

# notebook-experiment-liquidation-ratio

February 10, 2021

## 1 Experiment Analysis: The effect of liquidation ratio and price rescaling

Analyse the effect of setting the parameters `rescale_target_price` and `arbitrageur_considers_liquidation_ratio` to false.

- See `experiments/system_model_v3/experiment_liquidation_ratio.py`

## 2 Setup and Dependencies

```
[1]: # Set project root folder, to enable importing project files from subdirectories
from pathlib import Path
import os

path = Path().resolve()
root_path = str(path).split('notebooks')[0]
os.chdir(root_path)

# Force reload of project modules, sometimes necessary for Jupyter kernel
%load_ext autoreload
%autoreload 2

# Display framework versions for easy debugging
%pip show cadCAD
%pip show radcad
```

Name: cadCAD

Version: 0.4.23

Summary: cadCAD: a differential games based simulation software package for research, validation, and Computer Aided Design of economic systems

Home-page: <https://github.com/cadCAD-org/cadCAD>

Author: Joshua E. Jodesty

Author-email: [joshua@block.science](mailto:joshua@block.science)

License: LICENSE.txt

Location: `/home/bscholtz/workspace/reflexer/venv/lib/python3.8/site-packages`

Requires: `fn`, `funcy`, `pandas`, `pathos`

Required-by:

Note: you may need to restart the kernel to use updated packages.  
Name: radcad  
Version: 0.5.4  
Summary: A cadCAD implementation, for dynamical systems modelling & simulation  
Home-page: None  
Author: Benjamin Scholtz  
Author-email: ben@bitsofether.com  
License: None  
Location: /home/bescholtz/workspace/reflexer/venv/lib/python3.8/site-packages  
Requires: pathos, ray, pandas, boto3, tables  
Required-by:  
Note: you may need to restart the kernel to use updated packages.

```
[2]: # Import all shared dependencies and setup
from shared import *

import plotly.express as px
import plotly.graph_objects as go
from plotly.subplots import make_subplots
# import plotly.io as pio
# pio.renderers.default = "png"
from pprint import pprint
```

```
[3]: # Update dataframe display settings
pd.set_option('display.max_columns', 100)
pd.set_option('display.max_rows', 50)
```

### 3 Load Results

Using the experiment logs, select the experiment of interest from the specific HDF5 store file (these datasets are very large, and won't be committed to repo):

```
[4]: # experiment_results = 'experiments/system_model_v3/experiment_controller_sweep/
    ↪experiment_results.hdf5'
experiment_results = 'experiments/system_model_v3/experiment_liquidation_ratio/
    ↪experiment_results.hdf5'
```

```
[5]: experiment_results_keys = []
with pd.HDFStore(experiment_results) as store:
    experiment_results_keys = list(filter(lambda x: "results" in x, store.
    ↪keys()))
    exceptions_keys = list(filter(lambda x: "exceptions" in x, store.keys()))
```

```
[6]: # A list of all experiment result keys
experiment_results_keys
```

```
[6]: ['/results_2021-02-09T22:02:18.259338', '/results_2021-02-09T23:51:42.952840']
```

```
[7]: # A list of all experiment result exception keys
exceptions_keys
```

```
[7]: ['/exceptions_2021-02-09T22:02:18.259338',
      '/exceptions_2021-02-09T23:51:42.952840']
```

```
[8]: # Copy a results_ key from the above keys to select the experiment
experiment_results_key = 'results_2021-02-09T23:51:42.952840' # Or select last_
↳ result: experiment_results_keys[-1]
experiment_timestamp = experiment_results_key.strip('results_')
exceptions_key = 'exceptions_' + experiment_timestamp
experiment_timestamp
```

```
[8]: '2021-02-09T23:51:42.952840'
```

```
[9]: df_raw = pd.read_hdf(experiment_results, experiment_results_key)
df_raw.tail()
```

```
[9]:
```

				cdp_metrics	optimal_values	\
3732619	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...					{}
3732620	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...					{}
3732621	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...					{}
3732622	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...					{}
3732623	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...					{}

	sim_metrics	timedelta	cumulative_time	timestamp	blockheight	\
3732619	{}	3600	5184000	2017-03-02	0	
3732620	{}	3600	5184000	2017-03-02	0	
3732621	{}	3600	5184000	2017-03-02	0	
3732622	{}	3600	5184000	2017-03-02	0	
3732623	{}	3600	5184000	2017-03-02	0	

