



GLOBAL CHANGE INTERSECTORAL MODELING SYSTEM

# Uncovering Key Drivers of Future Virtual Water Trade and Global Water Use

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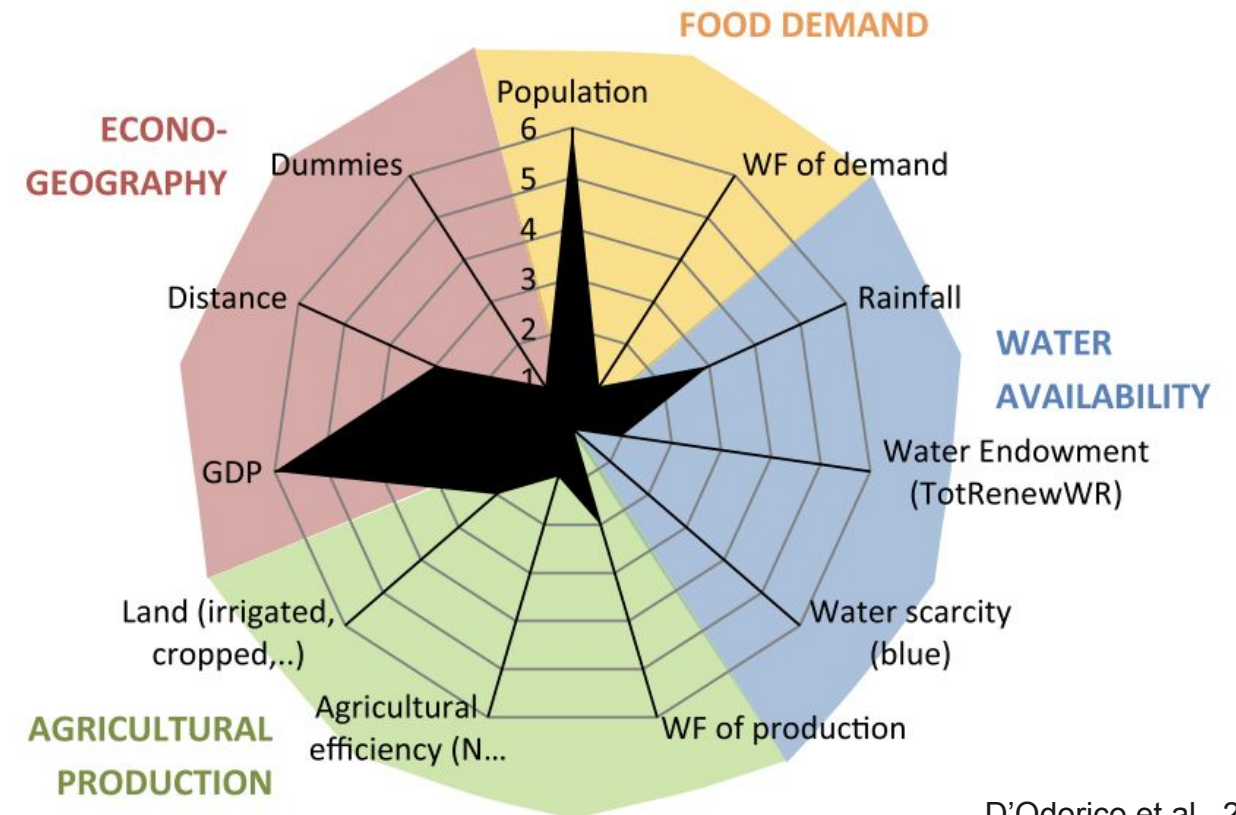


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- International trade redistributes water around the world, virtually, to allow regions to meet demands while having limited internal water resources or inadequate comparative advantages of land and labor productivity.
- Projected future growth, globalization, and continued market integration will drive further agricultural trade.
- Future projections of change significantly lack in the literature.



**Figure 13.** Drivers of VW trade identified in the publications reviewed in section 7. The radial coordinate expresses the number of publications reporting a significant dependence on each variable. Details are provided in the supplementary information (table S1 is available online at [stacks.iop.org/ERL/14/053001/mmedia](https://stacks.iop.org/ERL/14/053001/mmedia)).



Soil moisture and evapotranspiration consumed in crop growth

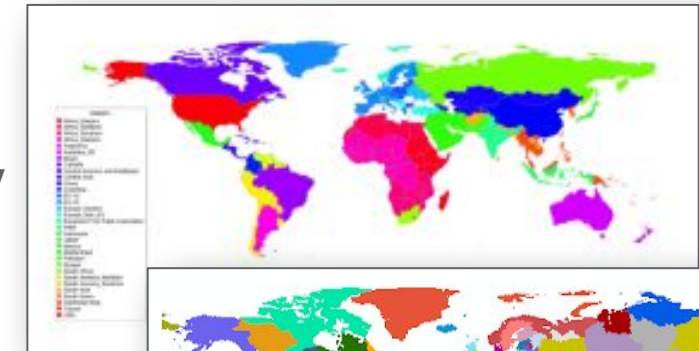


Fresh surface runoff and groundwater consumed in the production of a good or service

## Global Coverage

- **Leveraging a Scenario Discovery Framework we ask:**
  - **What are the main drivers of change for global and regional agricultural water use utilization and future virtual water trade across a large ensemble of uncertain futures?**
- GCAM is a **multisector, multi-regional, dynamic model** that **economically and physically links** energy, agriculture and water production, consumption, and trade.
- GCAM models **dynamic demand growth** in response to regional population and economic activity.
- GCAM includes **technology detail in energy** production, transformation and final demand sectors.
- GCAM includes **physical representations of crop management practices** in the agriculture sector.
  - Includes all commercial and natural lands in each land region.
- GCAM includes **dynamic economic modeling of the water** sector linked to energy and agriculture.

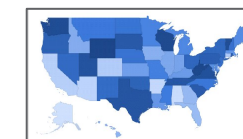
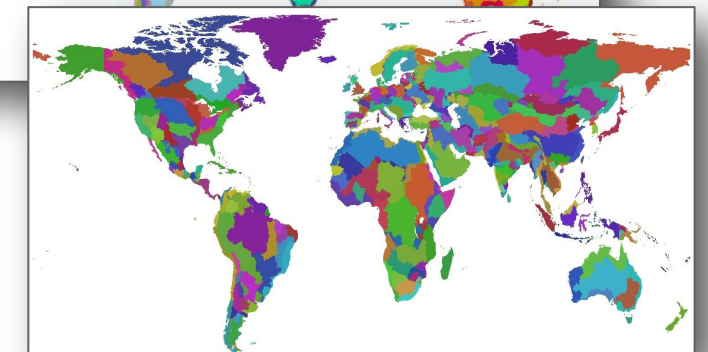
32 Energy & Economy Regions



