSubResource * \frac{P^{CurveExponent}}{(MidPrice^{CurveExponent} + P^{CurveExponent})}$$

# Bioenergy

## Purpose Grown Bioenergy:

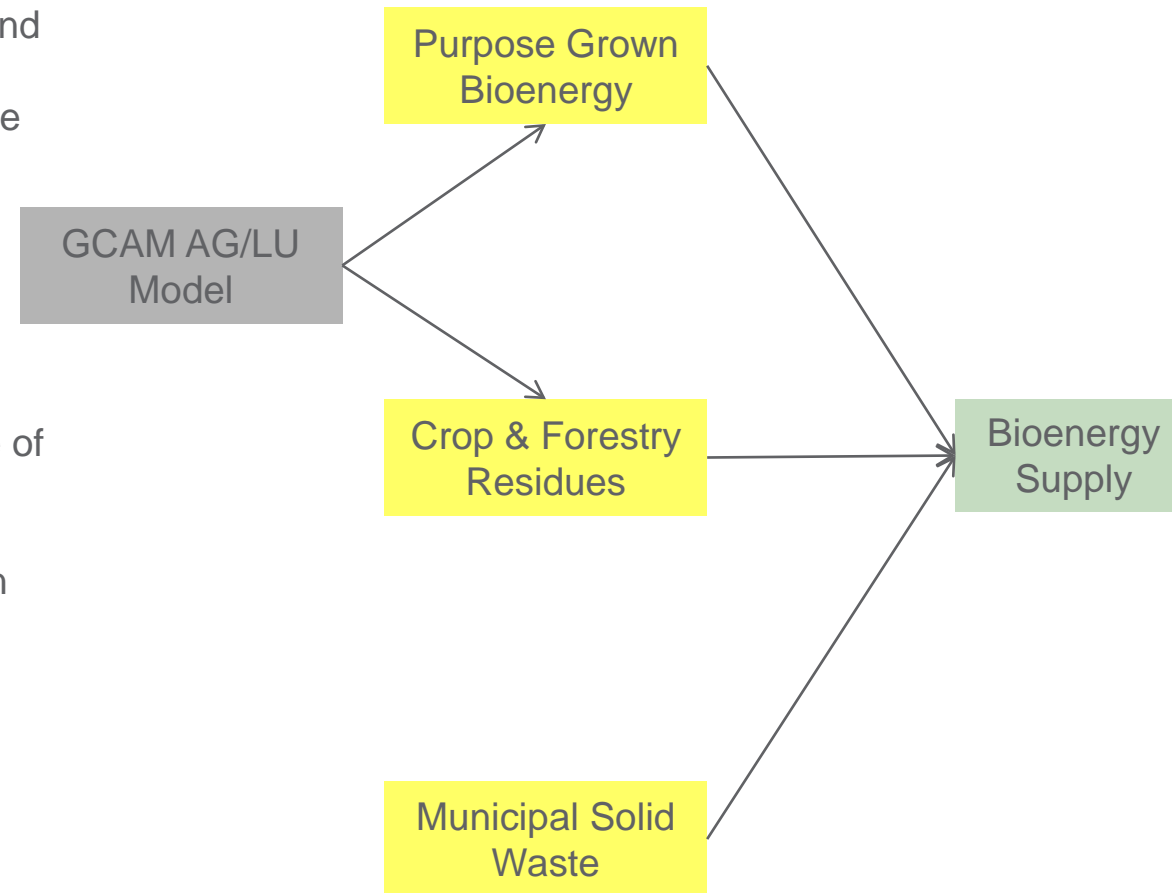
- Production depends on land allocation and regional yield from Ag model
- Land allocation depends on the profit rate of biomass AND all competing land uses
- Includes 1<sup>st</sup> and 2<sup>nd</sup> generation crops

## Crop & Forestry Residues:

- Potential production depends on crop production in ag model
- Fraction harvested depends on the price of bioenergy; higher prices lead to more production
- Some amount of residue must remain on the field for erosion control

## Municipal Solid Waste:

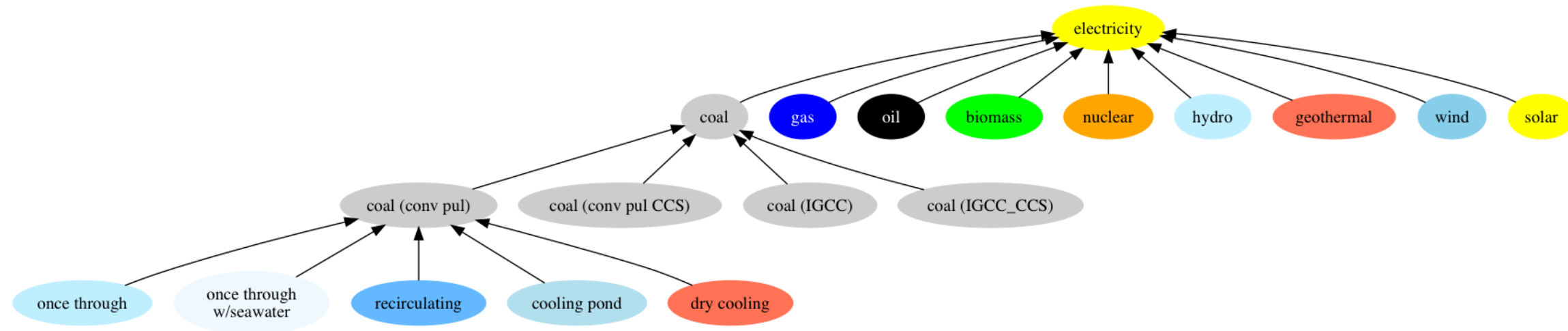
- Potential production depends population and income
- Fraction used for bioenergy depends on the price of bioenergy; higher prices lead to more production



**Note:** We also model traditional bioenergy. However, it is not added to the bioenergy resource pool and is instead consumed directly by the buildings sector. Similarly, we model 1<sup>st</sup> generation bioenergy (corn, sugar, oil crops), but it is converted directly to ethanol or diesel and not added to the bioenergy resource pool.



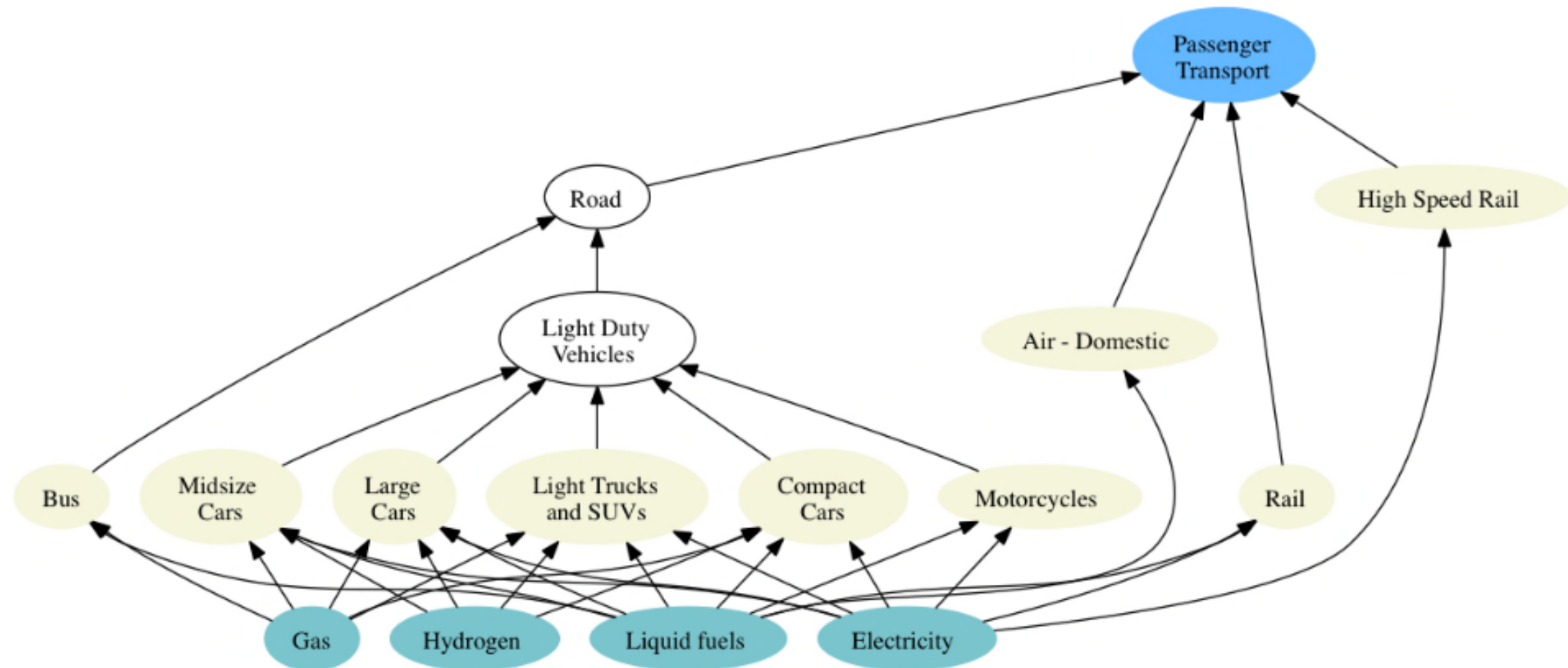
# Energy Transformation: Electric Power Sector



- 3 levels of nesting
  - Fuel (coal, gas, etc)
  - Technology (e.g., efficiency level, CCS)
  - Cooling system (thermo-electric techs only)
- Logit choice competition at each nest based on relative levelized costs of electricity generation
  - Capital + fixed O&M + variable O&M
  - Fuel costs, including emissions penalties
  - Backup-related costs for intermittent technologies

# Final energy demands: Transportation

- Passenger, freight, international aviation, and international shipping
  - Represented in physical units (passenger-km, tonne-km)



# Final energy demands: Transportation (Key Exogenous Inputs)

- By technology
  - Energy intensity
  - Load factor
  - Non-fuel costs
  - Calibration data

$$P_{j,i,r,t} = \frac{P_{f,r,t} * I_{j,i,r,t} + N_{j,i,r,t}}{L_{j,i,r,t}}$$

- By mode (passenger only)
  - Average speed
  - Valuation of time travelling

$$P_{i,r,t} = \sum_{j=1}^N (\alpha_{j,i,r,t} * P_{j,i,r,t}) + \frac{W_{r,t} * V_{i,r,t}}{S_{i,r,t}}$$

- By sector
  - Income elasticity
  - Price elasticity

$$D_{r,t} = D_{r,t-1} \left( \frac{Y_{r,t}}{Y_{r,t-1}} \right)^{\alpha} \left( \frac{P_{r,t}}{P_{r,t-1}} \right)^{\beta} \left( \frac{N_{r,t}}{N_{r,t-1}} \right)$$

*P*: price  
*r*: region  
*i*: mode  
*j*: technology  
*t*: time period  
*I*: fuel intensity  
*N*: non-fuel cost  
*L*: load factor