	eth_price	liquidity_demand	liquidity_demand_mean	\
3732619	300.0	-0.0	0.0	
3732620	300.0	-0.0	0.0	
3732621	300.0	-0.0	0.0	
3732622	300.0	-0.0	0.0	
3732623	300.0	-0.0	0.0	

				cdps	eth_collateral	\
3732619	open	time	locked	drawn	...	69768.4181
3732620	open	time	locked	drawn	...	69768.4181
3732621	open	time	locked	drawn	...	69768.4181
3732622	open	time	locked	drawn	...	69768.4181
3732623	open	time	locked	drawn	...	69768.4181

	eth_locked	eth_freed	eth_bitten	principal_debt	\
--	------------	-----------	------------	----------------	---

3732619	154827.528922	85059.110822	0.0	2.238129e+07
3732620	154827.528922	85059.110822	0.0	2.238129e+07
3732621	154827.528922	85059.110822	0.0	2.238129e+07
3732622	154827.528922	85059.110822	0.0	2.238129e+07
3732623	154827.528922	85059.110822	0.0	2.238129e+07

	rai_drawn	rai_wiped	rai_bitten	accrued_interest	\
3732619	3.769353e+07	1.531225e+07	0.0	16159.356162	
3732620	3.769354e+07	1.531225e+07	0.0	16159.356162	
3732621	3.769354e+07	1.531225e+07	0.0	16159.356162	
3732622	3.769354e+07	1.531225e+07	0.0	16159.356162	
3732623	3.769354e+07	1.531225e+07	0.0	16159.356162	

	interest_dripped	interest_wiped	interest_bitten	w_1	w_2	\
3732619	0	0	0	0.000002	0.0	
3732620	0	0	0	0.000002	0.0	
3732621	0	0	0	0.000002	0.0	
3732622	0	0	0	0.000002	0.0	
3732623	0	0	0	0.000002	0.0	

	w_3	system_revenue	stability_fee	market_price	market_price_twap	\
3732619	0.0	0.0	1.585490e-10	0.64498	0.644982	
3732620	0.0	0.0	1.585490e-10	0.64498	0.644982	
3732621	0.0	0.0	1.585490e-10	0.64498	0.644982	
3732622	0.0	0.0	1.585490e-10	0.64498	0.644982	
3732623	0.0	0.0	1.585490e-10	0.64498	0.644982	

	target_price	target_rate	eth_return	eth_gross_return	\
3732619	0.644951	-5.013736e-11	0.0	1.0	
3732620	0.644951	-5.013736e-11	0.0	1.0	
3732621	0.644951	-5.013736e-11	0.0	1.0	
3732622	0.644951	-5.013736e-11	0.0	1.0	
3732623	0.644951	-5.013736e-11	0.0	1.0	

	expected_market_price	expected_debt_price	error_star	\
3732619	0.66378	3.14	-0.00003	
3732620	0.66378	3.14	-0.00003	
3732621	0.66378	3.14	-0.00003	
3732622	0.66378	3.14	-0.00003	
3732623	0.66378	3.14	-0.00003	

	error_star_integral	market_slippage	RAI_balance	ETH_balance	\
3732619	-5449.108542	3.648923e-07	2.238129e+07	48118.243496	
3732620	-5449.108542	3.648923e-07	2.238129e+07	48118.243496	
3732621	-5449.108542	3.648923e-07	2.238129e+07	48118.243496	
3732622	-5449.108542	3.648923e-07	2.238129e+07	48118.243496	
3732623	-5449.108542	3.648923e-07	2.238129e+07	48118.243496	

```

UNI_supply                                uniswap_oracle \
3732619 10000000.0 <models.system_model_v3.model.parts.uniswap_or...
3732620 10000000.0 <models.system_model_v3.model.parts.uniswap_or...
3732621 10000000.0 <models.system_model_v3.model.parts.uniswap_or...
3732622 10000000.0 <models.system_model_v3.model.parts.uniswap_or...
3732623 10000000.0 <models.system_model_v3.model.parts.uniswap_or...

