# A Luna-tic Stablecoin Crash

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March 2023

# Background

- After remaining close to 1 US Dollar, the algorithmic stablecoin Terra UST lost more than 75 percent of its value in May, 2022, leading to a price collapse of the underlying LUNA token of 99.9 percent, an increase in LUNA supply by a factor of 19,000 and the erasure of more than 50 Billion U.S. Dollar.
- The system worked by allowing traders to convert a Terra UST coin into 1 U.S. Dollar worth of LUNA tokens and vice versa.
- Eventually, a sustained outflow or "burning" of UST coins into LUNA tokens resulted in the collapse of the system.
- The cryptocurrency market at large was affected. Barron's calculates that 600 Billion US Dollars were wiped out.
- Key figure cited in White House proposal on crypto regulation.
- Luna crash and recent FTX crash important driver of MiCA:
  - "Crypto assets & DeFi have the potential to pose real risk to financial stability", Lagarde June 20
  - "ECB president reiterates calls for 'MiCA II' in response to FTX collapse", Nov 28

## The gradual unfolding of the crash



# Objectives

Seek to understand these events qualitatively and quantitatively.

- Theory: Luna.
  - Build a theory that generates a gradual unfolding of the LUNA crash, given a rate of UST Terra burning.
  - Provide closed-form solutions in a benchmark case.
  - Compare qualitatively to the data.
- 2 Theory: Terra UST.
  - Build a theory that justifies the rate of UST Terra burning.
  - Derive bounds for Terra UST underpricing.
- Interpret the data.
  - Use the theory to back out the the theory-variables from the data to exactly justify the observations.

### Method of Quantitative Interpretation

Compare to

- Identification
- Estimation
- Calibration

# **Pricing Cryptocurrencies**

- NPV = 0? This is no different from fiat currencies.
- Benchmark: competing fiat currencies.
- "Exchange Rate Indeterminacy" (Kareken-Wallace, 1981):
  - Kareken Wallace, 1981: In an OLG model with two perfectly substitutable monies, their exchange rate Q<sub>t</sub> (or price of currency B in terms of currency A) is indeterminate and constant forever,

$$Q_t = Q_{t+1}$$

Manuelli-Peck, 1990: In a stochastic OLG model with two perfectly substitutable monies, the exchange rate Q<sub>t</sub> satisfies

$$Q_t = E_t \left[ \frac{v'(c_{2,t+1})}{u'(c_{1,t})} \frac{P_t}{P_{t+1}} Q_{t+1} \right]$$

where  $U_t = u(c_{1,t}) + E_t[v(c_{2,t+1})]$  and  $P_t$  is the price of date-*t* goods in terms of currency A.

► Schilling-Uhlig, 2019: Q<sub>t</sub> is a risk-adjusted martingale,

$$Q_{t} = \mathbb{E}_{t} \left[ \mathcal{M}_{t+1} Q_{t+1} \right], \text{ where } \mathcal{M}_{t+1} = \frac{u'(c_{t+1}) \frac{P_{t}}{P_{t+1}}}{\mathbb{E}_{t} \left[ u'(c_{t+1}) \frac{P_{t}}{P_{t+1}} \right]}$$

# Challenges

- With KW, MP & SU:, shouldn't this have worked?
  - KW, MP & SU: Market Price  $Q_t$  of Luna is a martingale.
  - ► Assume: surprise movements in Q<sub>t</sub> are uncorrelated with surprises regarding Terra UST inflows & outflows.
  - Result: the burning of Terra UST should have been absorbed by a higher market cap of Luna rather than its price collapse.
- One can fix that with **assuming correlations** of the needed kind. But why should they be there? Free parameter.
- Different perspective: (1) assumption for the evolution of market capitalization  $m_t = Q_t M_t$ .
- Tension with KW, MP & SU: given  $m_t$  and  $M_t$ , calculate  $Q_t$ .
- Imperfect substitutability of (crypto-)currencies? Terra UST as raison d'être for Luna?
- And how come, the **crash unfolded gradually**? Simple model & rational expectations: crash should be immediate.
- Solution is via (2) assumption: continual hope for resurrection.