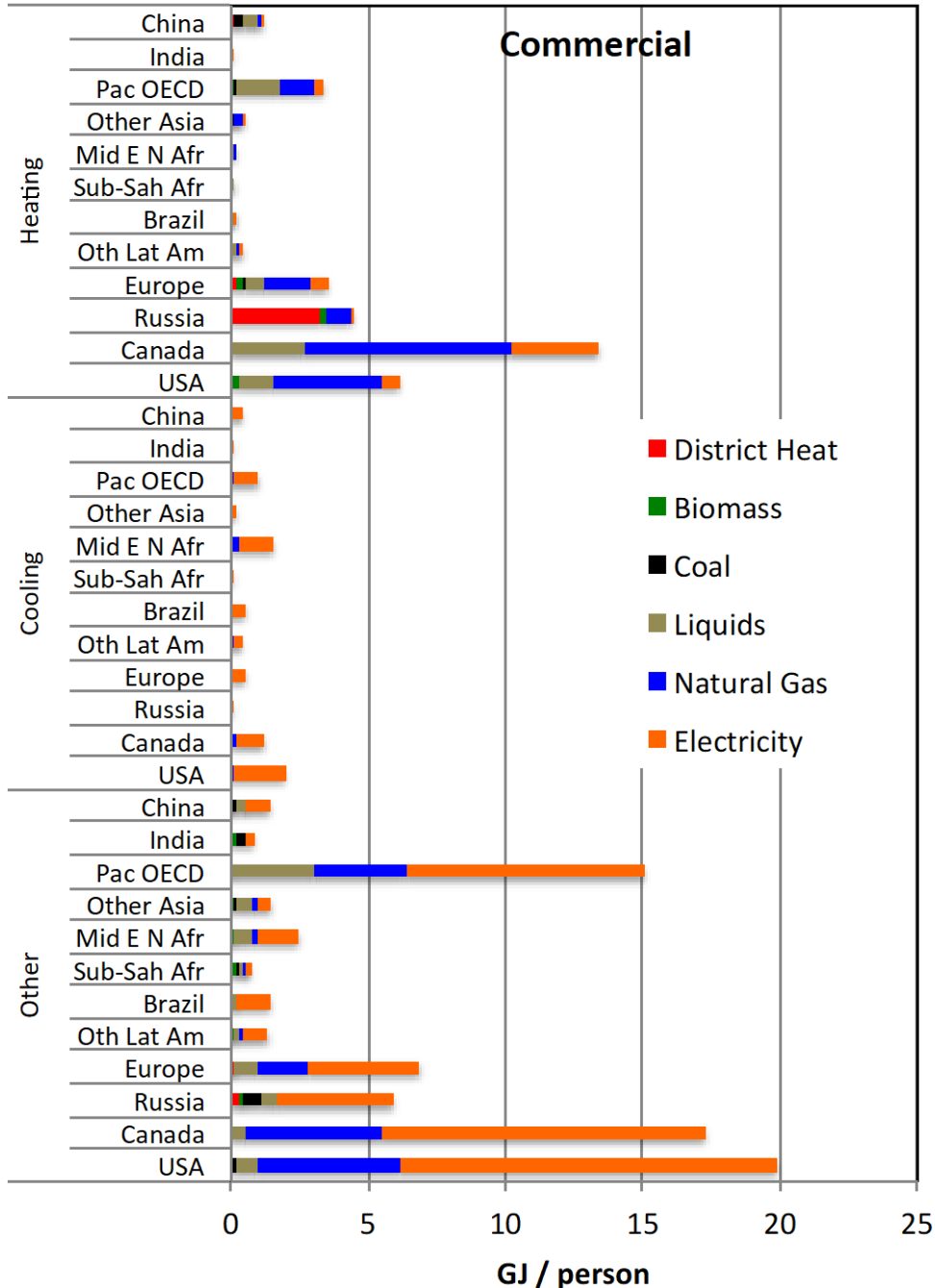
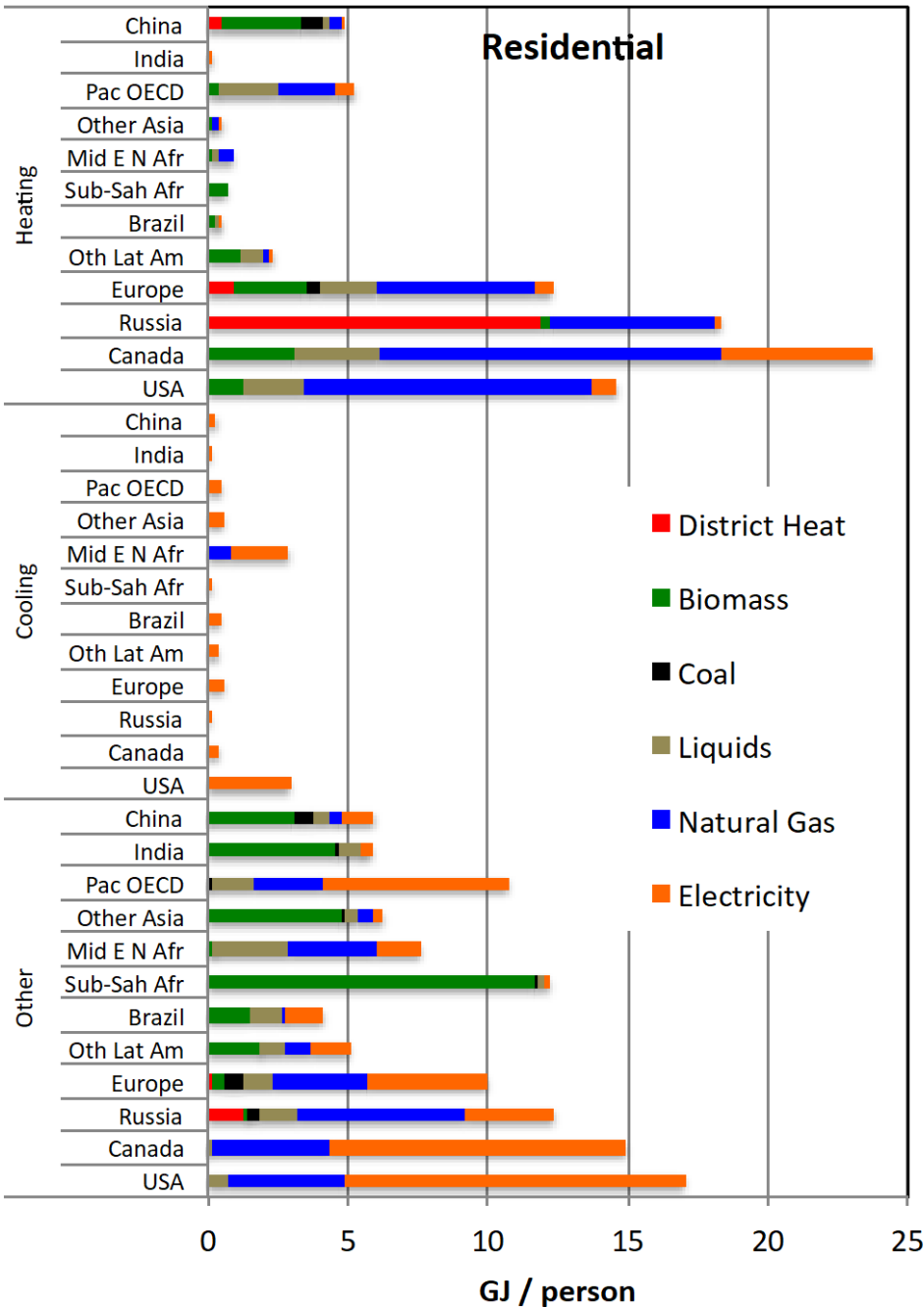
*α*: share of tech *j* within mode *i*  
*W*: wage rate  
*V*: time valuation  
*S*: speed

*D*: demand  
*Y*: per-capita income  
*P*: price  
*N*: population



# Final energy demand: Buildings

Per-capita Residential and Commercial Energy Use in 2010



# Final energy demands: Buildings (Key Exogenous Inputs)

- By technology
  - Efficiency
  - Non-fuel cost
  - Calibration data

- By service

- Satiation level
- Degree days
- Shell conductance

$$d_H = k_H(HDD \cdot \eta \cdot R - IG) \left[ 1 - \exp \left( -\frac{\ln 2}{\mu_H} \frac{i}{P_H} \right) \right]$$

$$d_C = k_C(CDD \cdot \eta \cdot R + IG) \left[ 1 - \exp \left( -\frac{\ln 2}{\mu_C} \frac{i}{P_C} \right) \right]$$

- By sector

- Floorspace satiation level

$$d_t = (s - a) \left[ 1 - \exp \left\{ -\frac{\ln 2}{\mu} I_t \left( \frac{P_t}{P_0} \right)^\epsilon \right\} \right] + a$$

$d$ : demand

H, C: heating, cooling

h: calibration coefficient

HDD, CDD: degree days

$\eta$ : shell conductance

R: floor to surface ratio

IG: internal gain heat

$i$ : per-capita income

$\mu$ : per-capita income at mid-point of satiation fn

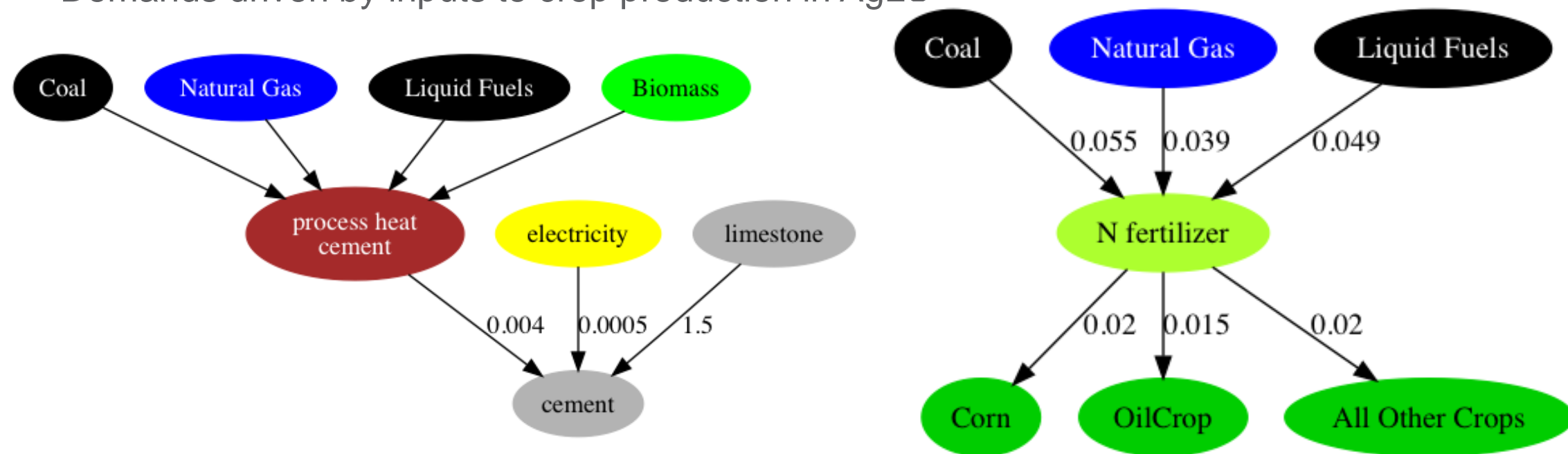
$s$ : satiation level

$a$ : minimum floorspace

$P$ : floorspace price

# Final energy demand: Industry

- Most of the industrial sector is modeled in aggregate form
  - Energy demand ~ GDP
- Cement
  - Process-based representation, in physical units
  - Exogenous final demand quantity
- N fertilizer
  - Process-based representation, in physical units
  - Demands driven by inputs to crop production in AgLU





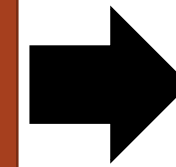
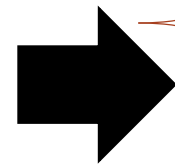


# The Land Sector in GCAM

# Inputs and Outputs

## Inputs

- Harvested area in historic period
- Land cover in historic period
- Production in historic period
- Consumption in historic period
- Cost of production
- Fertilizer application rates
- Water coefficients
- Carbon density, mature age
- Emissions factors
- Income elasticity of demand
- Price elasticity of demand
- Technical change
- Logit parameters
- FAO bilateral trade matrix



## Outputs

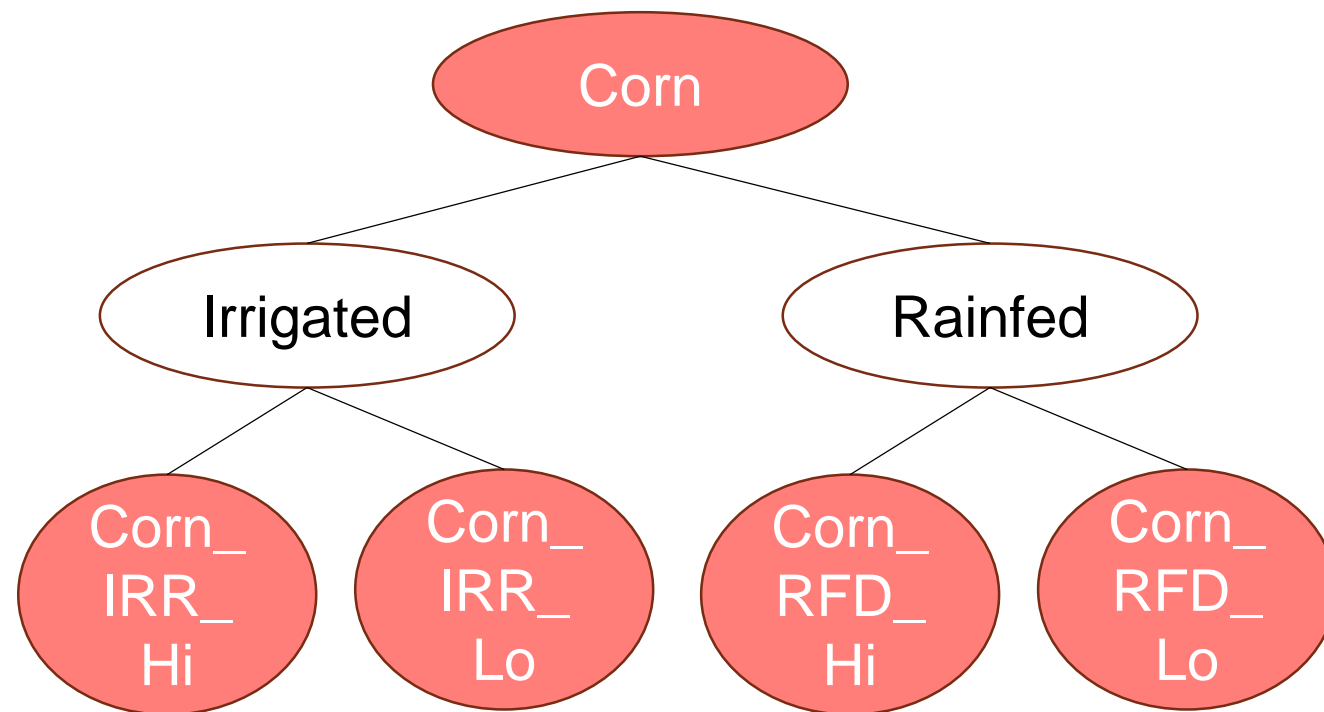
- Production
- Consumption
- Land use, land cover
- Yield
- Price
- Fertilizer use
- Water withdrawals
- Water consumption
- Land use change emissions
- Other land emissions

# Agricultural Demand

- GCAM currently models supply and demand for 13 crops, 6 animal categories, and bioenergy.
- We account for both food and non-food demand, including animal feed.
  - Demand for a given commodity changes over time in response to income, its price, and the price of substitutes.
  - Food, feed, and energy uses of crops are price responsive, but the price elasticity of demand for food is relatively small.
  - Demand is modeled at the 32 region level. Prices differ by each of the 32 regions.