```

```

simulation subset run substep timestep events
3732619      0    35   4      14    1440    NaN
3732620      0    35   4      15    1440    NaN
3732621      0    35   4      16    1440    NaN
3732622      0    35   4      17    1440    NaN
3732623      0    35   4      18    1440    NaN

```

Get experiment exceptions, tracebacks, and simulation metadata for further analysis:

```
[10]: exceptions_df = pd.read_hdf(experiment_results, exceptions_key)
exceptions_df.head()
```

```
[10]: exception traceback simulation run subset timesteps \
0      None      None          0    0      0      1440
1      None      None          0    0      1      1440
2      None      None          0    0      2      1440
3      None      None          0    0      3      1440
4      None      None          0    0      4      1440

```

```

parameters \
0 {"0":{"debug":false,"raise_on_assert":true,"fr...
1 {"0":{"debug":false,"raise_on_assert":true,"fr...
2 {"0":{"debug":false,"raise_on_assert":true,"fr...
3 {"0":{"debug":false,"raise_on_assert":true,"fr...
4 {"0":{"debug":false,"raise_on_assert":true,"fr...

```

```

initial_state
0 {'cdp_metrics': {}, 'optimal_values': {}, 'sim...
1 {'cdp_metrics': {}, 'optimal_values': {}, 'sim...
2 {'cdp_metrics': {}, 'optimal_values': {}, 'sim...
3 {'cdp_metrics': {}, 'optimal_values': {}, 'sim...
4 {'cdp_metrics': {}, 'optimal_values': {}, 'sim...

```

```
[11]: # Print the first 5 exceptions - indicating failed simulations
pprint(list(exceptions_df['exception'])[:5])
```

```
[None, None, None, None, None]
```

## 4 Post Process Results

```
[12]: from experiments.system_model_v3.post_process import post_process_results
      from experiments.system_model_v3.experiment_liquidation_ratio import params, \
      ↪SIMULATION_TIMESTEPS
```

```
* Number of timesteps: 1440 / 60.0 days
* Number of MC runs: 4
* Timestep duration: 0.004 seconds
* Control parameters: ['kp', 'ki', 'rescale_target_price',
'arbitrageur_considers_liquidation_ratio']
* Number of parameter combinations: 36
* Expected experiment duration: 13.824000000000002 minutes / 0.23040000000000002
hours
```

Remove substeps, add set\_params to dataframe, and add post-processing columns:

```
[13]: df = post_process_results(df_raw, params, set_params=['ki', 'kp', \
      ↪'liquidation_ratio', 'rescale_target_price', \
      ↪'arbitrageur_considers_liquidation_ratio'])
      df
```

```
[13]:
```

	index	cdp_metrics \
0	0	{}
1	18	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
2	36	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
3	54	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
4	72	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
...	...	...
207499	3732551	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
207500	3732569	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
207501	3732587	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
207502	3732605	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
207503	3732623	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...

	optimal_values	sim_metrics	timedelta	cumulative_time \
0	{}	{}	0	0
1	{}	{}	3600	3600
2	{}	{}	3600	7200
3	{}	{}	3600	10800
4	{}	{}	3600	14400
...	...	...	...	...
207499	{}	{}	3600	5169600
207500	{}	{}	3600	5173200
207501	{}	{}	3600	5176800
207502	{}	{}	3600	5180400

207503	{}	{}	3600	5184000
--------	----	----	------	---------

	timestamp	blockheight	eth_price	liquidity_demand	\
0	2017-01-01 00:00:00	0	294.069151	1.0	
1	2017-01-01 01:00:00	0	300.000000	0.0	
2	2017-01-01 02:00:00	0	300.000000	-0.0	
3	2017-01-01 03:00:00	0	300.000000	0.0	
4	2017-01-01 04:00:00	0	300.000000	-0.0	
...	...	...	...	...	
207499	2017-03-01 20:00:00	0	300.000000	0.0	
207500	2017-03-01 21:00:00	0	300.000000	0.0	
207501	2017-03-01 22:00:00	0	300.000000	-0.0	
207502	2017-03-01 23:00:00	0	300.000000	0.0	
207503	2017-03-02 00:00:00	0	300.000000	-0.0	

	liquidity_demand_mean	\
0	1.0000	
1	0.5000	
2	0.2500	
3	0.1250	
4	0.0625	
...	...	
207499	0.0000	
207500	0.0000	
207501	0.0000	
207502	0.0000	
207503	0.0000	

