Conceptual Road Map

Model Structure

The model consists of a nested structure:

- 1. Outer loop: Climate policy maker (social planner)
- 2. Inner loop: Two-Period General Equilibrium(GE) Model

Markov Decision Process Formalization

The system evolves as a Markov Decision Process (MDP) where:

$$S_t = f(a_t | S_{t-1})$$

Where:

- S_t is the state vector at time t containing:
 - \circ Economic variables from GE model (Y_t, C_t , etc.)
 - \circ Cumulative emissions $E_t = \sum_{i=0}^t \eta_i Y_i$
 - \circ Current damage function parameters $heta_t$
- a_t is the policy action vector: $[au_t, au_{t+1}]$
- $f(\cdot)$ is the transition function (Two-Period GE model)

Policy Maker's Problem

Objective Function

The policy maker maximizes expected social welfare:

$$\max_{a_t} \mathbb{E}\left[\sum_{t=0}^{\infty} eta^t \left(u(S_t) - D(E_t, heta_t) Y_t
ight)
ight]$$

Where:

- ullet eta is the social discount factor
- $u(S_t)$ is the social utility function

- $D(E_t, \theta_t)$ is the damage function
- ullet Y_t is aggregate output
- E_t is cumulative emissions
- $oldsymbol{ heta}_t$ is the vector of damage function parameters

Damage Function Learning

The damage function uncertainty evolves as:

$$\theta_t = g(E_t, \theta_{t-1}, \epsilon_t)$$

Where:

- $g(\cdot)$ represents the learning process
- ϵ_t is new information received in period t
- ullet Uncertainty band narrows as $t o\infty$
- ullet Distribution shape depends on E_t

Policy Action Space

 $a_t \in A$ where A is the discrete action space:

$$A = \{\tau_t \pm \{5, 10\} \text{ basis points}\}$$

Agent Expectations

Agents in the GE model form expectations of future carbon taxes:

$$\mathbb{E}_t[au_{t+1}] = h(a_t, \Omega_t)$$

Where:

- Ω_t is the information set at time t containing:
 - \circ Policy maker's announced rates $[au_t, au_{t+1}]$
 - Policy maker's objective function
 - \circ Current state S_t
- $h(\cdot)$ is the expectation formation function

Time Structure

1. Period t begins with state S_t

- 2. Policy maker observes S_t and chooses $a_t = [\tau_t, \tau_{t+1}]$
- 3. Agents observe a_t and form expectations
- 4. GE model solves for equilibrium
- 5. Damages realized, new information ϵ_t received
- 6. System transitions to S_{t+1}

Implementation Plan

1. State Management

```
mutable struct StateNode
    time::Int
    economic_state::Dict{String, Float64}  # Y_t, C_t, etc.
    emissions::Float64  # E_t
    damage_params::Vector{Float64}  # θ_t
    policy_actions::Vector{Float64}  # [τ_t, τ_{t+1}]
    parent::Union{Nothing, StateNode}
    children::Vector{StateNode}
    debug_log::Vector{String}
end
```

2. Transition Function Integration

1. GE Model as Inner Loop

- Use existing GE model as f(a_t|S_{t-1})
- Extend equilibrium computation to track emissions
- Add damage calculations to economic outcomes

2. State Transitions

```
function state_transition(current_state, policy_action)
    # 1. GE Model Equilibrium
    GE_equilibrium = compute_equilibrium(policy_action)

# 2. Emissions Accumulation
    E_t = current_state.emissions + η_t * GE_equilibrium["Y_t"]

# 3. Damage Learning
    θ_t = damage_function_learning(current_state.damage_params, E_t)

return new_state
end
```

3. Monte Carlo Policy Search

- 1. Graph Building
 - Start from initial state S₀
 - For each time period:
 - Generate policy action combinations
 - Compute state transitions
 - Track feasible paths

2. Path Evaluation

```
function evaluate_path(path)  
# Calculate expected social welfare:  
# \Sigma \beta^t(u(S_t) - D(E_t, \theta_t)Y_t)  
utility = sum(  
\beta^*t * (social_utility(state) - damage_function(state))  
for (t, state) in enumerate(path)  
)  
return utility end
```

4. Learning Implementation

1. Damage Function Learning

function damage_function_learning(θ_prev, E_t) # Update damage parameter distribution # Narrow uncertainty based on emissions # Return updated parameters end

2. Agent Expectations

```
function form_expectations(policy_announcement, state)
    # Form expectations about future tax rates
    # Based on policy maker's history and current state
end
```

5. Simulation Framework

1. Single Period

- i. Start with state S_t
- ii. Policy maker chooses [τ_t, τ_{t+1}]
- iii. Agents form expectations
- iv. GE model computes equilibrium
- v. Update state (emissions, damages, learning)
- vi. Transition to S_{t+1}

2. Monte Carlo Analysis

- Run multiple simulations
- Collect feasible paths
- Analyze policy effectiveness
- Identify robust strategies

6. Analysis Tools

1. Path Analysis

- Policy frequency analysis
- Transition path statistics
- Damage learning visualization

2. Economic Outcomes

- Emissions trajectories
- Output and welfare metrics
- · Uncertainty reduction tracking