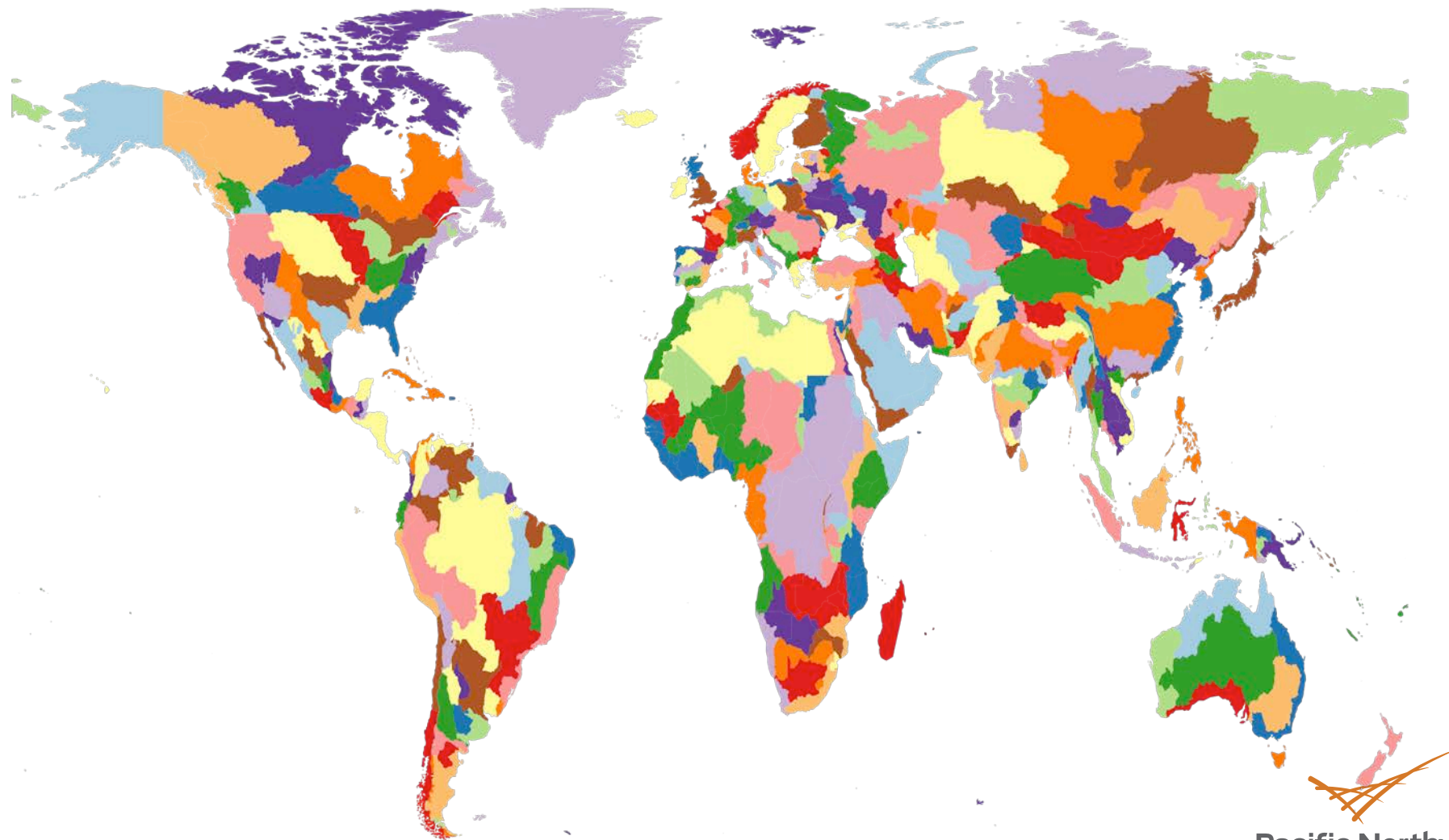


# Agricultural Technologies



- For each crop, we have four different technologies, each with a different yield, cost, and base year allocation.
- Changes in price or cost will alter the shares of each technology, enabling price-induced intensification.

# Land Use Regions



GCAM v5 has 384 land use regions, formed by the intersection of geopolitical regions and water basins.



# Land Use and Land Use Change

- The world is divided into **384** regions
- Farmers allocate land across a variety of uses in order to maximize profit
- There is a distribution of profits for each land type across each of the 384 regions
- The actual share of land allocated to a particular use is the probability in which that land type has the highest profit
- The variation in profit rates is due to variation in the cost of production
  - As the area devoted to a particular land use expands, cost increases
  - Yield is fixed within each region for each crop management practice



# Agricultural Supply

- Yield for each land type/management practice is exogenously calculated.
  - Base year derived from GTAP/FAO production and land area.
  - Yields increase over time based on exogenously specified technical change.
- Land area is endogenously calculated.
  - Each land type/management practice's share of area in its region is the probability its profit is the highest in that region.
- Supply = land \* yield



# Linking the Energy & Agricultural Sectors

- While we can explain the energy and agricultural systems separately, these two systems cannot be separated in practice. Choices made in one sector affect outcomes in another sector.
- This is true both in the real world and in GCAM. You cannot run the different components of the model separately.
- GCAM currently has three means of linking the energy and agriculture systems:
  - Bioenergy: supplied by the agricultural system, demanded by the energy system
  - Fertilizer: supplied by the energy system, demanded by the agricultural system
  - DDGS: supplied by the energy system, demanded by the agricultural system

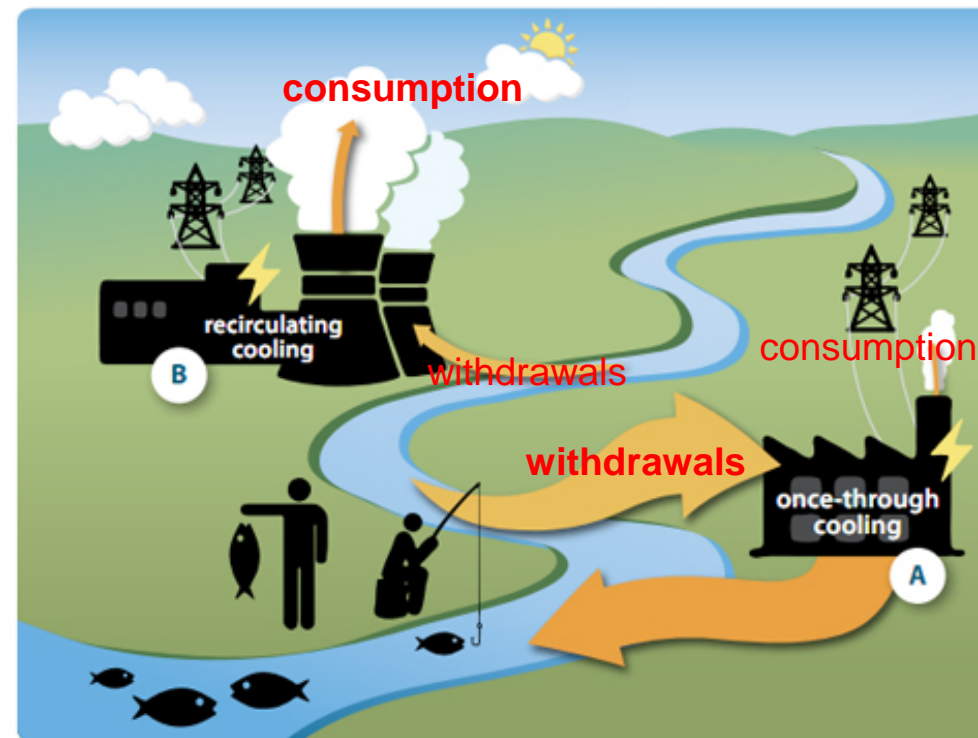


# The Water Sector in GCAM



# Water demands: definitions

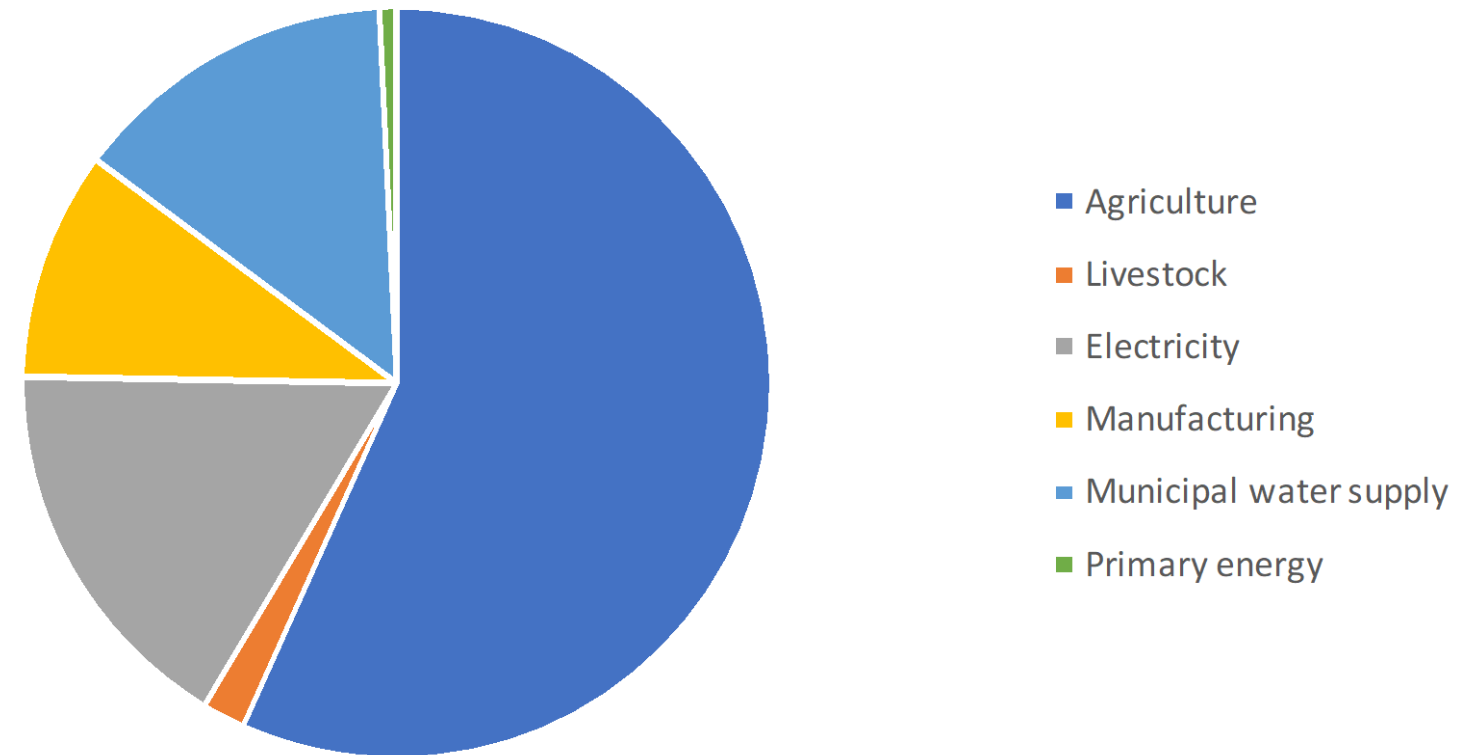
- Water withdrawals: water withdrawn from a surface or groundwater source
- Water consumption: water used in a way that it is removed from its immediate water environment
- Biophysical water consumption: crop evapo-transpiration
- Seawater: water withdrawn from the ocean or brackish estuaries