# 1. Theory: Luna. An intro model

- Two periods: before burning, after burning.
- Initial Luna stock: M.
- Price: Q. Rational traders: same price in both periods.
- Burning of *b* UST.
- Final Luna stock:

$$\tilde{M} = M + \frac{b}{Q}$$

or

$$Q\tilde{M} = QM + b$$

- (1) Assumption for final market cap  $\tilde{m} = Q\tilde{M}$ .
- Solve for Q:

$$Q = \frac{\tilde{m} - b}{M}$$

- Need:  $\tilde{m} > b$  for Q > 0. Note: it may be: m < b.
- Next: dynamics, gradual crash ...

# Gradual crash. TERRA UST peg: May 2022



### Gradual crash. USDC peg: March 2023

#### **USD Coin to USD Chart** : : ... Price Market Cap Compare with 7D 1M 3M 1Y YTD ALL H LOG 1D V 1.00 -0.98 0.96 0.92 0.90 0.88 5 9 10 11 1:00 AM USD 6 8 14 15 18 Feb

# 1. Theory: Luna. Dynamics.

- At t = 0: "MIT shock", burning  $b_t$  begins.
- Treat  $b_t$  as exogenous for this part.
- At t, and within the next time interval  $\Delta$ ,
  - ▶ burning of  $b_t \Delta$  UST coins into  $b_t \Delta / Q_t$  LUNA tokens.
  - Assumption (2): continual hope for resurrection. With exogenously given probability λ<sub>t</sub>Δ: the burning of UST stops and LUNA has an exogenously given market cap n<sub>t+Δ</sub>.
  - with probability  $1 \lambda_t \Delta$ : the burning of UST continues.
- When LUNA price reaches *ϵ* > 0: suspension of convertibility and exogenously given terminal market cap *m<sub>T</sub>* for LUNA.
- Rational traders price LUNA tokens, taking dilution into account.
- For  $\Delta \rightarrow 0$ : system of ODEs.
- Closed-form solution, when  $b_t \equiv b$ ,  $\lambda_t \equiv \lambda$ ,  $n_t \equiv n$ .

### Timeline



## Equations

Evolution of stock of tokens: forward in time.

$$M_{t+\Delta} = M_t + \frac{b_t \Delta}{Q_t} \tag{1}$$

Pricing equation: backwards in time.

$$Q_t = \frac{n_{t+\Delta}}{M_{t+\Delta}} \lambda_t \Delta + (1 - \lambda_t \Delta) Q_{t+\Delta}$$

#### Solve numerically?

Examine market cap  $m_t = M_t Q_t!$  Backward in time:

$$m_t + b_t \Delta = \lambda_t n_{t+\Delta} \Delta + (1 - \lambda_t \Delta) m_{t+\Delta}$$
(3)

(2)

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<sup>(2)</sup>

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(3)

### ODEs

Define: 
$$y_t = \lambda_t \left( \frac{n_{t+\Delta}}{m_{t+\Delta}} - 1 \right)$$
 (4)  
 $m_{t+\Delta} - m_t = -\lambda_t (n_{t+\Delta} - m_{t+\Delta})\Delta + b_t \Delta = -y_t m_{t+\Delta} \Delta + b_t \Delta$  (5)

$$Q_{t+\Delta} - Q_t = -y_t Q_{t+\Delta} \Delta \tag{6}$$

For  $\Delta \rightarrow 0$ :

$$\dot{m}_t = -\lambda_t (n_t - m_t) + b_t \tag{7}$$

$$\dot{Q}_t = -\lambda_t \left(\frac{n_t}{m_t} - 1\right) Q_t$$
 (8)

$$\dot{M}_t = \frac{b_t}{Q_t} \tag{9}$$

Exogenous:  $b_t, \lambda_t, n_t, m_T$ . T solves  $Q_T = \epsilon$ , given  $\epsilon$ .