235 Water Basins



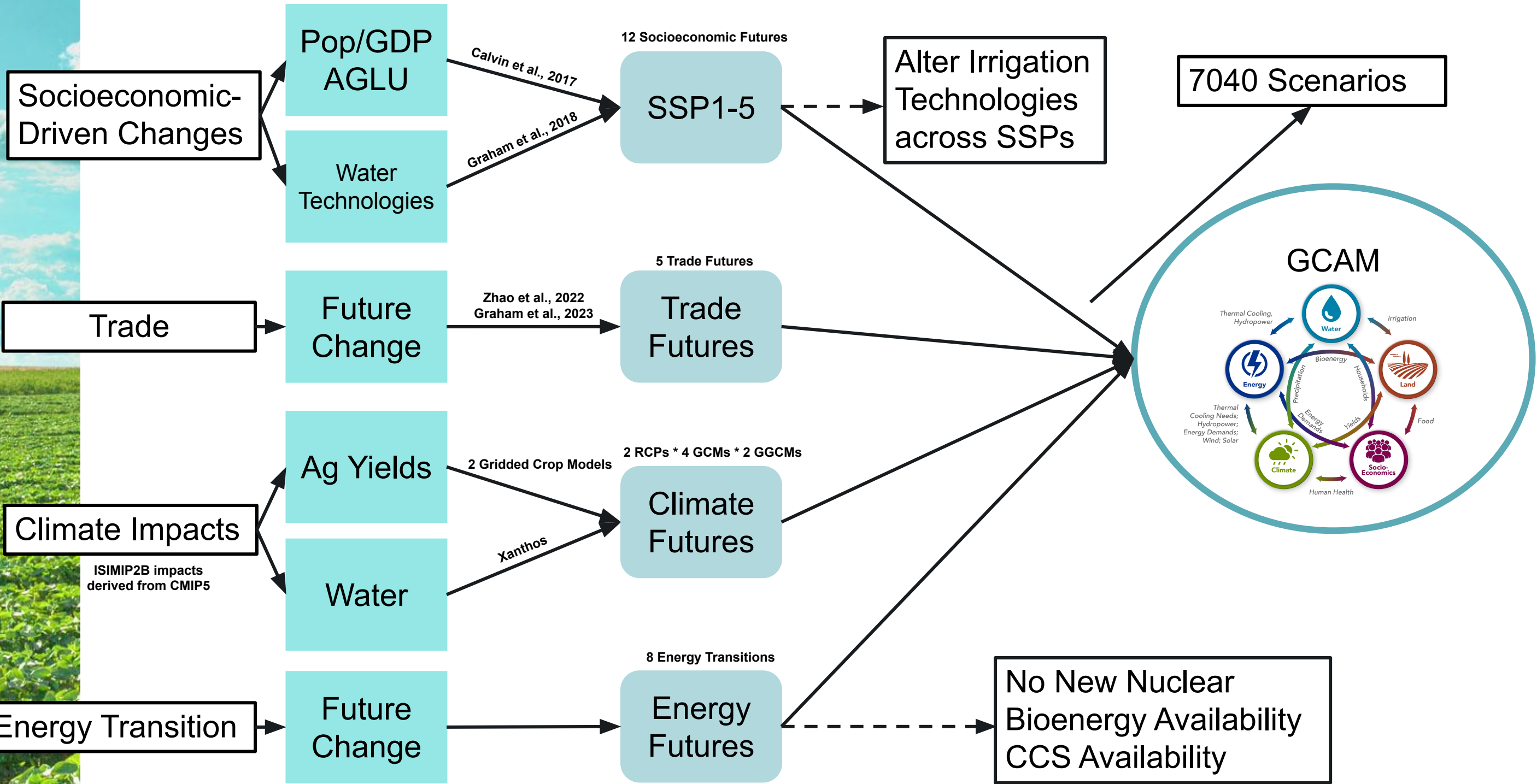
384 Land Regions



GCAM-USA (50 State)

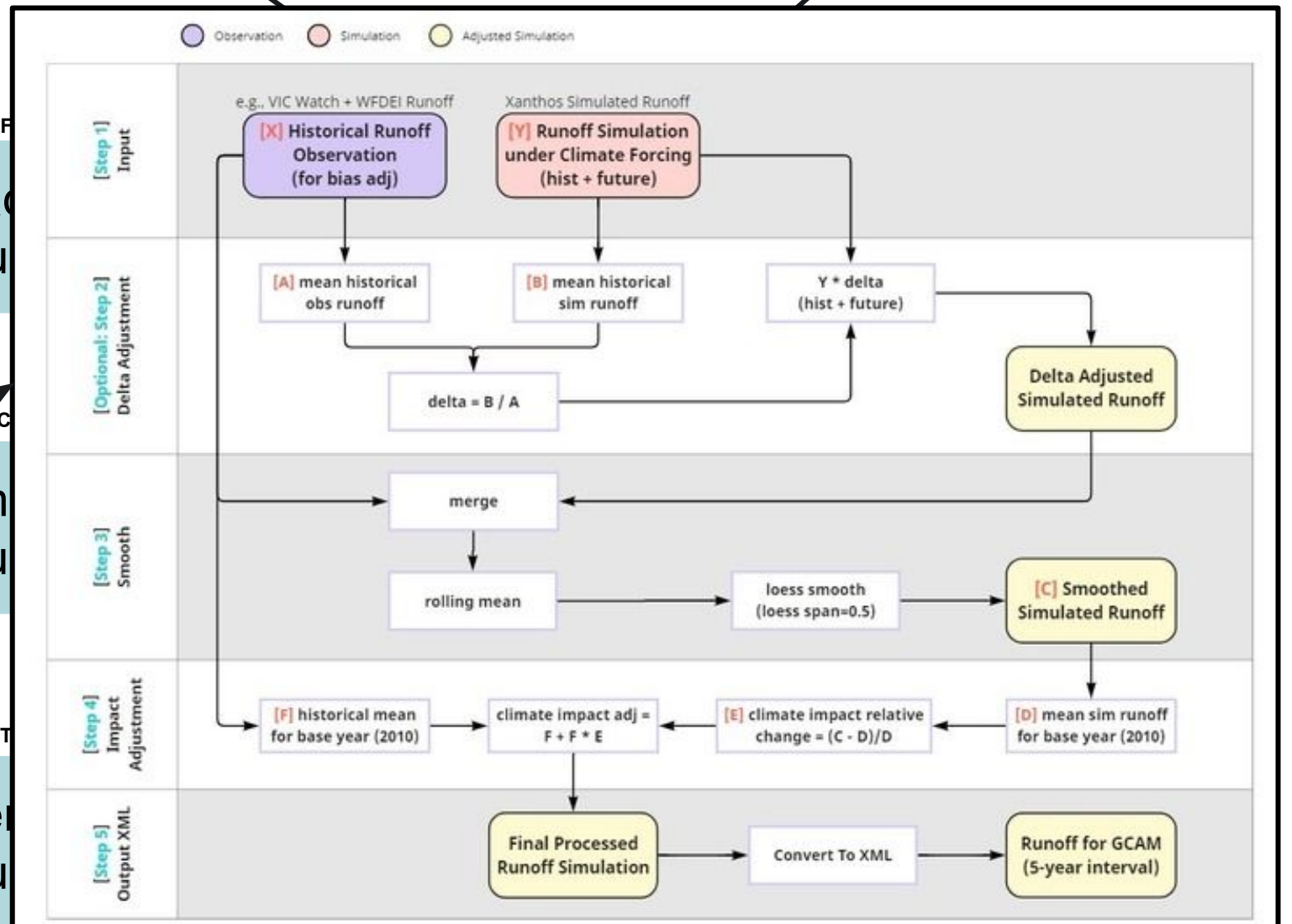
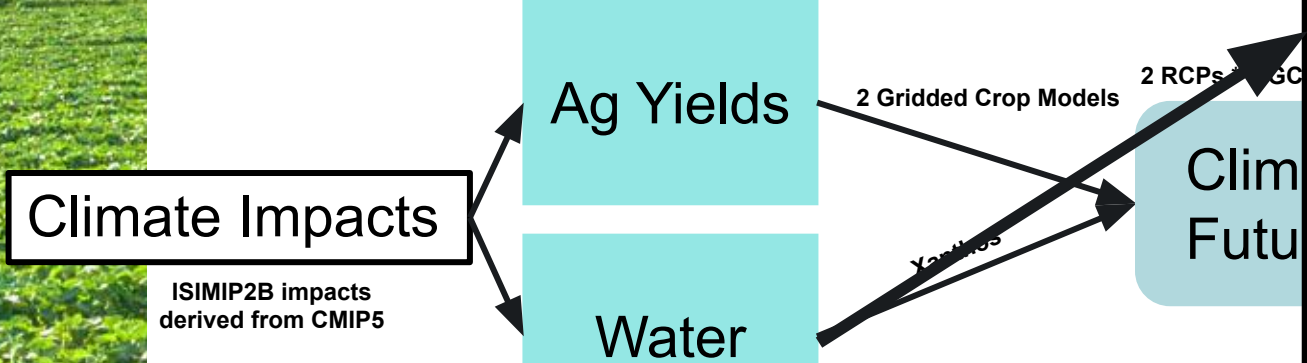
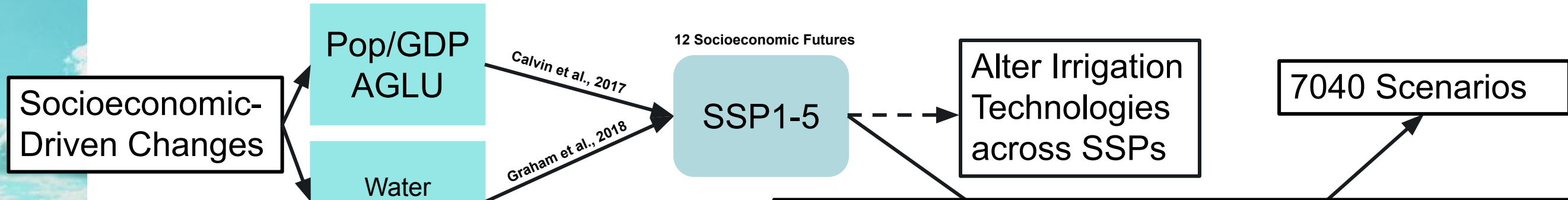