# Water demands: sectors

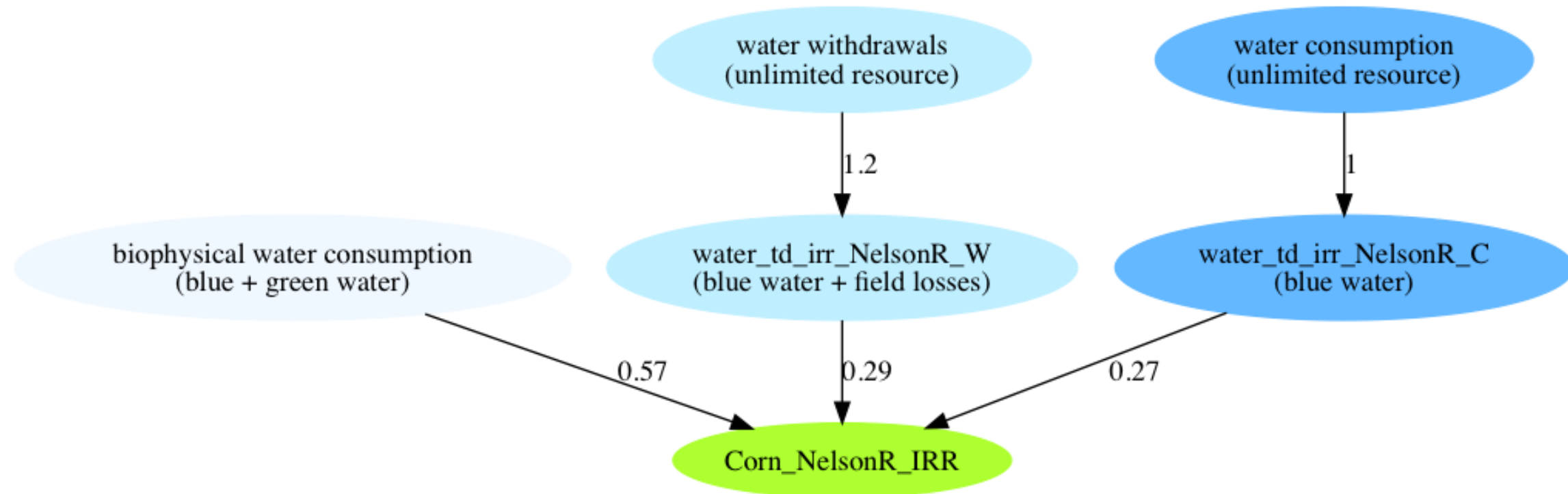
- Agriculture (irrigated crop production)
- Livestock
- Electric power sector (most thermo-electric cooling)
- Industrial manufacturing
- Municipal water (public water supply)
- Primary energy production

2010 Global Allocation of Water Withdrawals



# Agriculture

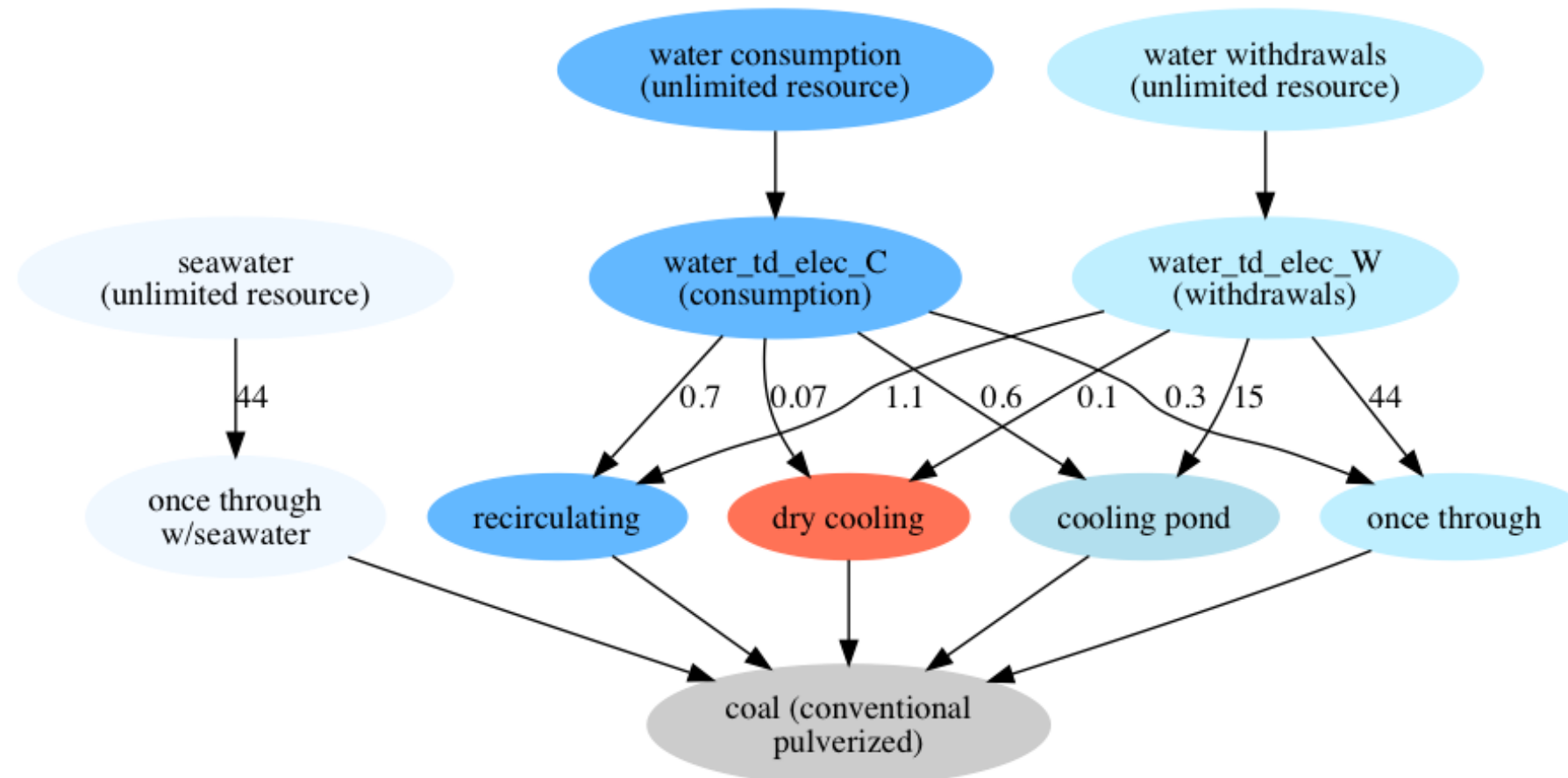
- Water input requirements of crop production
  - Unit:  $\text{m}^3$  of water per kg of crop
  - Specific to each region, basin, and crop type
  - Withdrawals (irrigated techs only): blue water plus field losses
  - Consumption (irrigated tech only): blue water
  - biophysical water consumption (irrigated and rainfed techs): blue + green water





# Electric sector

- Water input requirements of electric power generation
  - Unit: m<sup>3</sup> of water per GJ of electricity generation
  - Thermo-electric generation technologies are allowed up to 5 cooling system types
  - Future choice is generally exogenous as water prices are constant in GCAM 5.1

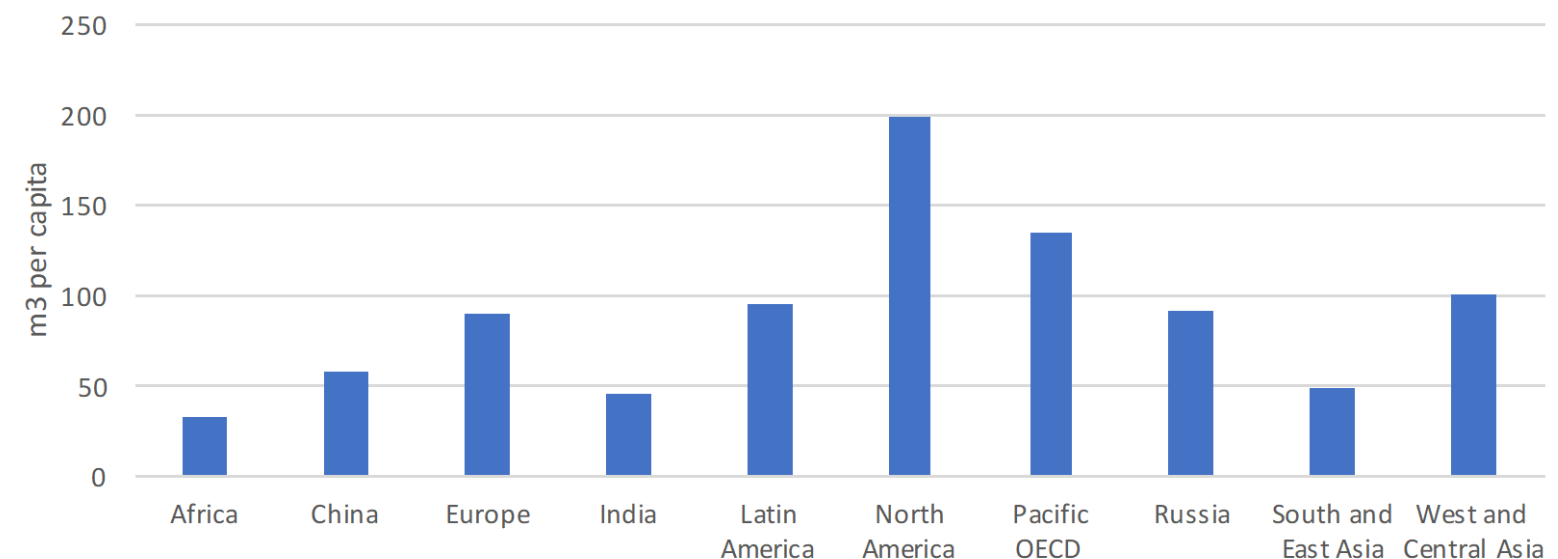


# Municipal water

- Public water supply
  - Historical data from FAO Aquastat
  - Includes industrial sector use of municipal water
  - Excludes residential self-supply (i.e., wells)
- Future per-capita water demands ( $pcW$ ) depend on per-capita GDP ( $pcGDP$ ), water prices ( $P$ ), and technological change ( $Tech$ )

$$pcW_t = pcW_{t-1} * \left( \frac{pcGDP_t}{pcGDP_{t-1}} \right)^{0.37} * \left( \frac{P_t}{P_{t-1}} \right)^{-0.33} * (1 - Tech_t)$$

Per-capita municipal water in 2010





## Remaining sectors

- The remaining sectors' water demands are estimated bottom-up
- Industrial manufacturing
  - All industrial demands of water except for thermo-electric power plant cooling and primary energy production
- Livestock
  - Animal drinking water and operations
- Primary energy production
  - All water use from operations
  - Coal, oil (conventional and unconventional), natural gas, and uranium





# Emissions in GCAM



# GCAM Emissions

GCAM projects emissions of greenhouse gases and air pollutants.

CO<sub>2</sub>, CH<sub>4</sub>, N<sub>2</sub>O, CF<sub>4</sub>, C<sub>2</sub>F<sub>6</sub>, SF<sub>6</sub>, HFC23, HFC32, HFC43-10mee, HFC125, HFC134a, HFC143a, HFC152a, HFC227ea, HFC236fa, HFC245fa, HFC365mfc, SO<sub>2</sub>, BC, OC, CO, VOCs, NO<sub>x</sub>, NH<sub>3</sub>

Future emissions are determined by the evolution of drivers (such as energy consumption, land-use, and population) and the mix of technologies. How this is represented in GCAM varies by emission type.

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**CO<sub>2</sub> Emissions:** GCAM is a process model for CO<sub>2</sub> emissions and reductions

- CO<sub>2</sub> emissions depend on specific technologies, whose use is explicitly determined by the model.
- The GCAM, in effect, produces a Marginal Abatement Curve for CO<sub>2</sub> as a carbon-price is applied within the model.
- Land-Use Change emissions are tracked separately
  - The above-ground (e.g. vegetation) carbon-content of land converted to forest exponentially approaches an exogenously-specified, region-dependent value.
  - Changes in the carbon content of soils due to land-use change also exponentially approach an equilibrium value

# GCAM GHG Emissions: non-CO<sub>2</sub> GHGs

**Non-CO<sub>2</sub> greenhouse gases:** are modeled as

$$Emissions = Em\_factor \bullet Activity\_Level \bullet (1 - MAC(Carbon - Price))$$

Non-CO<sub>2</sub> GHG emission factors only change due to exogenously specified Marginal Abatement Cost (MAC) curves, which encapsulate the technological detail for abatement that is not otherwise explicitly represented in GCAM.

- Below-zero (e.g. “no cost”) MAC mitigation is phased in even in reference cases.
- Under a carbon policy, the emission factor is reduced, as a function of the carbon price, as specified by the MAC curve.