	cdps	eth_collateral	\
0	None	154827.528922	
1	None	154827.528922	
2	None	105925.122117	
3	None	105925.122117	
4	None	105925.122117	
...	...	...	
207499	None	69768.418100	
207500	None	69768.418100	
207501	None	69768.418100	
207502	open time	locked	drawn ... 69768.418100
207503	open time	locked	drawn ... 69768.418100

	eth_locked	eth_freed	eth_bitten	principal_debt	rai_drawn	\
0	154827.528922	0.000000	0.0	1.000000e+07	1.000000e+07	
1	154827.528922	0.000000	0.0	1.450000e+07	1.450000e+07	
2	154827.528922	48902.406805	0.0	1.012023e+07	1.450000e+07	
3	154827.528922	48902.406805	0.0	1.012023e+07	1.450000e+07	
4	154827.528922	48902.406805	0.0	1.012023e+07	1.450000e+07	

...	...	...	...	...	...	...
207499	154827.528922	85059.110822	0.0	2.238127e+07	3.769352e+07	
207500	154827.528922	85059.110822	0.0	2.238128e+07	3.769352e+07	
207501	154827.528922	85059.110822	0.0	2.238128e+07	3.769353e+07	
207502	154827.528922	85059.110822	0.0	2.238129e+07	3.769353e+07	
207503	154827.528922	85059.110822	0.0	2.238129e+07	3.769354e+07	

	rai_wiped	rai_bitten	accrued_interest	interest_dripped	\
0	0.000000e+00	0.0	0.000000		0
1	0.000000e+00	0.0	5.707764		0
2	4.379765e+06	0.0	11.484158		0
3	4.379765e+06	0.0	17.260556		0
4	4.379765e+06	0.0	23.036956		0
...	...	...	...	...	
207499	1.531225e+07	0.0	16108.220517		0
207500	1.531225e+07	0.0	16121.004414		0
207501	1.531225e+07	0.0	16133.788320		0
207502	1.531225e+07	0.0	16146.572236		0
207503	1.531225e+07	0.0	16159.356162		0

	interest_wiped	interest_bitten	w_1	w_2	w_3	\
0	0	0	0.000000e+00	0.0	0.0	
1	0	0	8.276258e+00	0.0	0.0	
2	0	0	4.723892e-06	0.0	0.0	
3	0	0	2.696510e-12	0.0	0.0	
4	0	0	0.000000e+00	0.0	0.0	
...	...	...	...	...	...	
207499	0	0	2.363668e-06	0.0	0.0	
207500	0	0	2.363669e-06	0.0	0.0	
207501	0	0	2.334199e-06	0.0	0.0	
207502	0	0	2.334199e-06	0.0	0.0	
207503	0	0	2.334200e-06	0.0	0.0	

	system_revenue	stability_fee	market_price	market_price_twap	\
0	0.0	1.585490e-10	3.140000	0.000000	
1	0.0	1.585490e-10	3.140000	0.000000	
2	0.0	1.585490e-10	1.525000	0.000000	
3	0.0	1.585490e-10	3.133425	0.000000	
4	0.0	1.585490e-10	3.133425	0.000000	
...	...	...	...	...	
207499	0.0	1.585490e-10	0.644981	0.644982	
207500	0.0	1.585490e-10	0.644980	0.644982	
207501	0.0	1.585490e-10	0.644980	0.644982	
207502	0.0	1.585490e-10	0.644980	0.644982	
207503	0.0	1.585490e-10	0.644980	0.644982	

target_price	target_rate	eth_return	eth_gross_return	\
--------------	-------------	------------	------------------	---



0	3.140000	0.000000e+00	0.000000	0.0
1	2.165517	0.000000e+00	0.020168	1.0
2	2.165517	0.000000e+00	0.000000	1.0
3	2.165517	0.000000e+00	0.000000	1.0
4	2.165517	0.000000e+00	0.000000	1.0
...	...	...	...	...
207499	0.644952	-5.139649e-11	0.000000	1.0
207500	0.644952	-5.075558e-11	0.000000	1.0
207501	0.644952	-5.075558e-11	0.000000	1.0
207502	0.644951	-5.075558e-11	0.000000	1.0
207503	0.644951	-5.013736e-11	0.000000	1.0

	expected_market_price	expected_debt_price	error_star	\
0	3.140000	3.14	0.000000	
1	3.231683	3.14	3.140000	
2	1.569093	3.14	3.140000	
3	3.224303	3.14	3.140000	
4	3.224444	3.14	3.140000	
...	...	...	...	
207499	0.663780	3.14	-0.000030	
207500	0.663780	3.14	-0.000031	
207501	0.663780	3.14	-0.000030	
207502	0.663780	3.14	-0.000030	
207503	0.663780	3.14	-0.000030	