# Scenario Discovery Workflow



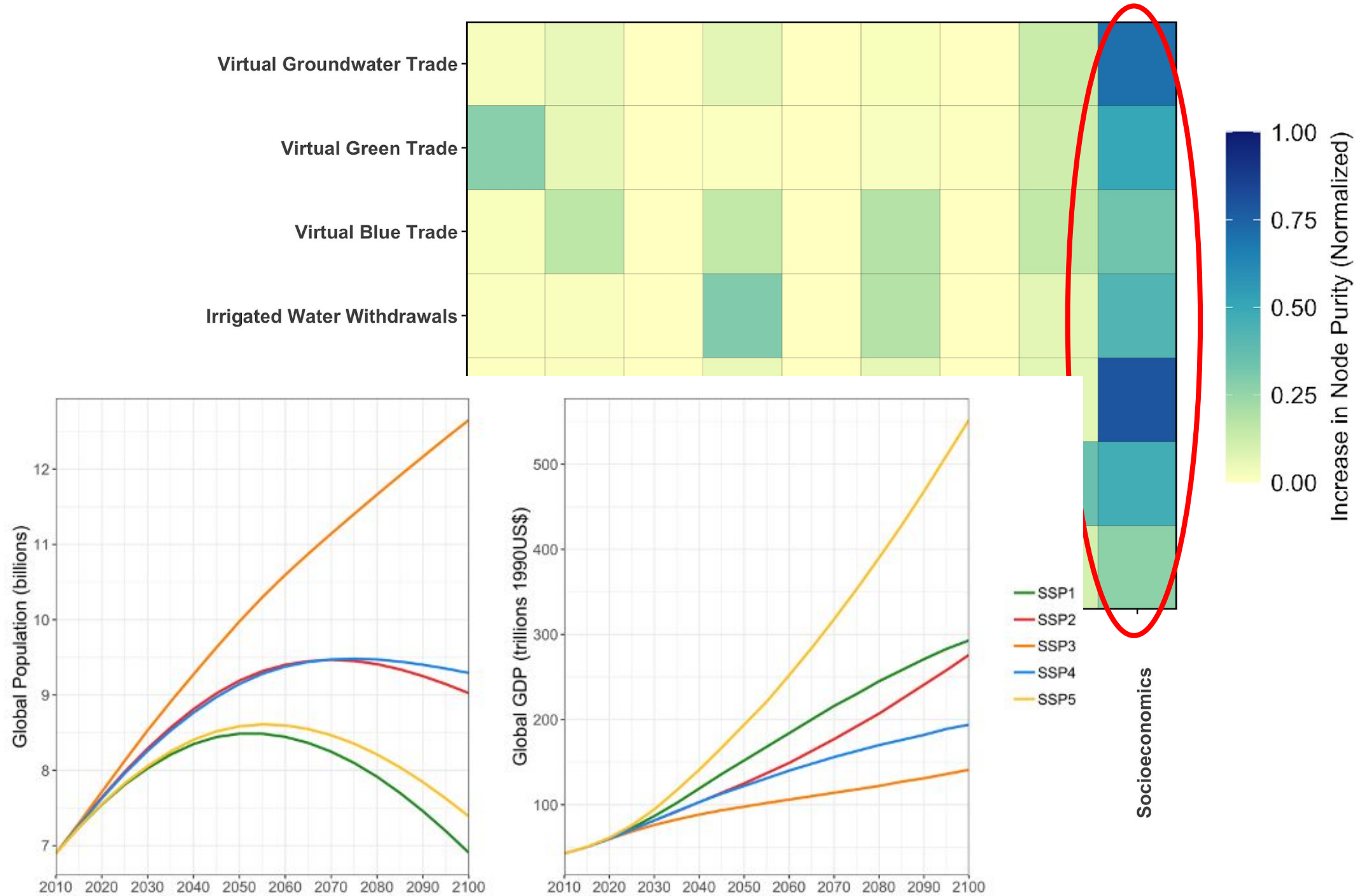


# Scenario Discovery Workflow



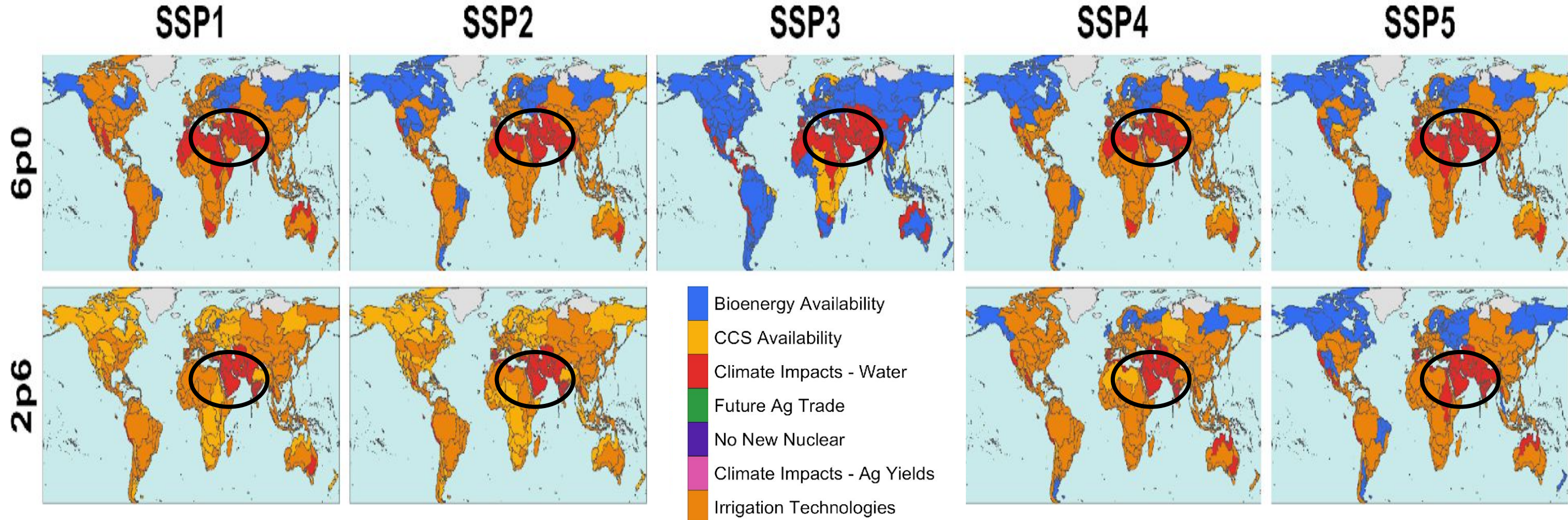


# Global drivers of change are dominated by socioeconomic divergences in the SSPs, but the story becomes increasingly complex with more refined scenario selection



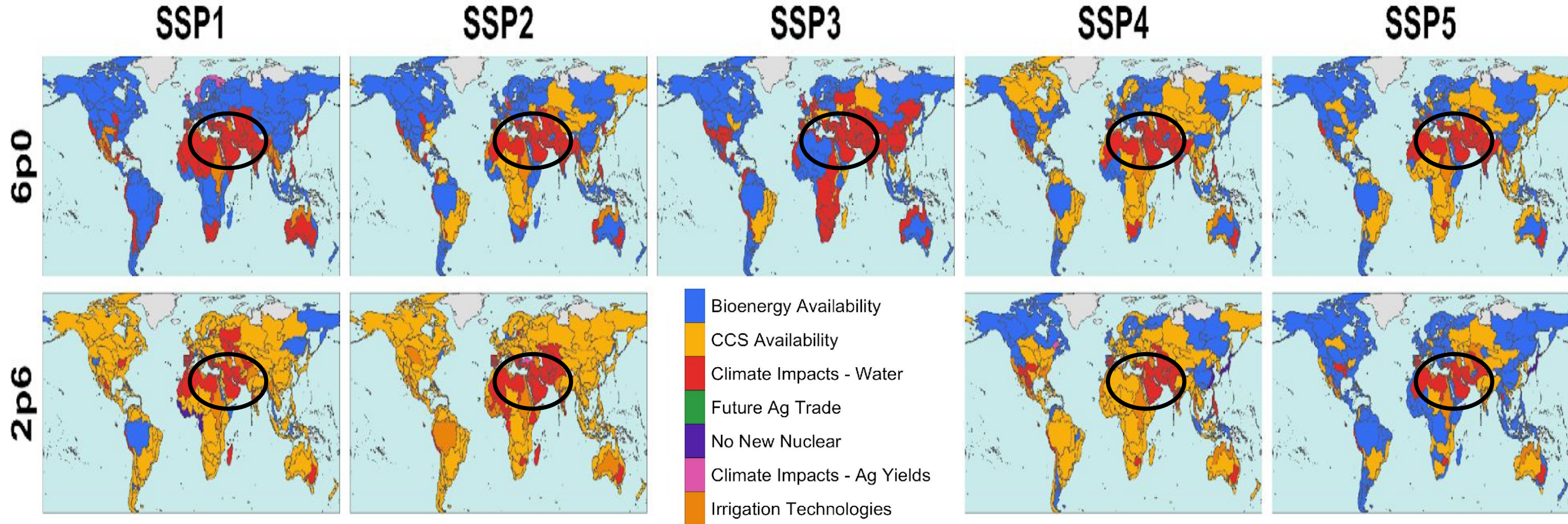


# Irrigation Water Withdrawals are largely driven by the accessibility of higher efficiency technologies and the water savings associated with such adoption