# GCAM Air Pollutant Emissions

**Air Pollutant Emissions** (SO<sub>2</sub>, NO<sub>x</sub>, etc.) are modeled as:

$$Emissions = Em\_factor \bullet Activity\_Level \bullet (1 - Em\_Controls(GDP_{per-capita}))$$

Projections use a global parameterization where emission factors decline as a function of GDP per capita

- This species-specific parameterization captures the general global trend of increasing pollutant controls over time.
- This does not capture regional and technological heterogeneity.
- *Note that the GCAM implementation of the SSP scenarios used a different approach, incorporating region, sector, and fuel specific pollutant emission factor pathways (Calvin et al 2016, Rao et al. 2016).*



# Emissions: Base Year Emissions

- CO2
  - Energy system: we read in global carbon contents for fossil fuels (e.g., coal, gas, oil). These are consistent with values from CDIAC. These carbon contents are used to compute emissions in all years (including the base year).
  - LUC: we read in carbon density, growth parameters, and historical land allocation and compute emissions in all years (including the base year).
- Non-CO2:
  - 2005 emissions calibrated to match the EDGAR\* data set (except BC & OC, where we use RCP inventories). In some cases (e.g., electricity), we supplement EDGAR with EPA to get technology-specific emissions. Additional information for fluorinated gases is from Velders et al.
  - *We are in the process of updating GCAM calibration to be more flexible and calibrate to the newly released CEDS historical emissions dataset, or other datasets as needed.*  
*[globalchange.umd.edu/ceds](http://globalchange.umd.edu/ceds)*

# Emissions: Non-CO2 Drivers

- Energy System
  - Emissions in the energy system can be driven by input (e.g., fuel consumed by a particular technology) or output (e.g., fuel or service produced by a particular technology).
  - Emissions information is technology-specific. As a result, different technologies that produce the same output can have different emissions per unit of activity.
  - For most gases and species, we model drivers of emissions in detail. However, for some F-gases, the driver data (e.g., fire extinguishers) depends only on GDP.
- Agriculture and Land-Use
  - Emissions in the agricultural system can be driven by output (e.g., for crop production) or land area (e.g., for open burning).
  - Emissions information is crop and region specific in GCAM. However, inventory data is region specific, but generally not crop specific (or land-class specific) other than for rice production.

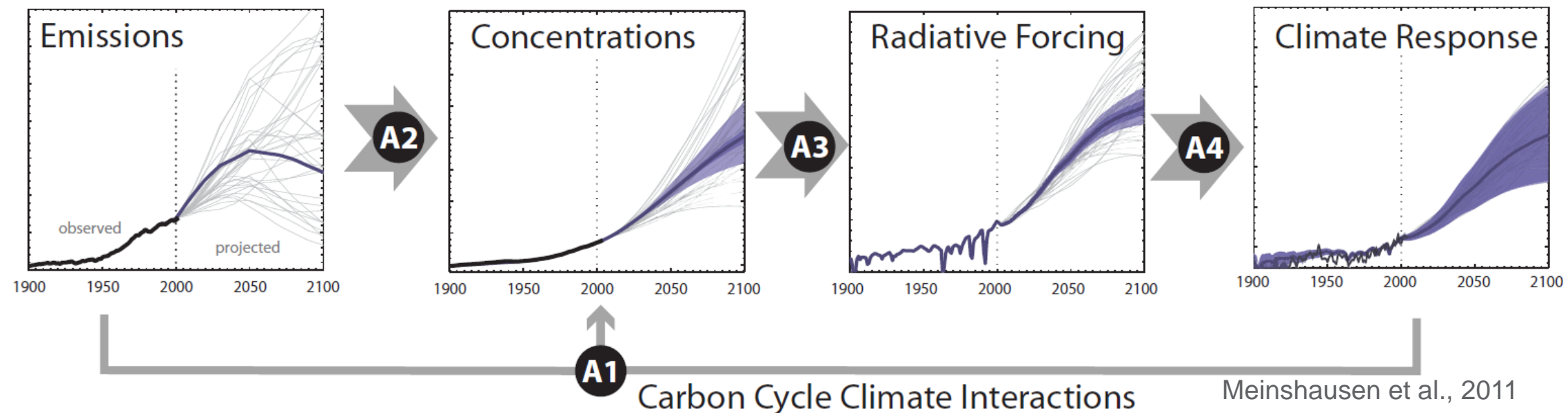


# The Earth System in GCAM



# The Climate System: Approach

- GCAM now uses **Hector** (no longer MAGICC) simple carbon/climate model to compute climate related outputs.
- GCAM passes emissions to the climate model
  - Fossil fuel & Industrial CO<sub>2</sub>, Land-Use Change CO<sub>2</sub>, CH<sub>4</sub>, N<sub>2</sub>O, 26 halocarbons, SO<sub>2</sub>, CO, NO<sub>x</sub>, NMVOCs, BC, OC
- Hector computes atmospheric CO<sub>2</sub> concentrations, radiative forcing (direct and indirect), temperature change, air-land/air-sea fluxes, ocean heat flux...



# Hector: Science

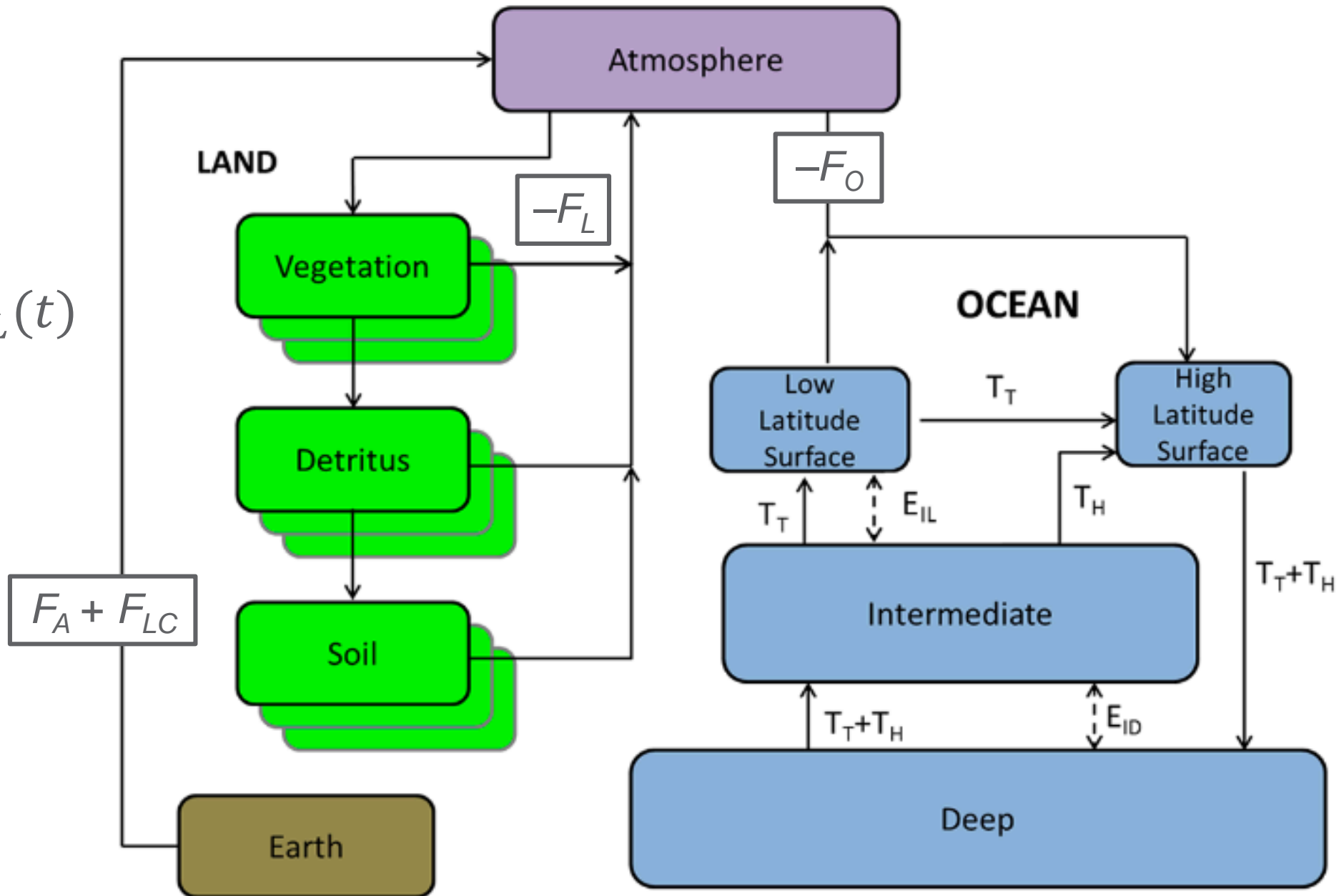
$$\Delta[CO_2]_t = F_A(t) + F_{LC}(t) - F_O(t) - F_L(t)$$

$F_A$  : Fossil fuel emissions

$F_{LC}$  : Land-use change emissions

$F_O$  : Atm.  $\rightarrow$  ocean flux

$F_L$  : Atm.  $\rightarrow$  land flux



# Terrestrial C cycle

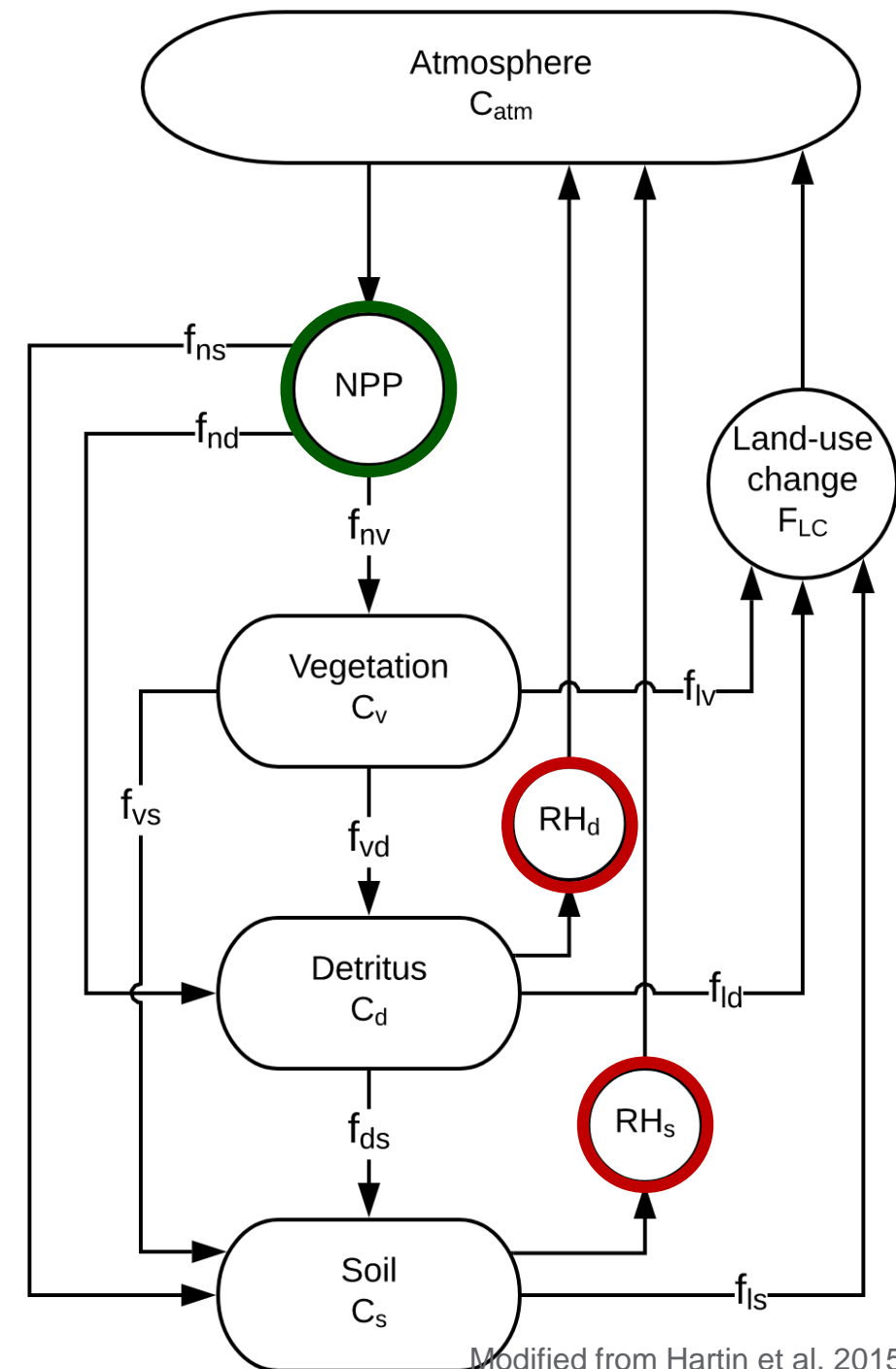
**NPP: Net primary productivity** (photosynthesis – plant respiration) increases with atmospheric  $\text{CO}_2$  concentration according to parameter  $\beta$ :

$$NPP(t) \sim f(NPP(0), [CO_2]_t \mid \beta)$$

**RH: Heterotrophic respiration** (respiration by microbes) from soil and detritus increases with atmospheric temperature according to parameter  $Q_{10}$ :

$$RH_x \sim f(T_{atm} \mid Q_{10,x})$$

At each time step, **C gain from NPP** and **C loss from decomposition and land-use change (LC)** are distributed between vegetation, soil, and detritus pools according to fractionation parameters  $f_{ab}$



Modified from Hartin et al. 2015 GMD

# Ocean C cycle

**Surface boxes** exchange  $\text{CO}_2$  with the atmosphere based on the  $\text{CO}_2$  gas transfer velocity ( $k$ ), the solubility of  $\text{CO}_2$  ( $\alpha$ ), and the ocean-atmosphere gradient of  $\text{CO}_2$  partial pressure.

$$F_i(t) = k\alpha(p\text{CO}_{2,i} - p\text{CO}_{2,\text{atm}})$$

An **ocean chemistry submodel** relates total C in each ocean box to the dissolved organic C (DIC) used for these calculations.