	error_star_integral	market_slippage	RAI_balance	ETH_balance	\
0	0.000000	0.000000e+00	1.000000e+07	106777.606153	
1	5652.000000	0.000000e+00	1.450000e+07	73708.353400	
2	16932.000000	5.143311e-01	1.012023e+07	105703.331685	
3	28166.000000	-1.054705e+00	1.012023e+07	105703.331685	
4	39354.000000	0.000000e+00	1.012023e+07	105703.331685	
...	...	...	...	...	
207499	-5541.109474	3.694992e-07	2.238127e+07	48118.278670	
207500	-5518.109904	3.694992e-07	2.238128e+07	48118.269794	
207501	-5495.109012	3.694992e-07	2.238128e+07	48118.261028	
207502	-5472.108117	3.648923e-07	2.238129e+07	48118.252262	
207503	-5449.108542	3.648923e-07	2.238129e+07	48118.243496	

	UNI_supply	uniswap_oracle	\
0	10000000.0	None	
1	10000000.0	None	
2	10000000.0	None	
3	10000000.0	None	
4	10000000.0	None	
...	...	...	
207499	10000000.0	None	
207500	10000000.0	None	

207501	10000000.0	None
207502	10000000.0	<models.system_model_v3.model.parts.uniswap_or...
207503	10000000.0	<models.system_model_v3.model.parts.uniswap_or...

	simulation	subset	run	substep	timestep	events	\
0	0	0	1	0	0	NaN	
1	0	0	1	18	1	NaN	
2	0	0	1	18	2	NaN	
3	0	0	1	18	3	NaN	
4	0	0	1	18	4	NaN	
...	...	...	...	...	...	...	
207499	0	35	4	18	1436	NaN	
207500	0	35	4	18	1437	NaN	
207501	0	35	4	18	1438	NaN	
207502	0	35	4	18	1439	NaN	
207503	0	35	4	18	1440	NaN	

	eth_collateral_value	collateralization_ratio	ki	\
0	4.553000e+07	1.450000	-5.000000e-09	
1	4.644826e+07	1.479244	-5.000000e-09	
2	3.177754e+07	1.450000	-5.000000e-09	
3	3.177754e+07	1.450000	-5.000000e-09	
4	3.177754e+07	1.450000	-5.000000e-09	
...	...	...	...	
207499	2.093053e+07	1.450000	-2.000000e-10	
207500	2.093053e+07	1.450000	-2.000000e-10	
207501	2.093053e+07	1.450000	-2.000000e-10	
207502	2.093053e+07	1.450000	-2.000000e-10	
207503	2.093053e+07	1.450000	-2.000000e-10	

	kp	liquidation_ratio	rescale_target_price	\
0	2.000000e-07	1.45	True	
1	2.000000e-07	1.45	True	
2	2.000000e-07	1.45	True	
3	2.000000e-07	1.45	True	
4	2.000000e-07	1.45	True	
...	...	...	...	
207499	5.000000e-06	1.45	False	
207500	5.000000e-06	1.45	False	
207501	5.000000e-06	1.45	False	
207502	5.000000e-06	1.45	False	
207503	5.000000e-06	1.45	False	

	arbitrageur_considers_liquidation_ratio	target_price_scaled
0	True	4.553000
1	True	3.140000
2	True	3.140000

3	True	3.140000
4	True	3.140000
...	...	...
207499	False	0.935180
207500	False	0.935180
207501	False	0.935180
207502	False	0.935180
207503	False	0.935179

[207504 rows x 58 columns]

```
[14]: %%capture
# Save the processed results to the same HDF5 store file
df.to_hdf(experiment_results, key=f'processed_results_{experiment_timestamp}')
```

## 5 Control Parameters

```
[15]: from radcad.core import generate_parameter_sweep

param_sweep = generate_parameter_sweep(params)
```

```
[16]: df_control_parameters = df[['subset', 'kp', 'ki']]

df_control_parameters = df_control_parameters.drop_duplicates(subset=['kp',
↪ 'ki'])
df_control_parameters
```

```
[16]:
```