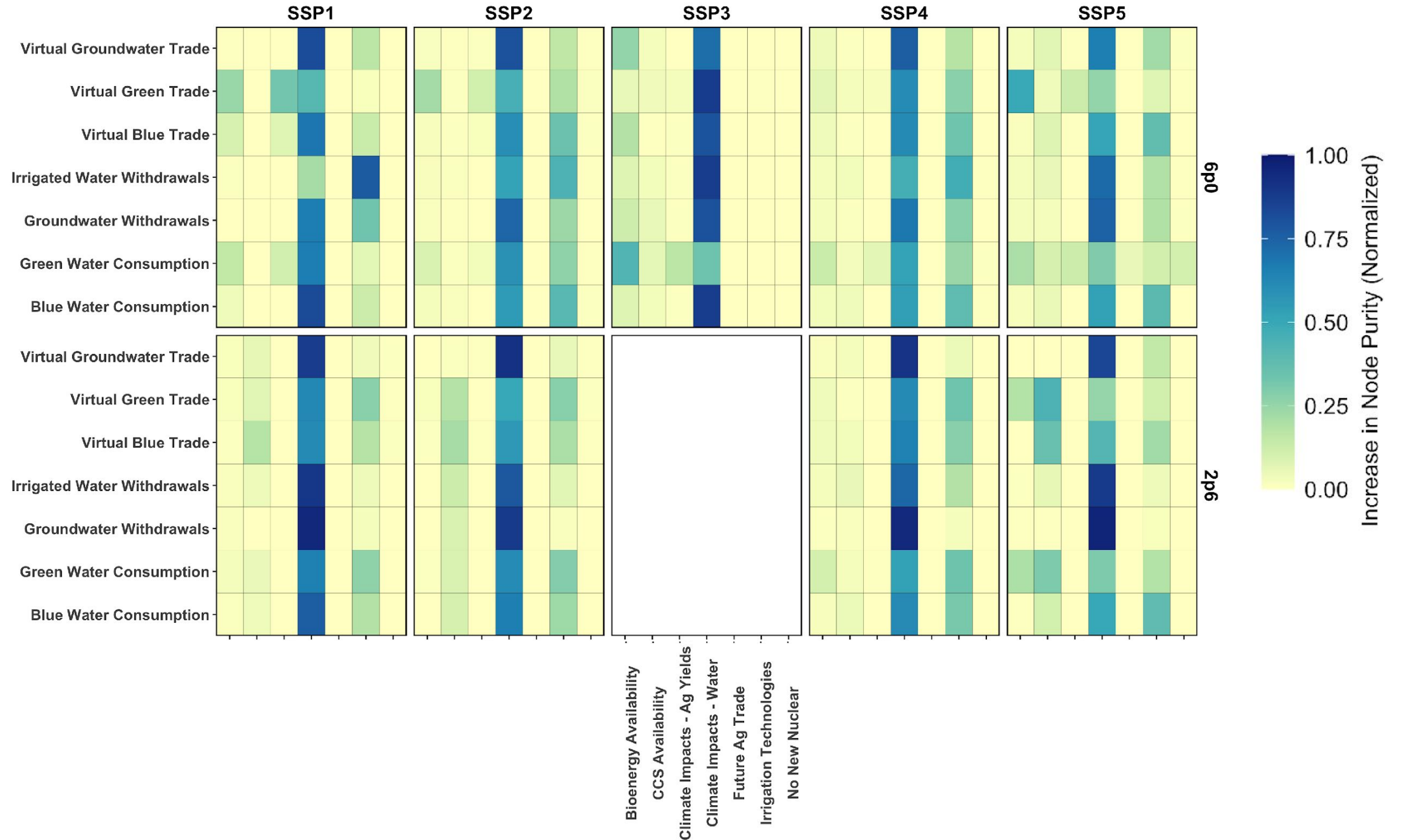
In some basins, despite compounding influences, the main drivers of change for multiple metrics remain consistent across SSPs and RCPs, highlighting the importance of select drivers in some regions





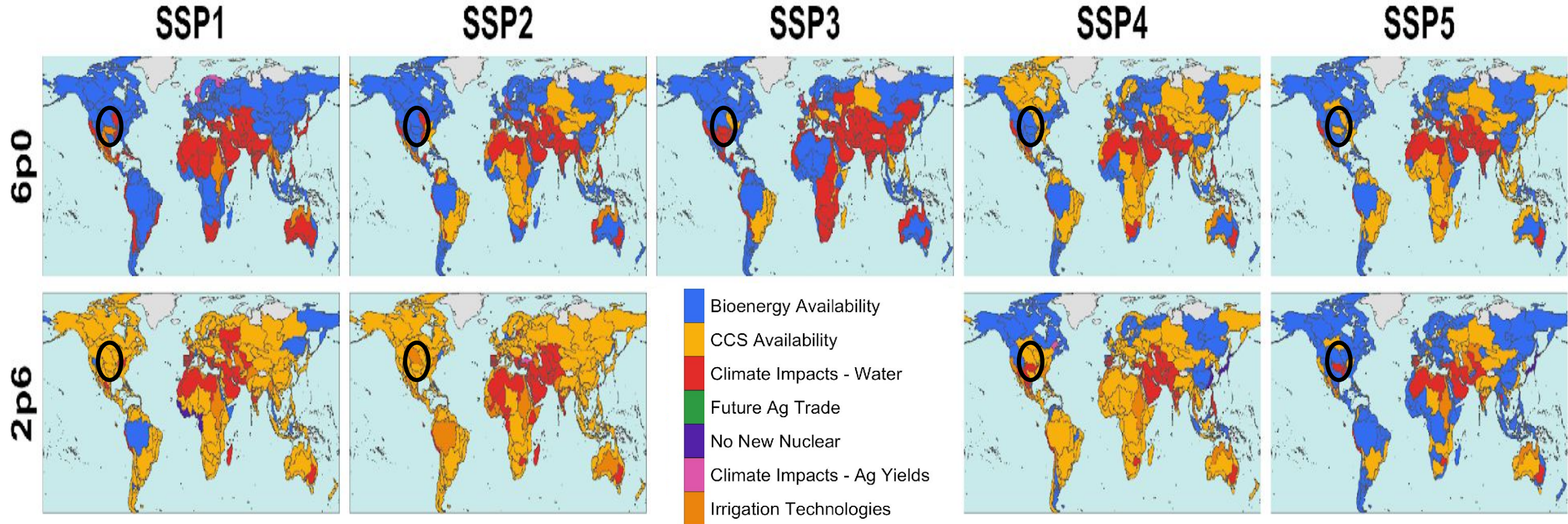
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Arabian\_Peninsula in 2100





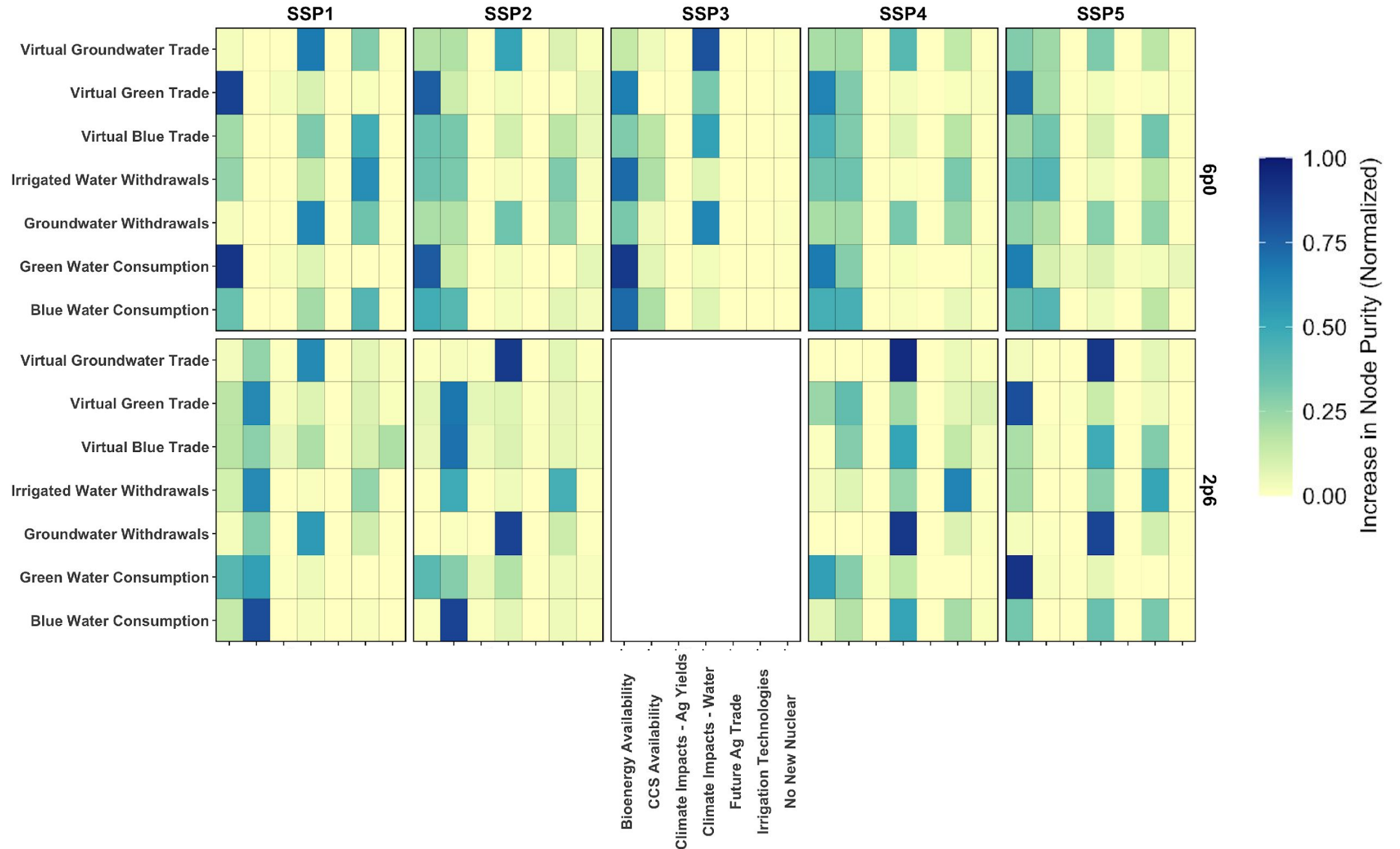
Virtual Blue Water Trade can have differing main drivers in a basin, depending on socioeconomic and climatic conditions – highlighting regional sensitivities to such dynamics





