

Dynamic Modeling for Optimal Cryptoeconomic Policies

Pitch Proposal

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Motivation

- Cryptoeconomies are highly dynamic
 - the nature of blockchain is decentralized
 - many sources of exogenous shocks
- But most crypto monetary policies today are static:
 - deterministic token supply e.g. Bitcoin
 - fixed burn rate e.g. Ethereum
 - naive dynamic burning e.g. Binance Coin
- Problems occur in the long term
 - larger shocks from RBCs
 - when issuance schedule ends

MA Thesis Pitch Proposal

└ Introduction

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shocks from the real economy: Banks failing, stock market, interest rates
shocks from other cryptoeconomies: Luna-Terra, FTX/Alameda

The crypto monetary system

Core Mechanisms:

- ① Minting (increases token supply)
 - initial coin offering (ICO)
 - block rewards via staking or mining (happens constantly)

⇒ Most cryptoeconomies are inflationary by default
- ② Burning (decreases token supply)
 - buy back and burn (BBB)
 - partial burning of transaction fees

Research Question

How should we design cryptoeconomic policies (staking and burning) that respond dynamically to endogenous and exogenous shocks?

- Cryptoeconomy: blockchain/cryptocurrency-based economy e.g. the Ethereum blockchain
- Similar to how central banks implement fiscal and monetary policies e.g. interest rate, taxation, money supply
- Policy goal: maximize welfare for good actors, disincentivize malicious actors

Literature Review

- Micro foundations
 - [Nisan et al., 2007], etc. Algorithmic game theory
 - [Gans and Holden, 2022], etc. Mechanism design
 - [Budish, 2018, Gans and Gandal, 2019] Costs of securing blockchain
- Models of the Cryptoeconomy
 - [Yermack, 2015] Bitcoin is not a currency
 - [Cong et al., 2021, Cong et al., 2022] Asset pricing model featuring token issuance as means of platform financing and user growth
 - [Athey et al., 2016, Catalini and Gans, 2020] Partial equilibrium
 - [Schilling and Uhlig, 2019a, Schilling and Uhlig, 2019b] Crypto vs. Fiat as mediums of exchange with transaction fees
 - [Catalini et al., 2020] Demand and supply of PoS tokens under attack
 - [Biais et al., 2020] An OLG model of prices and transactional benefits
 - [Bolt and Van Oordt, 2020] Modeled crypto exchange rates with quantity theory of money

Contributions to the Literature

Gaps in the literature:

- Most cryptoeconomic models feature a fixed token supply schedule
- Traditional models for optimal fiscal policy (e.g. Ramsey) and monetary policy (e.g. NK) are not directly transferrable to cryptoeconomies
- No literature has been established on optimal economic policies for cryptocurrencies in general (some on CBDC and stablecoins)

Novelty of this research:

- Applies dynamic general equilibrium methods to model and optimize crypto policies for staking and burning

The baseline model (Biais et al. 2020, Journal of Finance)

Idea: The fundamental value of crypto comes from the stream of future transactional benefits [Tirole, 1985]

e.g. access to unique goods, not expropriated/taxed/constrained by government, direct internet access

- Setup: Overlapping generations; discrete time; exogenous endowment
- Three actors: users, validators (miners), hackers
- Three financial assets: a risk-free asset, a standard currency (dollar), a cryptocurrency (Bitcoin)
- Sources of shocks: endowment, transactional benefit, fees

The baseline model: Users

Young users: $c_t^y = e_t - s_t - (1 + \varphi_t)q_t p_t - \hat{q}_t \hat{p}_t$

Old users: $c_{t+1}^o = s_t(1 + r_t) + (1 - h_{t+1})q_t p_{t+1} + \hat{q}_t \hat{p}_{t+1}$
 $+ \theta_{t+1} q_t p_{t+1}$

e_t : endowment

s_t : quantity of risk-free assets held

q_t, \hat{q}_t : quantities of crypto and dollars held

p_t, \hat{p}_t : prices (in units of consumption goods) of crypto and dollars

h_{t+1} : portion of crypto hacked by hackers

φ_t : transaction fees involved in using crypto (exog.)

θ_{t+1} : transactional benefits from using crypto (exog.) (assume $\theta_{t+1} \geq 1$)

The baseline model: Validators and hackers

$$\text{Validators: } c_{t+1}^v = (X_{t+1} - X_t)p_{t+1} + \varphi_{t+1}q_{t+1}p_{t+1}$$

$$\text{Hackers: } c_{t+1}^h = h_{t+1}q_t p_{t+1}$$

X_t : stock token supply

$X_{t+1} - X_t$: increase in token supply (newly minted tokens)

The baseline model: Market clearing

Markets for financial assets:

$$\text{crypto: } q_t = X_t$$

$$\text{dollars: } \hat{q}_t = m$$

$$\text{risk-free assets: } s_t = 0$$

Market for consumption goods (by Walras's Law):

$$c_t^y + c_t^o + c_t^v + c_t^h = e_t$$

The baseline model: Solution

A young user in period t solves:

$$\begin{aligned} \max_{s_t, q_t, \hat{q}_t} & u(c_t^y) + \beta \mathbb{E}_t u(c_{t+1}^o) \\ \text{s.t. } & c_t^y \geq 0 \end{aligned}$$

* Information set at period t includes $\{\theta_t, \varphi_t, \pi_t\}$

From FOCs obtain the equilibrium pricing equation:

$$p_t = \underbrace{\frac{1}{1+r_t}}_{\text{discount}} \mathbb{E}_t \left[\underbrace{\frac{u'(c_{t+1}^o)}{\mathbb{E}_t[u'(c_{t+1}^o)]}}_{\text{risk-neutral prob}} \underbrace{(1-h_{t+1})}_{\text{hack risk}} \underbrace{\frac{1+\theta_{t+1}}{1+\varphi_t}}_{\text{net transact. benefits}} p_{t+1} \right] \quad (1)$$

Modifications to the baseline model

- 1 Introduce two-part transaction fee with burning
- 2 Introduce an interest-bearing staking mechanism
- 3 Modify validators' budget constraint: incorporates costs of validation under PoS and PoW [Biais et al., 2019, Saleh, 2019, Saleh, 2021]
- 4 Use vector autoregression models for exogenous process in variable transaction fees and transaction benefit (e.g. sign restrictions)
- 5 Improvement in model calibration (estimation of transactional benefits and fees)

Proposed Project Timeline

- Spring '23: As final project for ECMA33603 (Macro & Financial Frictions), replicate the baseline model and implement modification 1 and 2 (endogenous money supply)
- Summer '23: More model modifications and seek analytical solutions; gather ideas from industry internship in Ethereum research & dev
- Fall '23 - Winter '24: Data collection, model estimation
- Spring '24: Finish paper

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