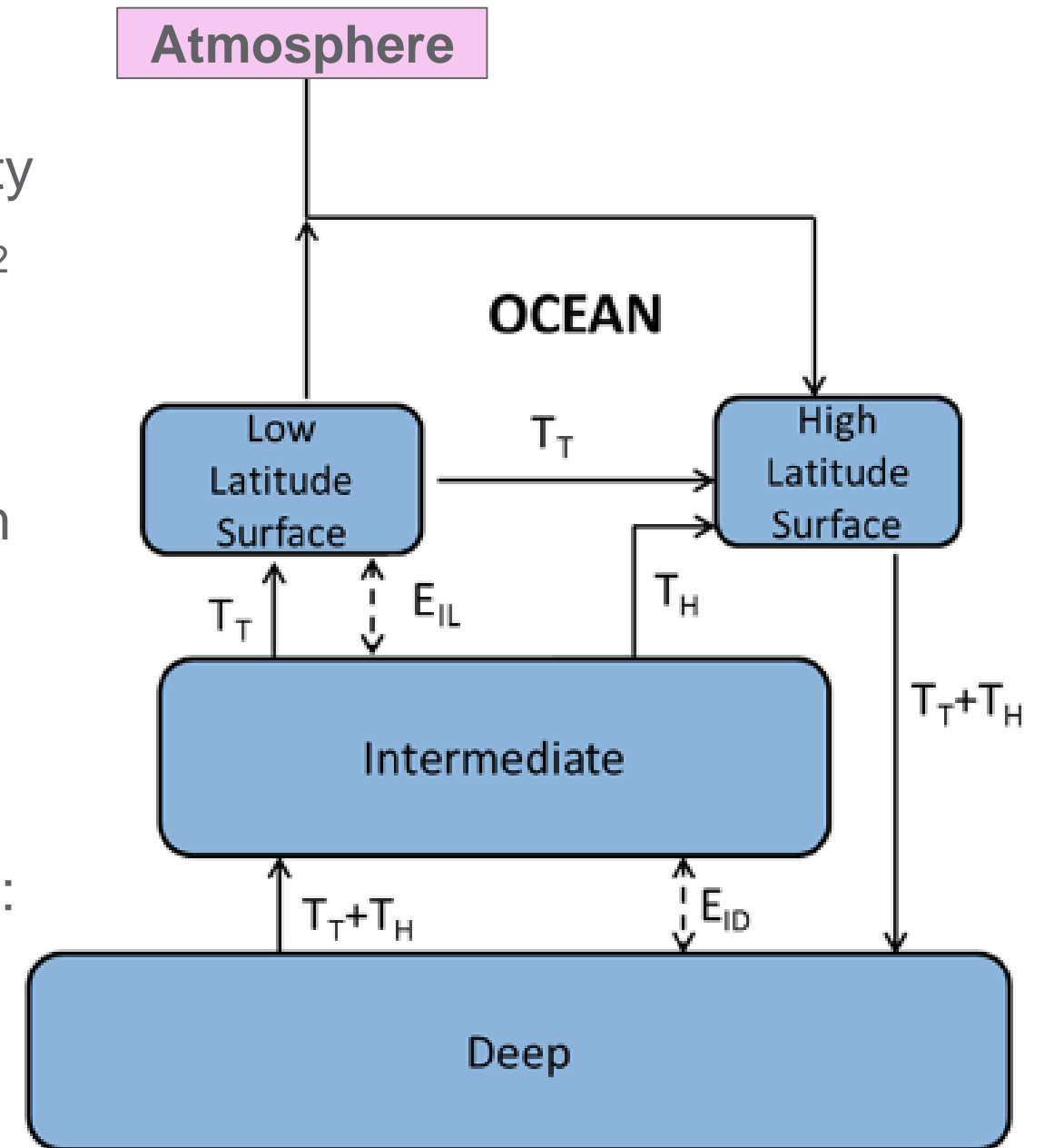
A **simplified model of global ocean circulation** moves water (and dissolved C therein) between ocean boxes, according to the following empirical parameters:

$T_T$ : Thermohaline circulation

$T_H$ : High-latitude circulation

$E_{ID}$ : Water mass exchange – intermediate to deep

$E_{LI}$ : Water mass exchange – low-latitude to intermediate

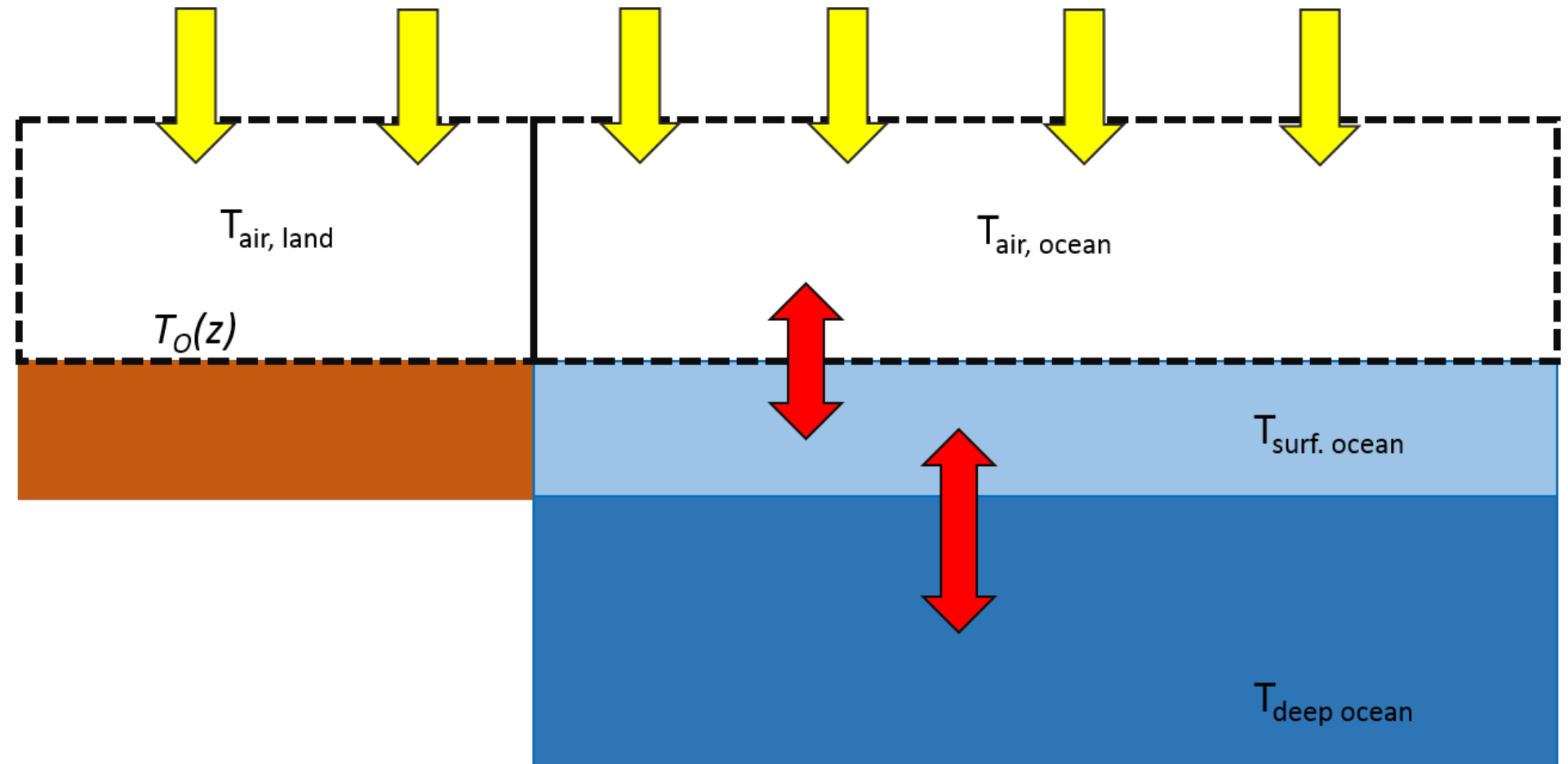




# Incorporated a diffusive ocean energy balance model into Hector - DOECLIM

## Hector outputs:

Air temperature over land  
Air temperature over ocean  
Ocean surface temperature  
Global Mean Temperature  
Ocean heat uptake/flux



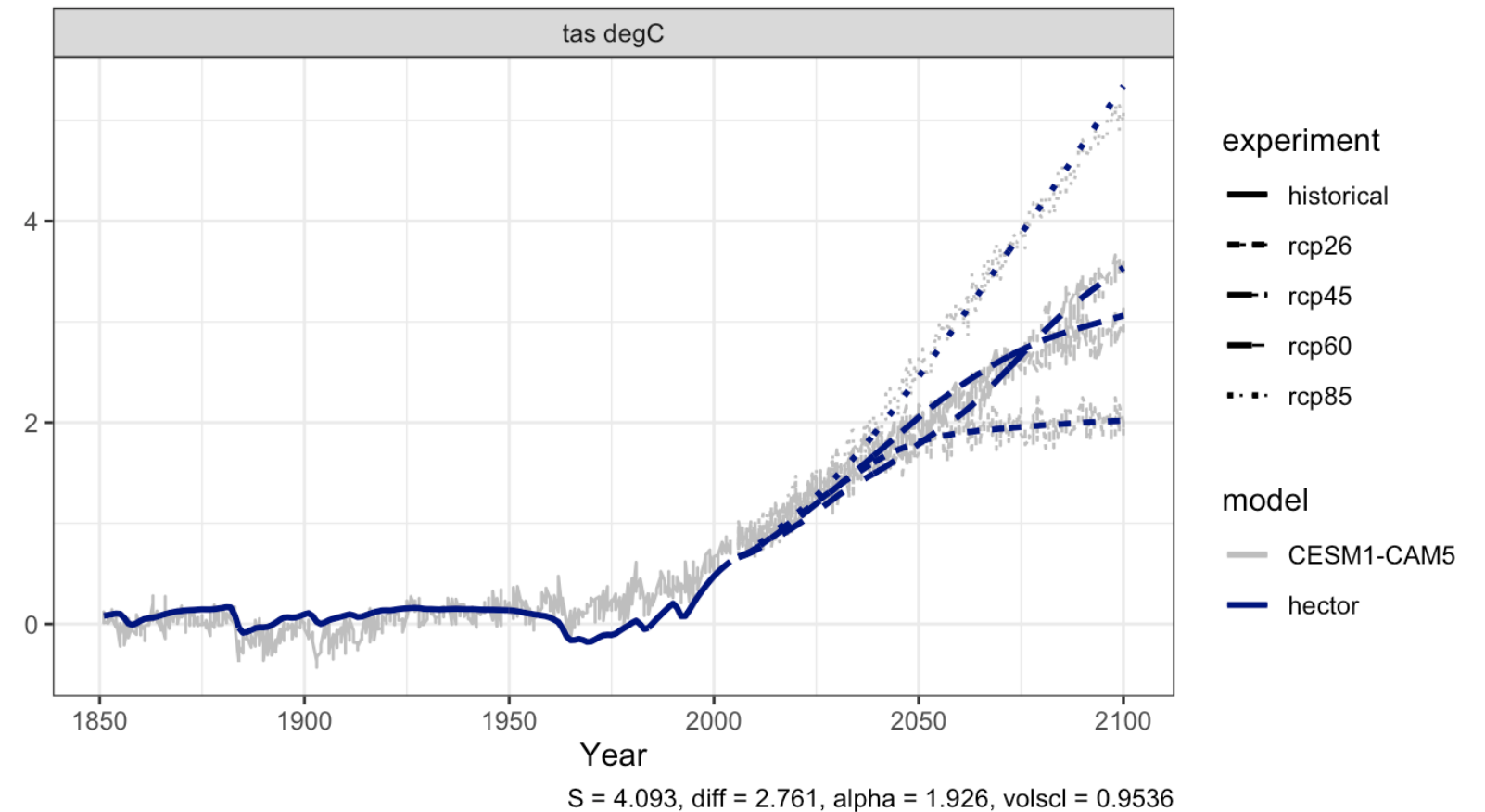
# Hector emulates the CMIP5 median and individual models\*