# Virtual Blue Water Trade can have differing main drivers in a basin, depending on socioeconomic and climatic conditions – highlighting regional sensitivities to such dynamics

Arkansas\_White\_Red in 2100





- First such study to examine a large suite of future uncertainty to uncover key drivers in water utilization and trade.
- While drivers show regional heterogeneity, ***dry areas are shown to often be driven by the climate-driven water supply responses*** while other basins show differences across socioeconomic and climate futures.
- Such basin-level differences allows us to further our exploration to understand the multisectoral drivers behind these changes in water.
  - Agricultural demand & production
  - Energy demand and mix
  - Regional trade
- Despite the differences across basins at the end of the century, ***it is important to understand how these drivers change in time.***



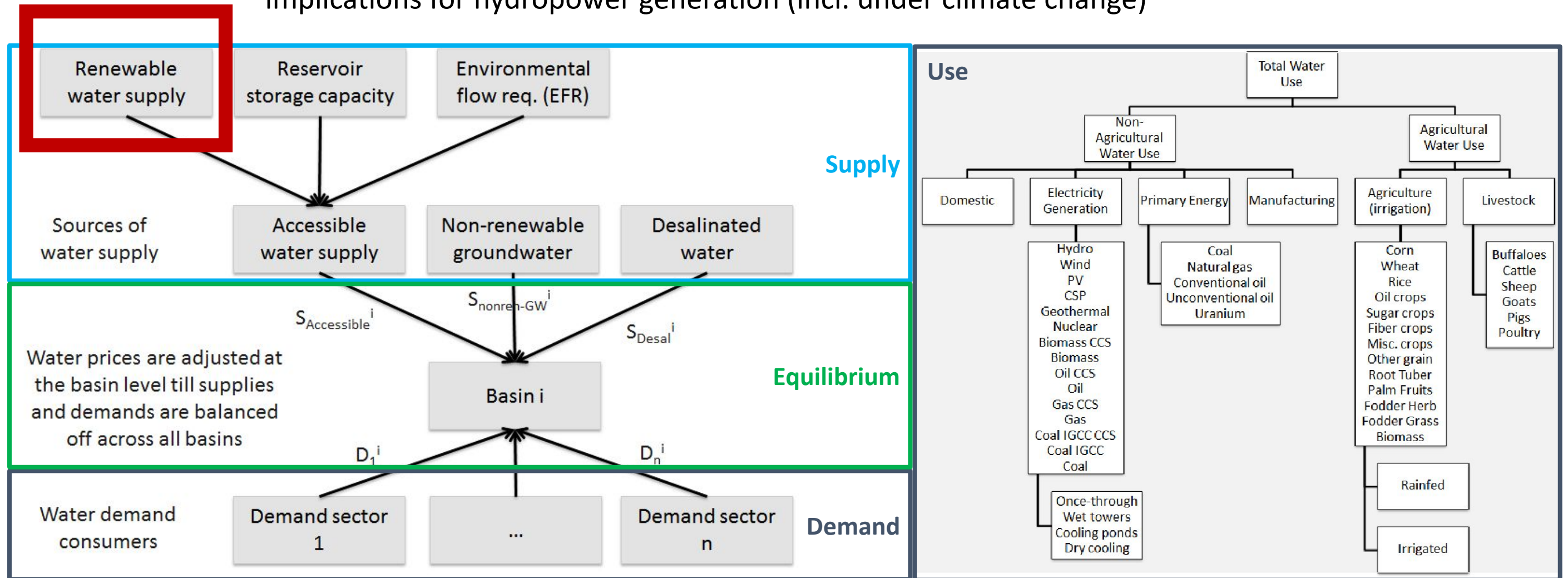
# Thank you

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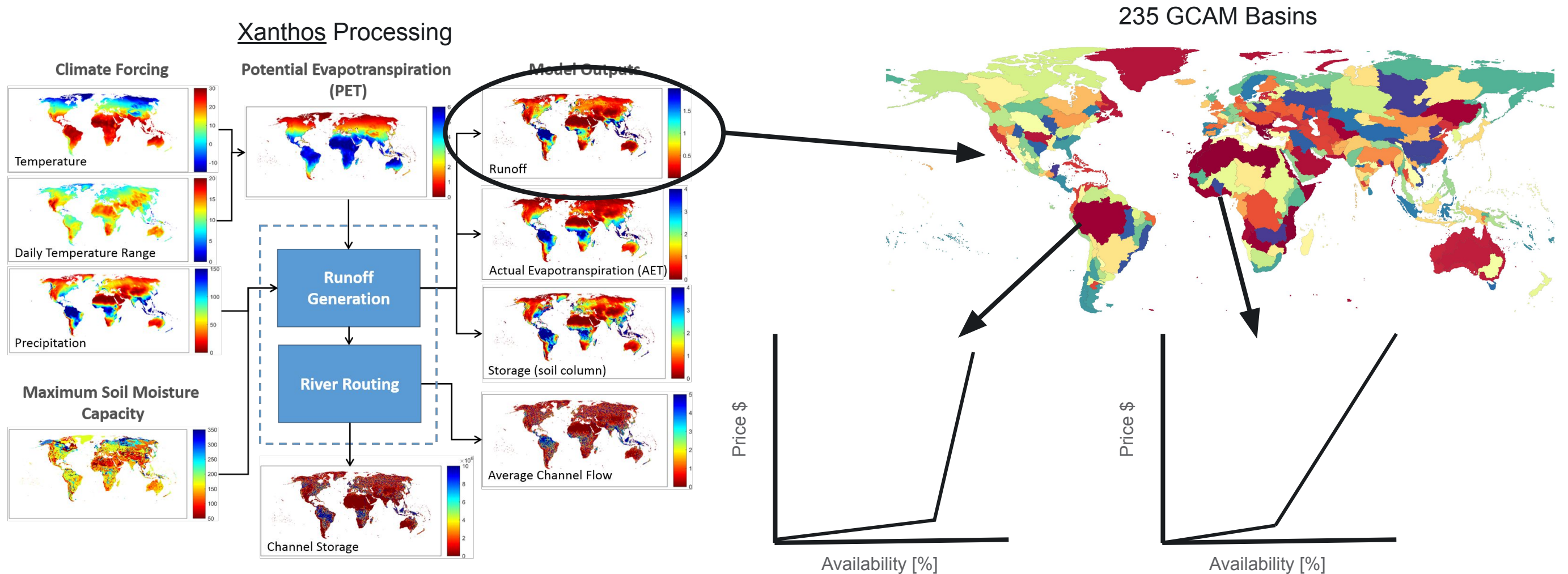
# Water Demand and Supply are Balanced Across 235 Global River Basins. Water demands by tech. (incl. hydro) are tracked. Water availability produced by a separate hydrology model.

We have a hydrology model that generates information about renewable water supply, and its implications for hydropower generation (incl. under climate change)



Kim et. al (2016)

Water prices allow us to capture two key dynamics: increasing shadow price with increasing scarcity; and preferential water allocation to classes of users (e.g., agriculture versus electricity)