Define the relative burn rate

$$\alpha_t = \frac{b_t}{\lambda_t n_t} \tag{10}$$

# A Simulation

Parameters:

 $\begin{array}{ll} \mbox{initial market cap:} & m_t & = 30 \mbox{ for } t < 0 \\ & \mbox{initial price:} & Q_t & = 100 \mbox{ for } t < 0. \mbox{ Thus, } M_t = .3 \mbox{ for } t < 0 \\ & \mbox{exit market cap:} & n & = 30 \\ & \mbox{exit rate:} & \lambda & = 0.05 \\ & \mbox{relative burn rate:} & \alpha = \frac{b}{\lambda n} & = \left\{ \begin{array}{c} 0 & \mbox{for } t < t^* \\ 0.5 \mbox{ for } t \geq t^* \end{array} \right. \\ & \mbox{stopping price:} & Q_T = \epsilon & = 0.1 \\ & \mbox{stopping market cap:} & m_T & = (1 - \kappa)n \mbox{ for the "stop decline rate"} \kappa \end{array}$ 

Three scenarios:

- **1** Burning start  $t^* = 0$ . Stop decline rate  $\kappa = 0$ , i.e.  $m_T = n$ .
- **2** Burning start  $t^* = 50$ . Stop decline rate  $\kappa = 0$ , i.e.  $m_T = n$ .
- Surning start  $t^* = 50$ . Stop decline rate  $\kappa = 0.9$ , i.e.  $m_T = n/10$ .







### Data



### 3. Interpret the data

#### Method of quantitative interpretation:

use data to back out  $n_t$ ,  $\lambda_t$ ,  $m_T$ .

## Calculations

Accounting: 
$$M_{t+\Delta} = M_t + \frac{b_t \Delta}{Q_t^a}$$
 (11)  
Define:  $x_t = -\frac{Q_{t+\Delta} - Q_t}{Q_{t+\Delta}\Delta}$  (12)

Market capitalization is  $m_t = Q_t M_t$  for all t. Thus,

$$m_{t+\Delta} - m_t = -x_t m_{t+\Delta} \Delta + \frac{Q_t}{Q_t^a} b_t \Delta$$
(13)

$$Q_{t+\Delta} - Q_t = -x_t Q_{t+\Delta} \Delta \tag{14}$$

Compare to theory:

Define: 
$$y_t = \lambda_t \left( \frac{n_{t+\Delta}}{m_{t+\Delta}} - 1 \right)$$
 (15)

$$m_{t+\Delta} - m_t = -y_t m_{t+\Delta} \Delta + b_t \Delta$$
(16)

$$Q_{t+\Delta} - Q_t = -y_t Q_{t+\Delta} \Delta \tag{17}$$

Thus, one can infer  $y_t$  from the data, but not  $n_t$  and  $\lambda_t$  separately. Any data can be interpreted as arising from that solved-for  $y_t$ .

### $Data \rightarrow Quantitative Interpretation$



### Luna price: theory vs data



### Burn rate and price floor: theory



### Burn rate and price floor: data





# Distribution of threshold P for burning UST: data.



# Conclusions

- Terra-Luna crash.
- Challenges:
  - Kareken-Wallace etc: why a crash?
  - ② Gradual rather than immediate?
- Theory:
  - Impose assumptions on market capitalization.
  - 2 Impose that traders hope for resurrection, until price hits floor  $\epsilon$ .
  - ODEs. They can be solved.
  - Olosed-form solution in special case.
  - Stylized version of the data.
- Method of quantitative interpretation:
  - Use the data to back out theory variables.
    - Examine the theory variables to understand the crash.
    - Exit probability declined to zero as end drew near.
  - Threshold probability of collapse, when burning UST was above 50% for more than 80% of UST holders.