\* CMIP6 near future

## Hector Outputs:

Atmospheric CO<sub>2</sub>  
Concentrations of GHGs  
Radiative Forcing – total and individual  
Global Mean Temperature  
Ocean Surface Temperature  
Net Primary Production  
Heterotrophic Respiration  
Ocean Acidification

CESM1-CAM5 vs. Hector Output



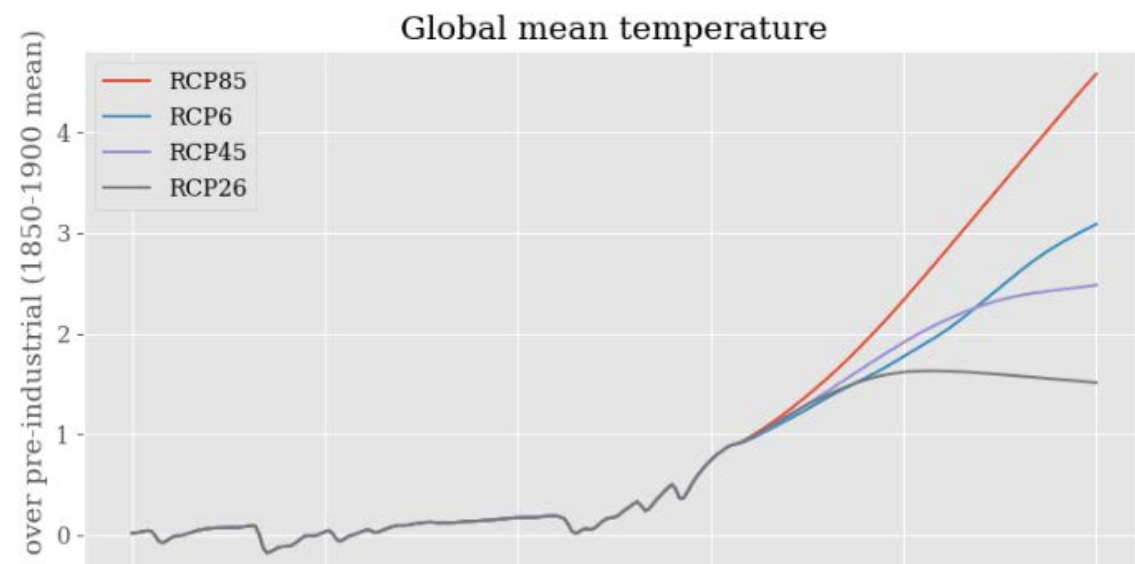
Dorheim et al., submitted JAMES

# User Friendly Applications - pyhector and rhector

```
import pyhector
from pyhector import rcp26, rcp45, rcp60, rcp85

import matplotlib.pyplot as plt

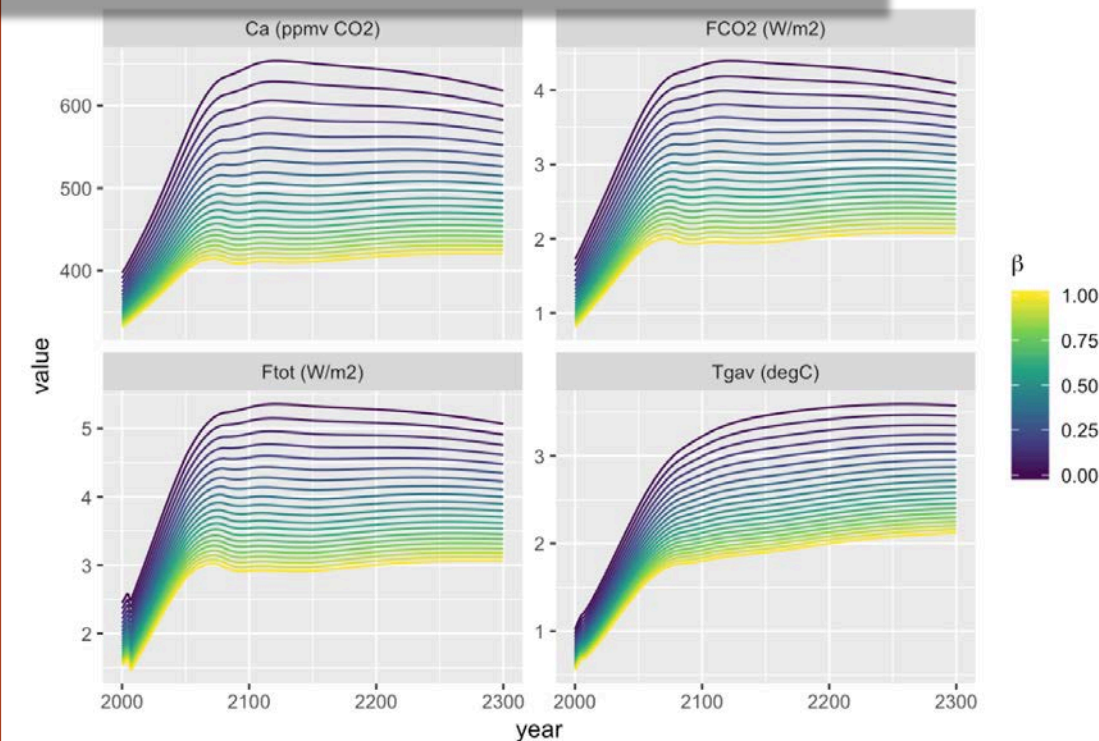
for rcp in [rcp26, rcp45, rcp60, rcp85]:
    output = pyhector.run(rcp, {"core": {"endDate": 2100}})
    temp = output["temperature.Tgav"]
    # Adjust to 1850 - 1900 reference period
    temp = temp.loc[1850:] - temp.loc[1850:1900].mean()
    temp.plot(label=rcp.name.split("_")[0])
plt.title("Global mean temperature")
plt.ylabel("°C over pre-industrial (1850-1900 mean)")
plt.legend(loc="best")
plt.show()
```



<https://github.com/openclimatedata/pyhector>

```
run_with_beta <- function(value) {
  setvar(core, NA, BETA(), value, "(unitless)")
  reset(core)
  run(core)
  result <- fetchvars(core, 2000:2300)
  result[["beta"]] <- value
  result[["variable_unit"]] <- with(result, {
    sprintf("%s (%s)", variable, units)
  })
  result
}
```

```
mapped <- Map(run_with_beta, seq(0, 1, 0.05))
sensitivity_beta <- Reduce(rbind, mapped)
```





# **Socioeconomics and Trade in GCAM**



# Representing Economic Activity

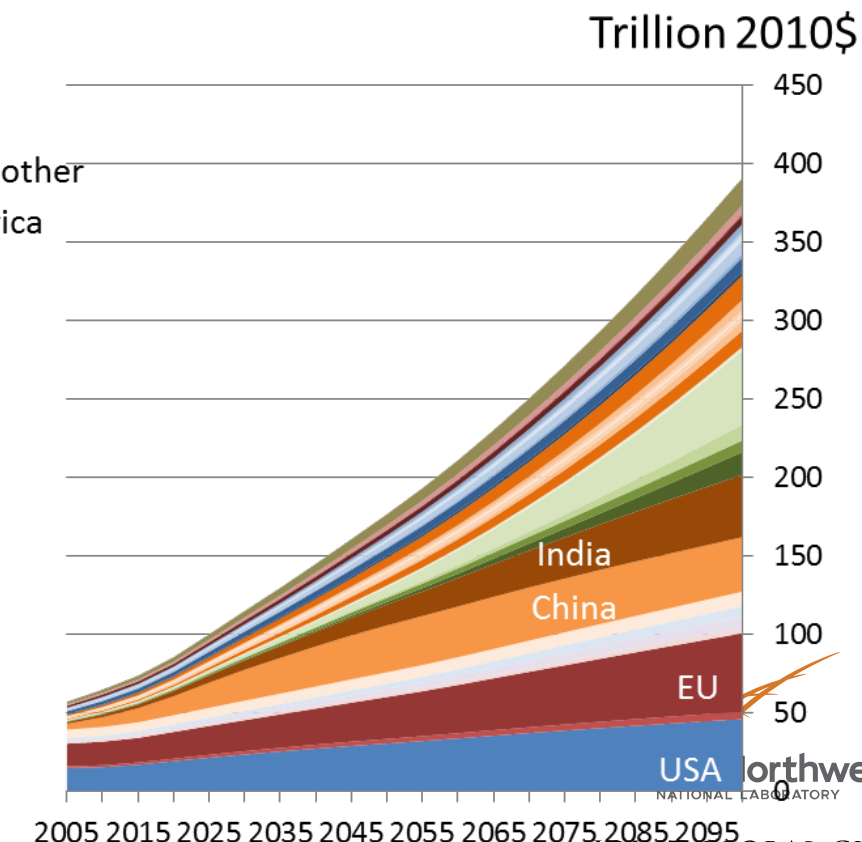
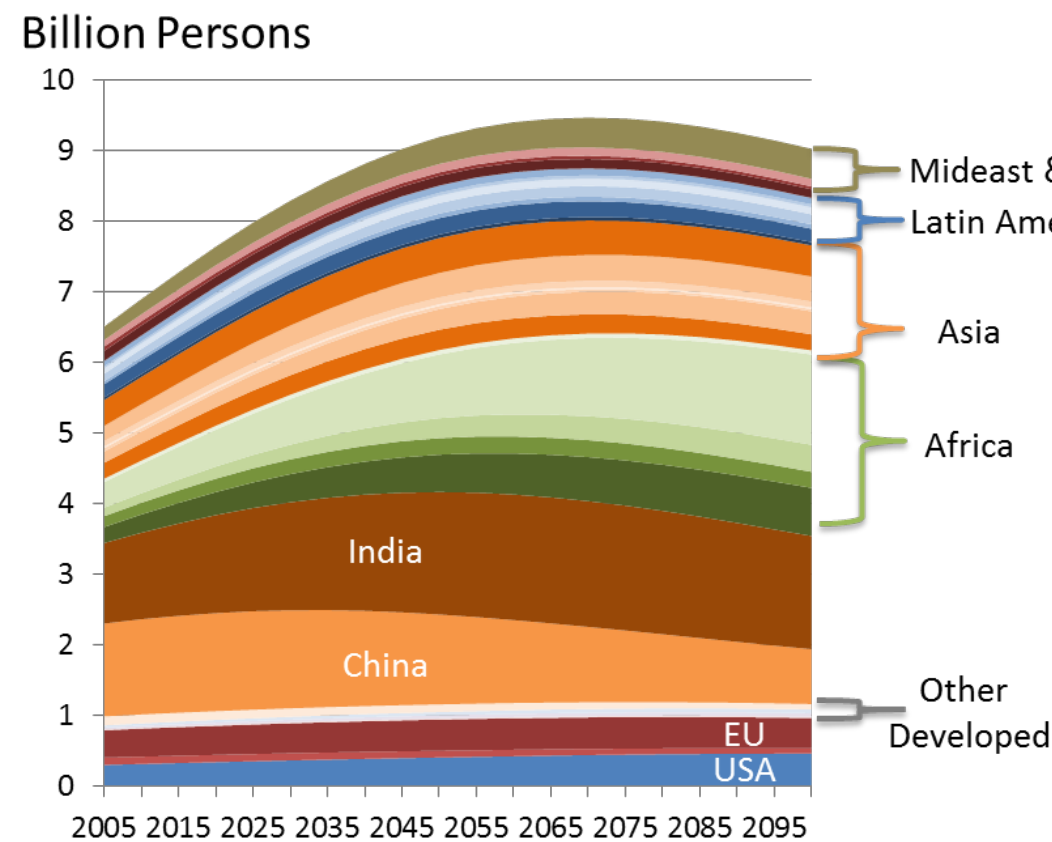
- GDP is a driver in GCAM, based on population and labor productivity, both of which can grow over time

$$GDP_{r,t+1} = POP_{r,t+1} * (GDP/cap_{r,t}) * (1 + g_{r,t})^{t\_step}$$

- Where:
  - r = region, t = model period, t\_step = years in model period
  - POP = population
  - g = annual rate of growth of per capita income

# Socioeconomic projections for core version

- GDP and population are based on the Middle of the Road scenario from Shared Socioeconomic Pathways,
- These projections are modified in the near-term to reflect near-term trends
- The core version of the model not explicitly include other SSP assumptions beyond population and GDP, however the full GCAM SSPs are available as part of the release



# Trade In GCAM

- Heckscher-Ohlin
  - Commodities such as coal, gas, and oil are each traded in a single global market.
  - Each region will see the same global price and independently decide how much each will supply and demand of each commodity given that price.
  - A region's net trade position is dynamic depending on economics, technical change, demand, growth, resources, etc.
- Armington Style Trade for Primary Ag
  - Armington-style distinction between domestic and imported goods but not with fixed bilateral for Primary Ag goods such as Corn, Rice, purpose grown bio-energy etc.
  - Imports are from a single Global pool
- Fixed Inter-Region Trade
  - Some commodities such as meat and dairy and trade volumes are simply held fixed at their historical value for the rest of the simulation.
- No Trade of Secondary energy goods
  - Note that secondary energy products such as Electricity or Refined Liquids are assumed to not be traded at all between the 32 GCAM regions.