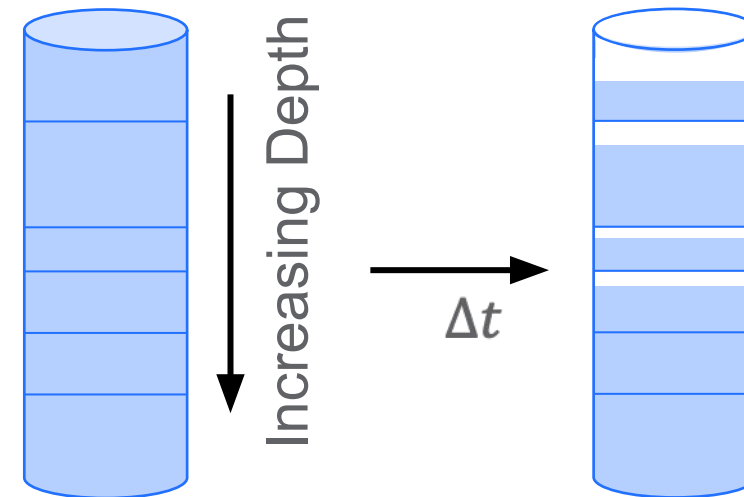
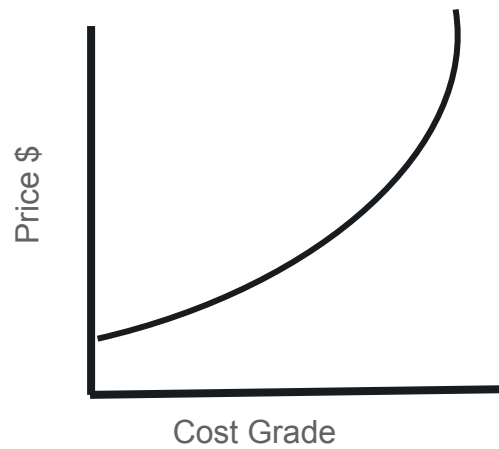
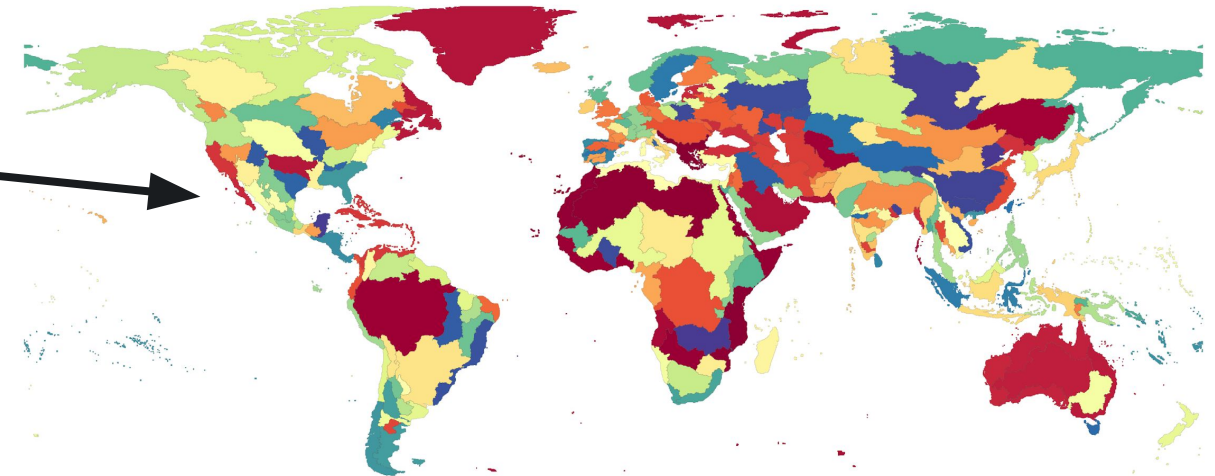


1. Calculate maximum runoff and accessible water\* for each year using Xanthos and 1970-2015 WATCH data
2. Aggregate 0.5° gridded data to 5-year moving averages at GCAM basin scale.
- 3A. For basins with no historical nonrenewable groundwater depletion find the accessible percentage of maximum runoff (accessible/runoff) **cost curve inflection point**
- 3B. For basins with historical nonrenewable groundwater extraction, back calculate accessible portion by  $\frac{(Demands - Depletion)}{runoff}$  averaged over historical years **cost curve inflection point**
4. Set initial supply grades (0%, accessible percent, 100%) and cost grades (~0, 0.001, 10), then interpolate for 20 total grades



Superwell  
Derived unit cost of groundwater extraction over time at 50km grid cells

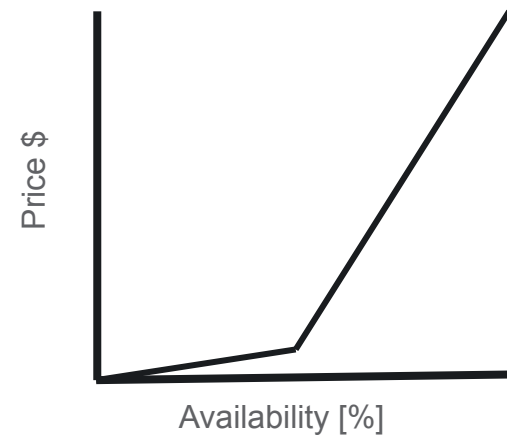
- Estimates of groundwater availability from previous studies on porosity, depth-to-groundwater, and aquifer thickness in addition to WHYMAP
- Constrained to only allow for 25% use of all physical water estimated in aquifer
- Calculations within require an assumption of confined aquifer, however results assumed to be from unconfined aquifer



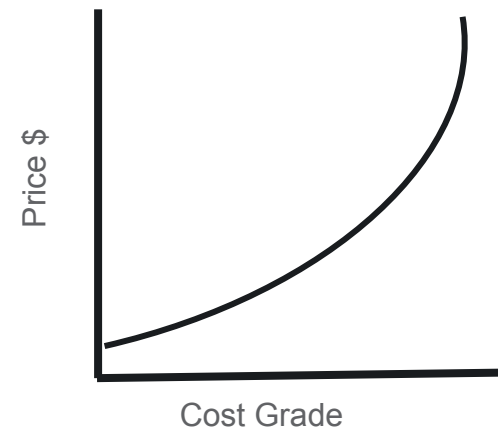
1. Aggregate unit costs from Superwell 50km grid to GCAM basin scale and transform into 24 unique cost grades
2. Historical groundwater depletion from WaterGap is placed into grade\_hist which is pulled during historical calibration
3. Added to GCAM as a subresource with price interactions with renewable water cost curves.

# Surface Water and Groundwater Price Interaction

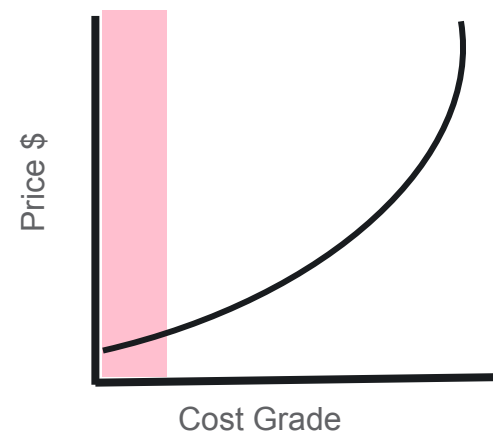
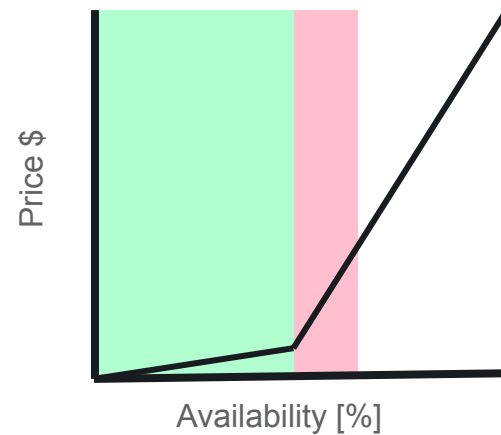
Renewable Cost Curve  
Basin A



Groundwater Cost Curve  
Basin A

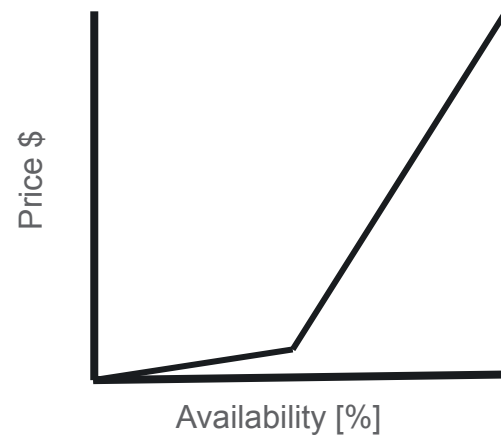


1. Water withdrawals come from cheap renewable sources first.
2. If demands exceed the accessible portion in any given timestep, a price interaction between groundwater and renewable water occurs where water is drawn from the cheaper source first.