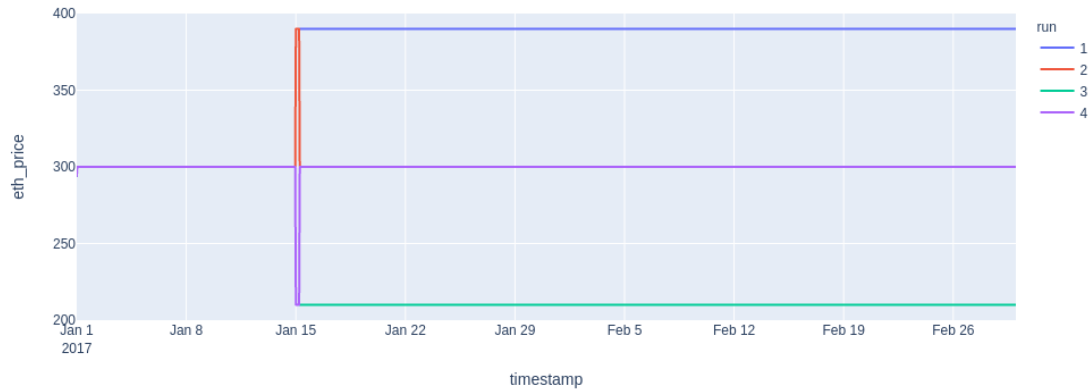
	subset	kp	ki
0	0	2.000000e-07	-5.000000e-09
5764	4	2.000000e-07	-1.000000e-09
11528	8	2.000000e-07	-2.000000e-10
17292	12	1.000000e-06	-5.000000e-09
23056	16	1.000000e-06	-1.000000e-09
28820	20	1.000000e-06	-2.000000e-10
34584	24	5.000000e-06	-5.000000e-09
40348	28	5.000000e-06	-1.000000e-09
46112	32	5.000000e-06	-2.000000e-10

## 6 Simulation Analysis

```
[17]: df.query('subset == 0')[['timestamp', 'eth_price', 'run']].plot(
    title="ETH price shocks (positive and negative step and impulse; one shock_
↪ type for each run)",
    x='timestamp',
    y='eth_price',
```

```
color='run'
)
```

ETH price shocks (positive and negative step and impulse; one shock type for each run)



Experiment parameter subsets:

```
[53]: df[['subset', 'kp', 'ki', 'rescale_target_price',
        ↪ 'arbitrageur_considers_liquidation_ratio']] \
        .drop_duplicates(subset=['kp', 'ki', 'rescale_target_price',
        ↪ 'arbitrageur_considers_liquidation_ratio'])
```

```
[53]:
```

	subset	kp	ki	rescale_target_price	\
0	0	2.000000e-07	-5.000000e-09	True	
1441	1	2.000000e-07	-5.000000e-09	True	
2882	2	2.000000e-07	-5.000000e-09	False	
4323	3	2.000000e-07	-5.000000e-09	False	
5764	4	2.000000e-07	-1.000000e-09	True	
7205	5	2.000000e-07	-1.000000e-09	True	
8646	6	2.000000e-07	-1.000000e-09	False	
10087	7	2.000000e-07	-1.000000e-09	False	
11528	8	2.000000e-07	-2.000000e-10	True	
12969	9	2.000000e-07	-2.000000e-10	True	
14410	10	2.000000e-07	-2.000000e-10	False	
15851	11	2.000000e-07	-2.000000e-10	False	
17292	12	1.000000e-06	-5.000000e-09	True	
18733	13	1.000000e-06	-5.000000e-09	True	
20174	14	1.000000e-06	-5.000000e-09	False	
21615	15	1.000000e-06	-5.000000e-09	False	
23056	16	1.000000e-06	-1.000000e-09	True	
24497	17	1.000000e-06	-1.000000e-09	True	
25938	18	1.000000e-06	-1.000000e-09	False	