# Implementing Policies in GCAM





# Emissions-Related Policies

- Carbon or GHG prices:
  - Users can specify the price of carbon or GHGs directly
  - Emissions will vary depending on other scenario drivers
- Emissions constraints:
  - Users can specify the total amount of emissions (CO<sub>2</sub> or GHG)
  - Model will calculate the price of carbon needed to reach the constraint
- Climate constraints:
  - Users can specify a climate variable (e.g., concentration or radiative forcing) target for a particular year
  - Users determine whether that target can be exceeded prior to the target year
  - Model will adjust carbon prices in order to find the least cost path to reaching the target
  - (This type of policy increases model run time significantly)

# Energy-Related Policies

- We can impose constraints (lower & upper bounds) on energy consumption.
  - The model will solve for the tax (upper bound) or subsidy (lower bound) required to reach the given constraint.
  - Within an individual sector, these constraints can be share constraints (e.g., fraction of electricity that comes from solar power).
    - ✓ This allows us to model renewable portfolio standards and biofuels standards.
  - Across sectors, these must be quantity constraints.

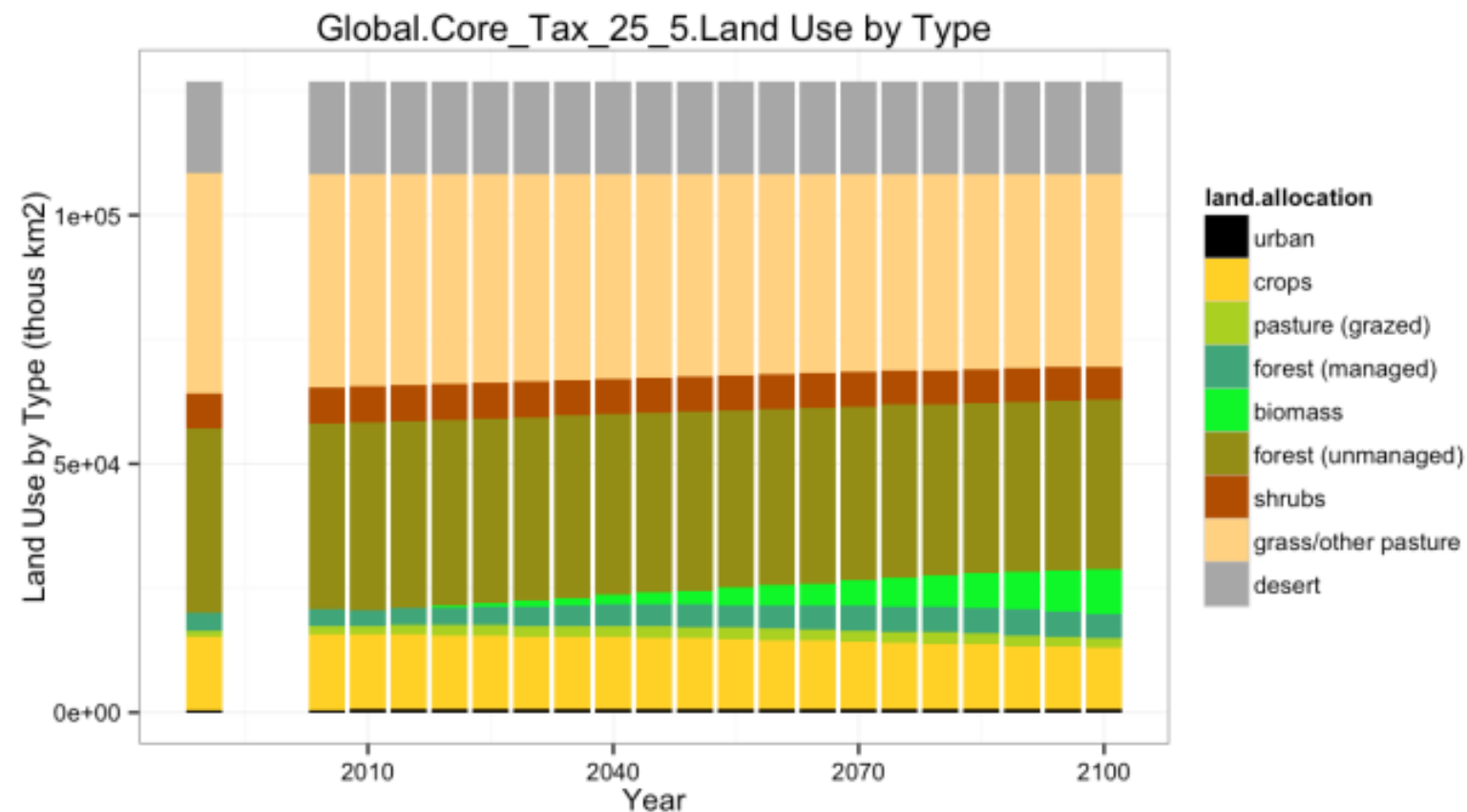


# Land-Related Policies

- Protected Lands (REDD):
  - With this policy, we can set aside some land, removing it from economic competition. This land cannot be converted to crops, pasture, or any other land type.
  - The default in GCAM is to protect 90% of all non-commercial ecosystems.
- Valuing carbon in land:
  - In a policy regime, we can choose to put a price on land-use change CO2 emissions that is equal to the price on fossil fuel and industrial CO2 emissions.
  - We model this policy as a subsidy to land-owners for the holding carbon stocks.
- Bioenergy constraints (upper or lower):
  - We can impose constraints on bioenergy within GCAM. Under such a policy, GCAM will calculate the tax or subsidy required to ensure that the constraint is met.
  - By default a bioenergy constraint in GCAM is imposed based on the amount of subsidy available for net negative emissions.

# Land-Related Policies

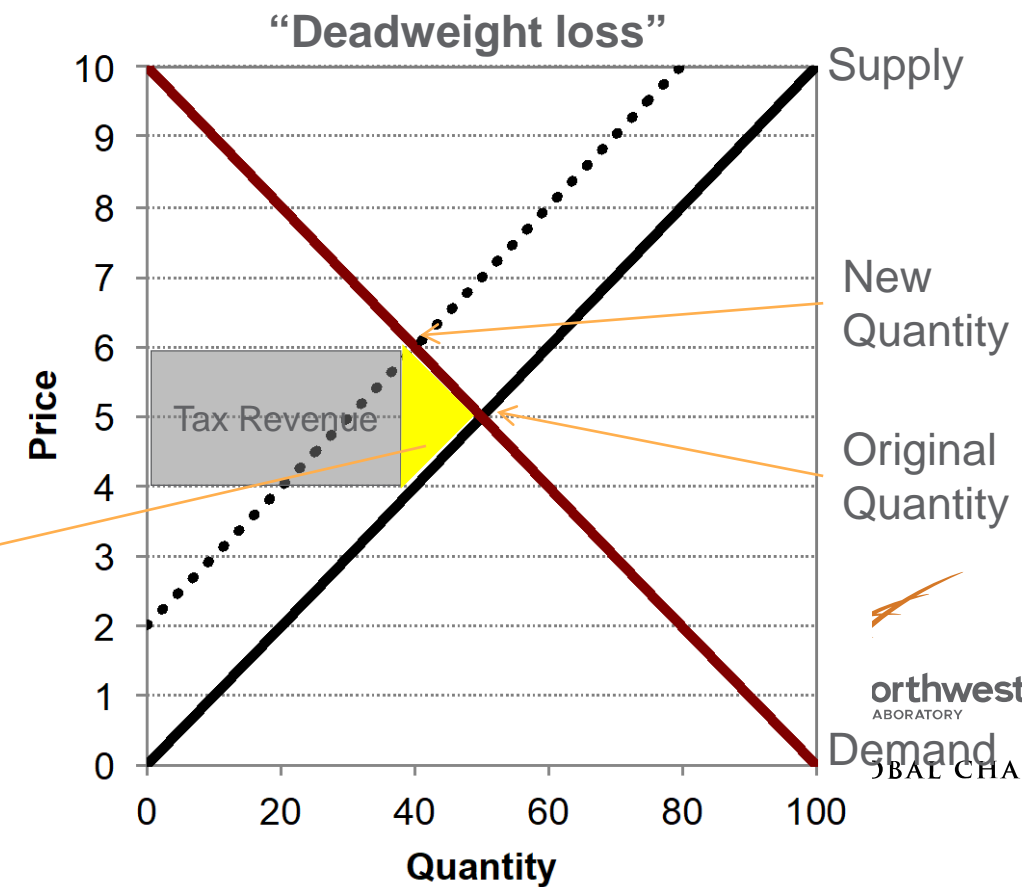
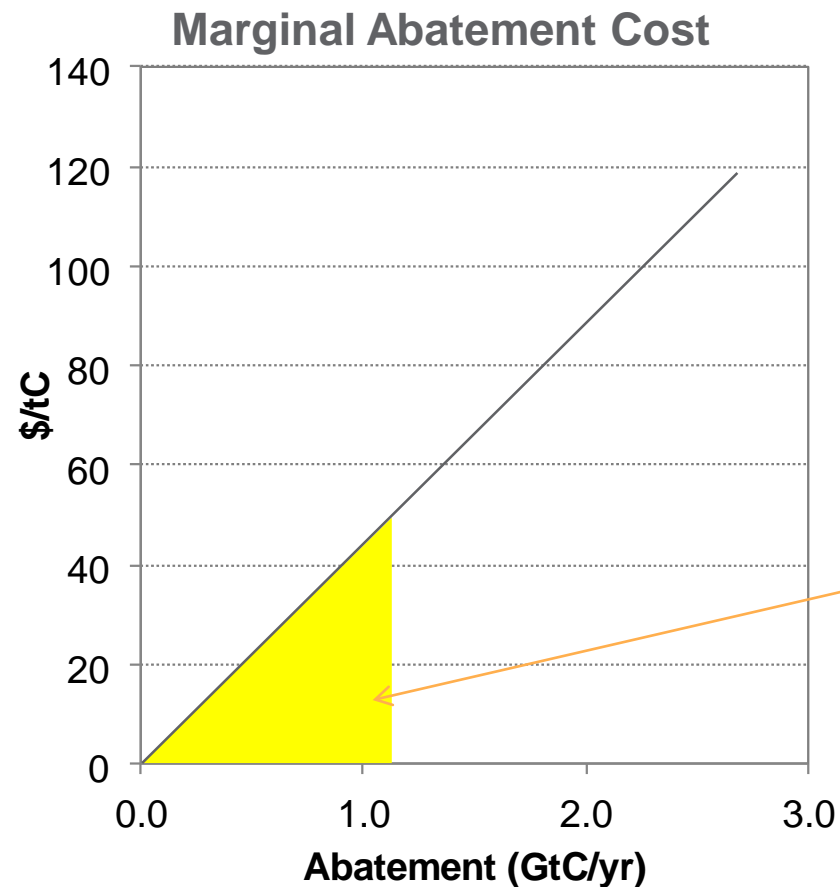
- Under the default assumption in GCAM, 90% of non-commercial ecosystems are protected in GCAM. This means that they cannot be used for crop or bioenergy production.





# Other Markets: Policy Cost

- GCAM can compute the cost of a some policies.
- The cost metric used is the area under the marginal abatement cost (MAC) curve. This area under the MAC curve commonly referred to as “deadweight loss” (i.e., the change in producer and consumer surplus.)
- Currently, we are not modeling this cost as affecting GDP in GCAM.



# Treatment of Existing Policies in GCAM

- Question:
  - Does the GCAM reference scenario include other climate and energy policies?
- Answer:
  - To the extent that these exist in the base year, they will be calibrated into the GCAM reference scenario.
  - However, we do not explicitly include any proposed climate or energy policies in the reference scenario.



An aerial photograph of a coastal landscape. The top half of the image shows a patchwork of green fields and brown, possibly agricultural or industrial, areas. A winding river or stream flows through the landscape. The bottom half of the image shows a blue body of water, likely a bay or estuary, with a white sandy beach visible along the shoreline. The text "Thank you" is overlaid in the center of the image.

**Thank you**