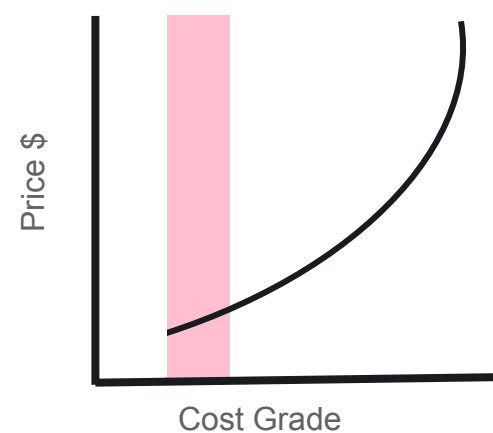
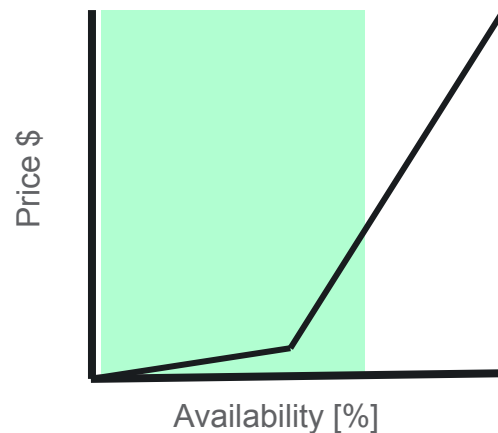
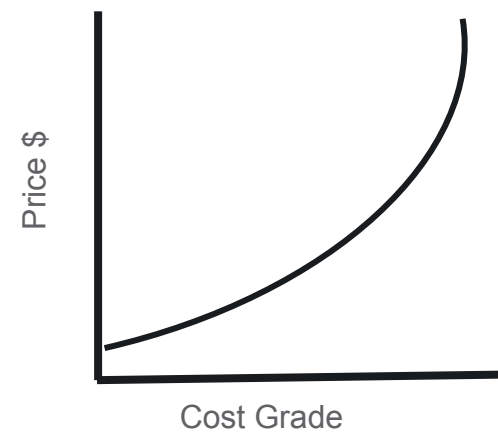




Renewable cost curve  
Basin A

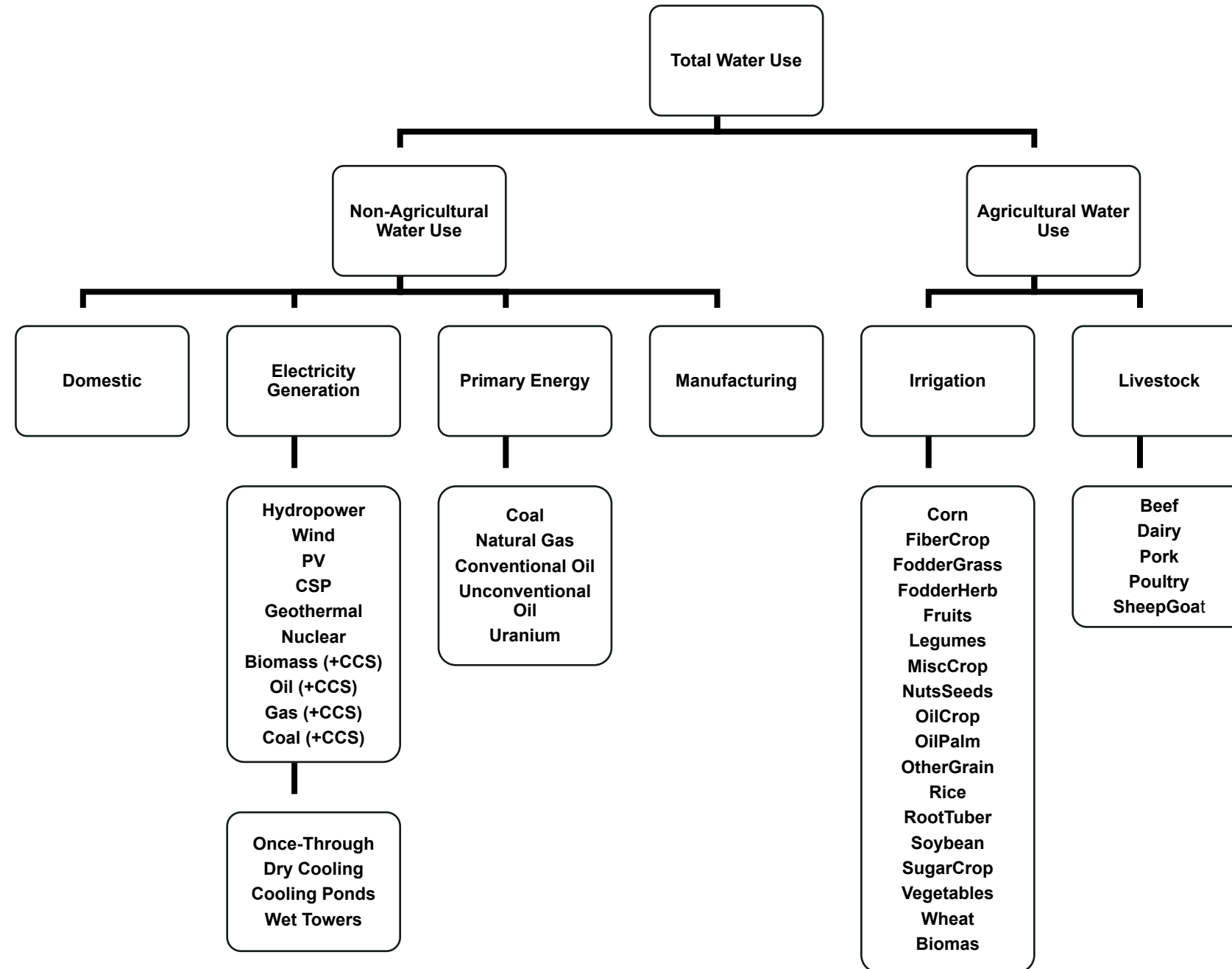


Groundwater cost curve  
Basin A



1. Water withdrawals come from cheap renewable sources first.
2. If demands exceed the accessible portion in any given timestep, a price interaction between groundwater and renewable water occurs where water is drawn from the cheaper source first.
3. As groundwater is exhausted in low-cost grades, the point of interaction between renewable and groundwater is pushed to higher prices
4. More water than is deemed accessible must be pulled in order to start the price interaction with groundwater

# Sectoral Water Demands in GCAM





- Ag production is separated into Irrigated (IRR) or Rainfed (RFD) to calculate trade and virtual water trade at the basin level,  $b$ , for crop  $c$ .
  - As demand is calculated at regional scales, proportionate values of production and water are used to downscale to basin level
- Groundwater contributions to virtual water trade are calculated based on the proportion of nonrenewable groundwater depletion,  $GWD$  to blue water withdrawals,  $BWW$ , within the specified basin

## Virtual Water Trade

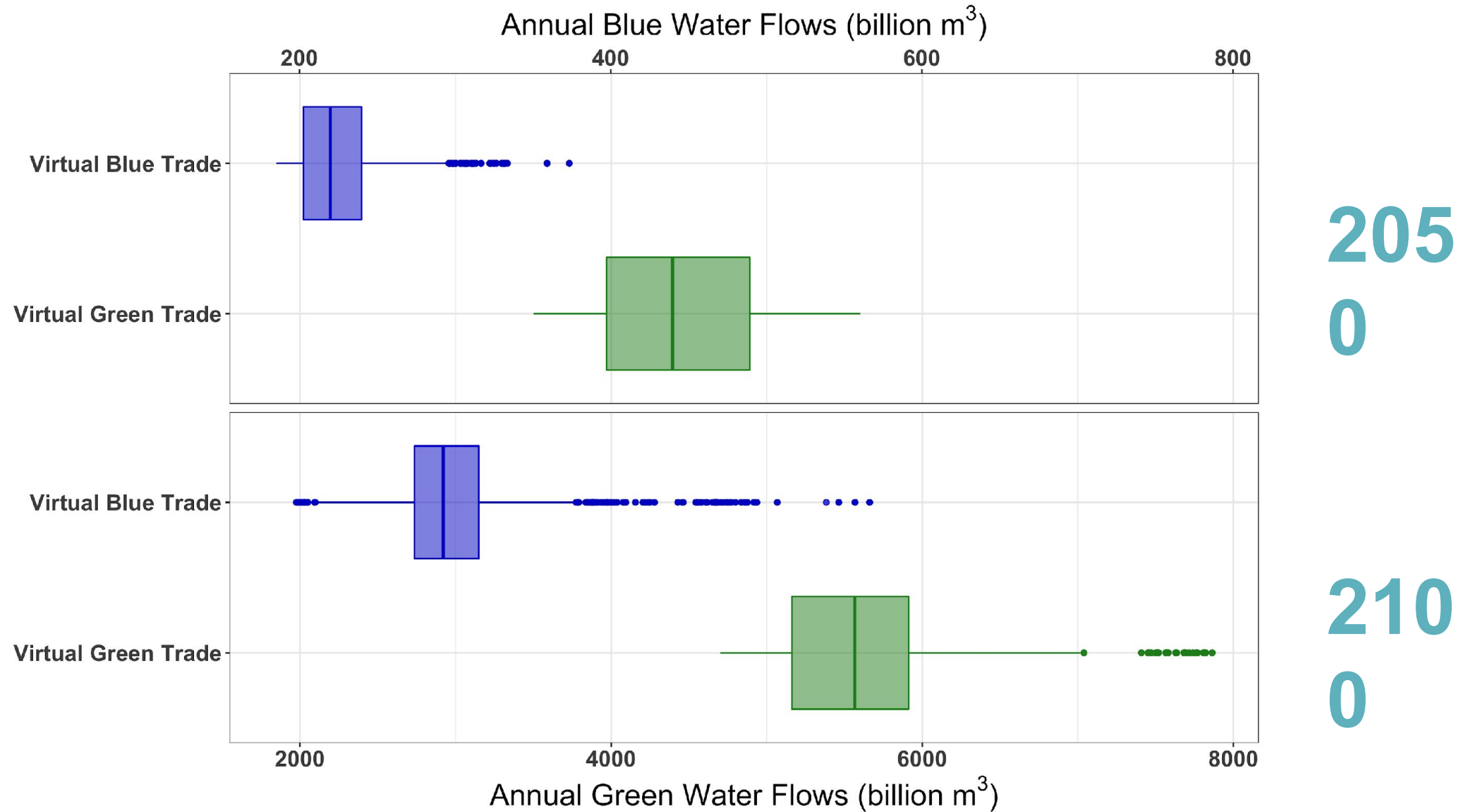
$$VBE_{b,c}(t) = BWC_{b,c}(t) * \left( \frac{E_{b,c,IRR}(t)}{P_{b,c,IRR}(t)} \right)$$

$$VGE_{b,c}(t) = GWC_{b,c}(t) * \left( \frac{E_{b,c,RFD}(t)}{P_{b,c,RFD}(t)} \right)$$