27379	19	1.000000e-06	-1.000000e-09	False
28820	20	1.000000e-06	-2.000000e-10	True
30261	21	1.000000e-06	-2.000000e-10	True
31702	22	1.000000e-06	-2.000000e-10	False
33143	23	1.000000e-06	-2.000000e-10	False
34584	24	5.000000e-06	-5.000000e-09	True
36025	25	5.000000e-06	-5.000000e-09	True
37466	26	5.000000e-06	-5.000000e-09	False
38907	27	5.000000e-06	-5.000000e-09	False
40348	28	5.000000e-06	-1.000000e-09	True
41789	29	5.000000e-06	-1.000000e-09	True
43230	30	5.000000e-06	-1.000000e-09	False
44671	31	5.000000e-06	-1.000000e-09	False
46112	32	5.000000e-06	-2.000000e-10	True
47553	33	5.000000e-06	-2.000000e-10	True
48994	34	5.000000e-06	-2.000000e-10	False
50435	35	5.000000e-06	-2.000000e-10	False

	arbitrageur_considers_liquidation_ratio
0	True
1441	False
2882	True
4323	False
5764	True
7205	False
8646	True
10087	False
11528	True
12969	False
14410	True
15851	False
17292	True
18733	False
20174	True
21615	False
23056	True
24497	False
25938	True
27379	False
28820	True
30261	False
31702	True
33143	False
34584	True
36025	False
37466	True
38907	False

40348	True
41789	False
43230	True
44671	False
46112	True
47553	False
48994	True
50435	False

```
[52]: fig = px.line(  
    df.query('run == 1'),  
    title="Price response for all control parameter subsets, first run",  
    x="timestamp",  
    y=["market_price", "market_price_twap", "target_price"],  
    facet_col="subset",  
    facet_col_wrap=4,  
    facet_row_spacing=0.04, # default is 0.07 when facet_col_wrap is used  
    facet_col_spacing=0.04, # default is 0.03  
    height=2000  
)  
fig.show()
```



Get the initial target price to test stability conditions:

```
[19]: initial_target_price = df['target_price'].iloc[0]
      initial_target_price
```

[19]: 3.14

Find all controller constant subsets where the price goes to zero:

```
[20]: df_market_price_zero = df.query("market_price <= 0.1*@initial_target_price")
df_market_price_zero[['subset', 'kp', 'ki']].drop_duplicates(subset=['kp', 'ki'])
```

```
[20]:      subset      kp      ki
38192      26  0.000005 -5.000000e-09
43864      30  0.000005 -1.000000e-09
49612      34  0.000005 -2.000000e-10
```

Find all controller constant subsets where the price goes to infinity:

```
[21]: df_market_price_infinity = df.query("market_price > 10*@initial_target_price")
df_market_price_infinity[['subset', 'kp', 'ki']].drop_duplicates(subset=['kp', 'ki'])
```

```
[21]:      subset      kp      ki
4205      2  2.000000e-07 -5.000000e-09
```

Create dataframe of stable simulation scenarios.

Stability is defined as: 1. The market price and scaled target price remaining within 0.1x and 10x the starting price, for all timesteps

```
[22]: df['stable_price'] = False
df.loc[df.eval("""
0.1*@initial_target_price < market_price <= 10*@initial_target_price and 0.
↳1*@initial_target_price < target_price_scaled <= 10*@initial_target_price
"""), 'stable_price'] = True
df_stable_price = df.groupby("subset").filter(lambda x: all(x.query('timestep >=
↳24*2')['stable_price'])) # and x['timestep'].max() == SIMULATION_TIMESTEPS
df_stable_price['subset'].unique()
```

```
[22]: array([ 0,  1,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15, 16, 17, 18,
        19, 20, 21, 22, 23, 24, 25, 27, 28, 29, 31, 32, 33, 35])
```