## Virtual Groundwater Trade

$$VGWE_{b,c}(t) = VBE_{b,c}(t) * \frac{GWD_b(t)}{BWW_b(t)}$$

# Future Virtual Water Trade Projections





# Main Drivers of Nonrenewable Groundwater Withdrawals (2100)

SSP1

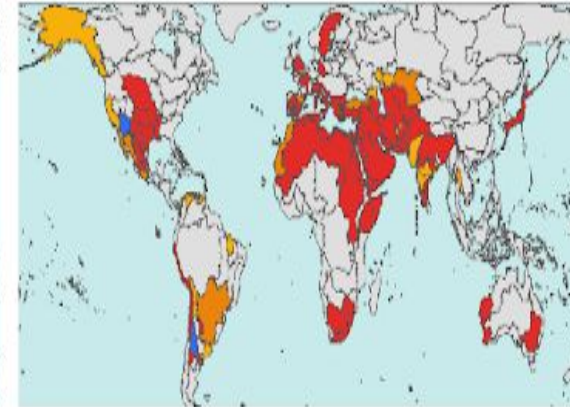
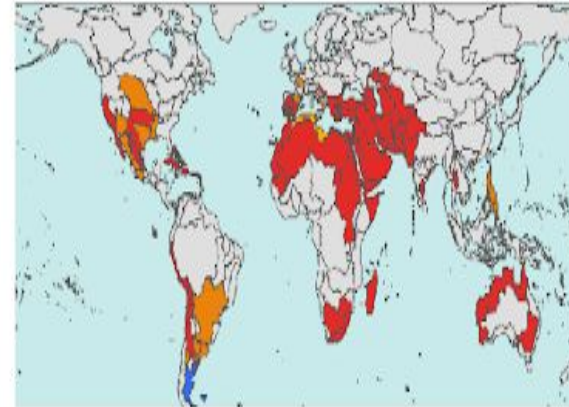
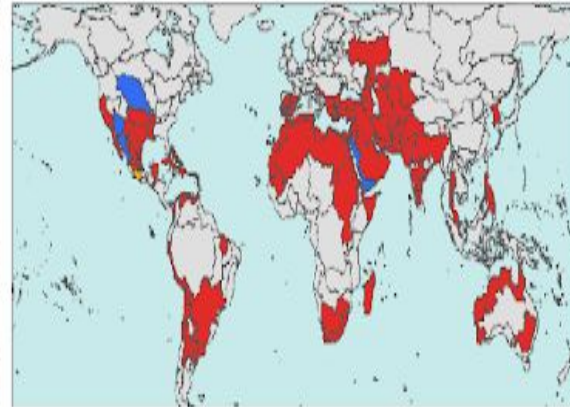
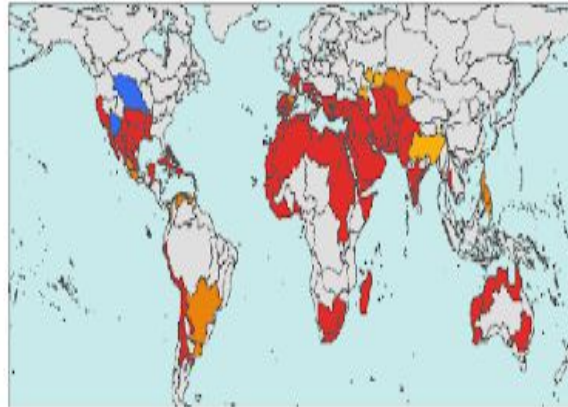
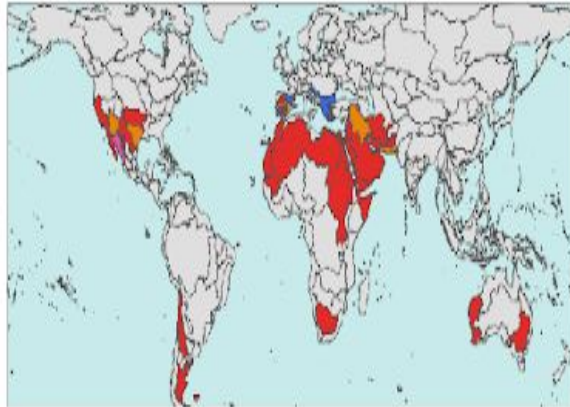
SSP2

SSP3

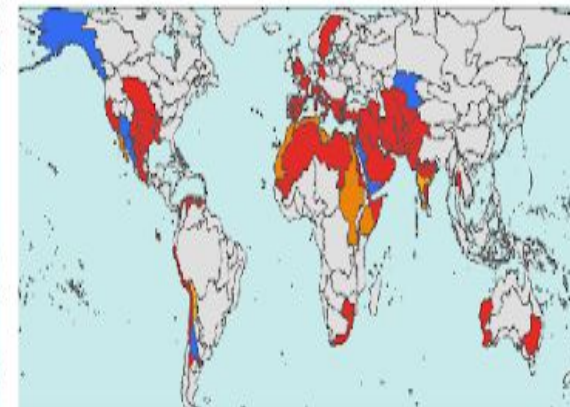
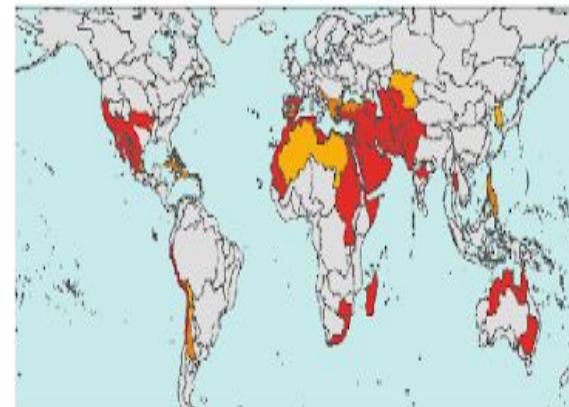
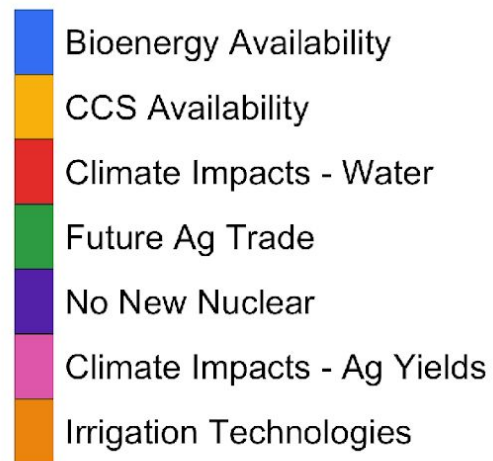
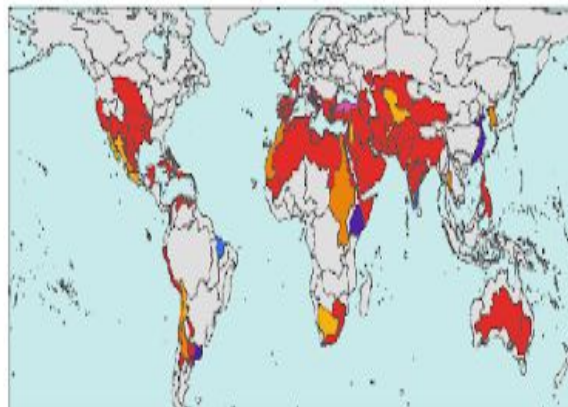
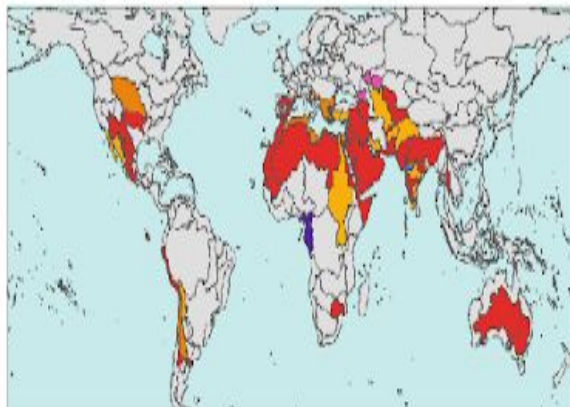
SSP4

SSP5

6p0



2p6



# Mean Global Electricity Generation