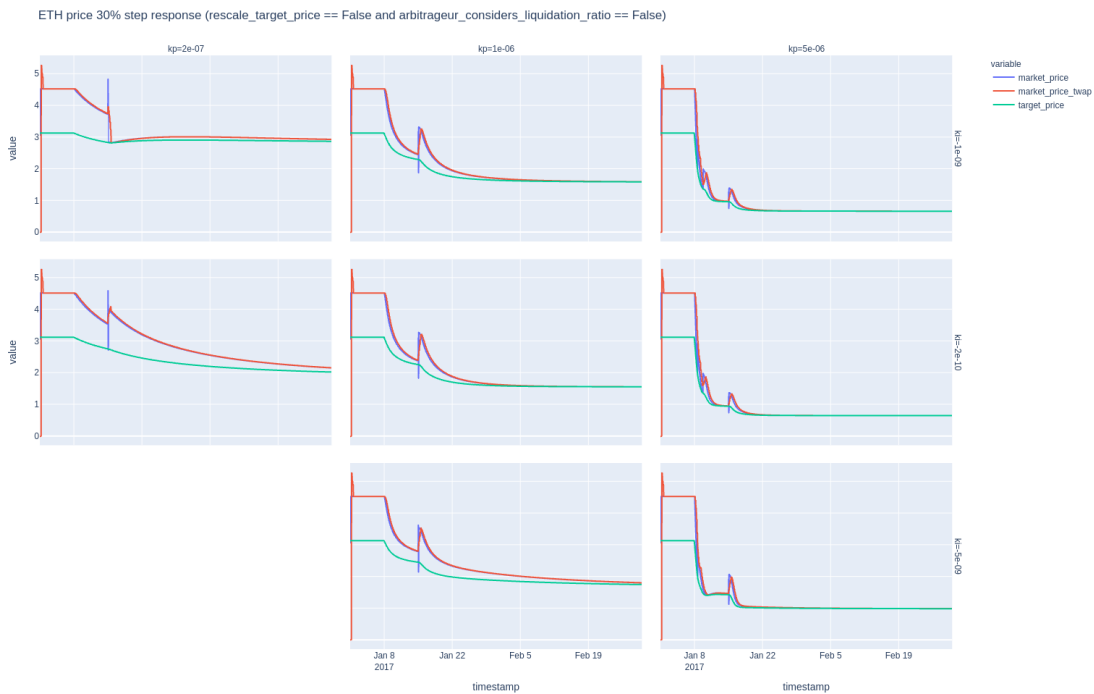
```
[35]: fig = px.line(
    df_stable_price.query('run == 1 and rescale_target_price == False and
↳arbitrageur_considers_liquidation_ratio == False'),
    title="ETH price 30% step response (rescale_target_price == False and
↳arbitrageur_considers_liquidation_ratio == False)",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price"],
    facet_col="kp",
    facet_row="ki",
```



```

        facet_col_wrap=2,
        height=1000
    )
    fig.show()

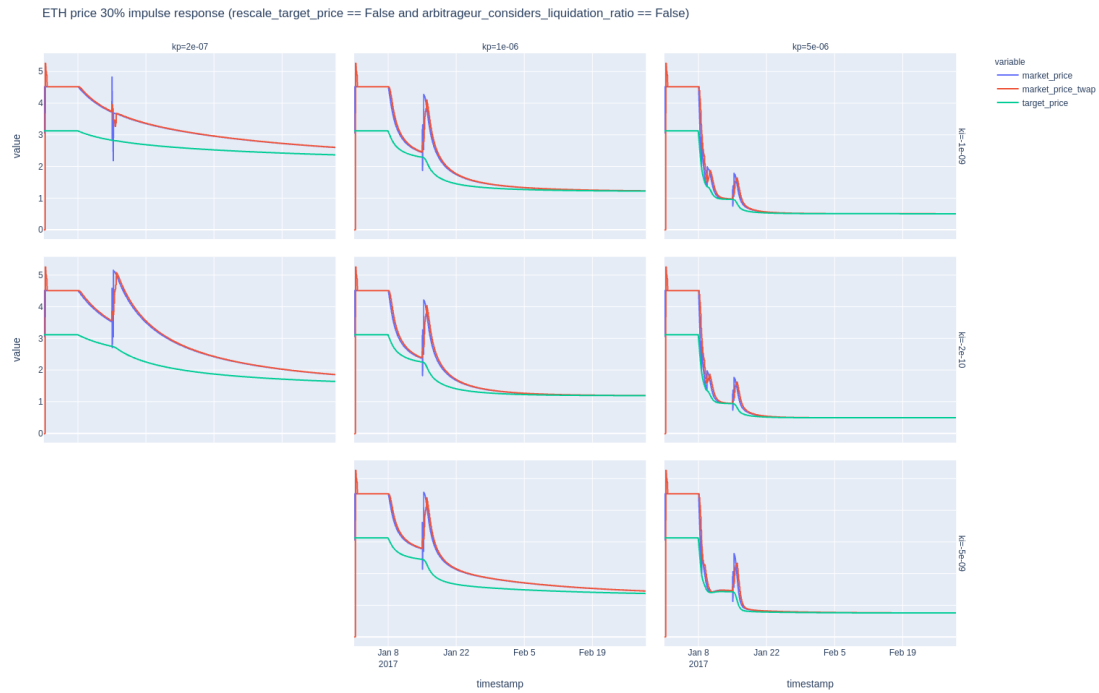
```



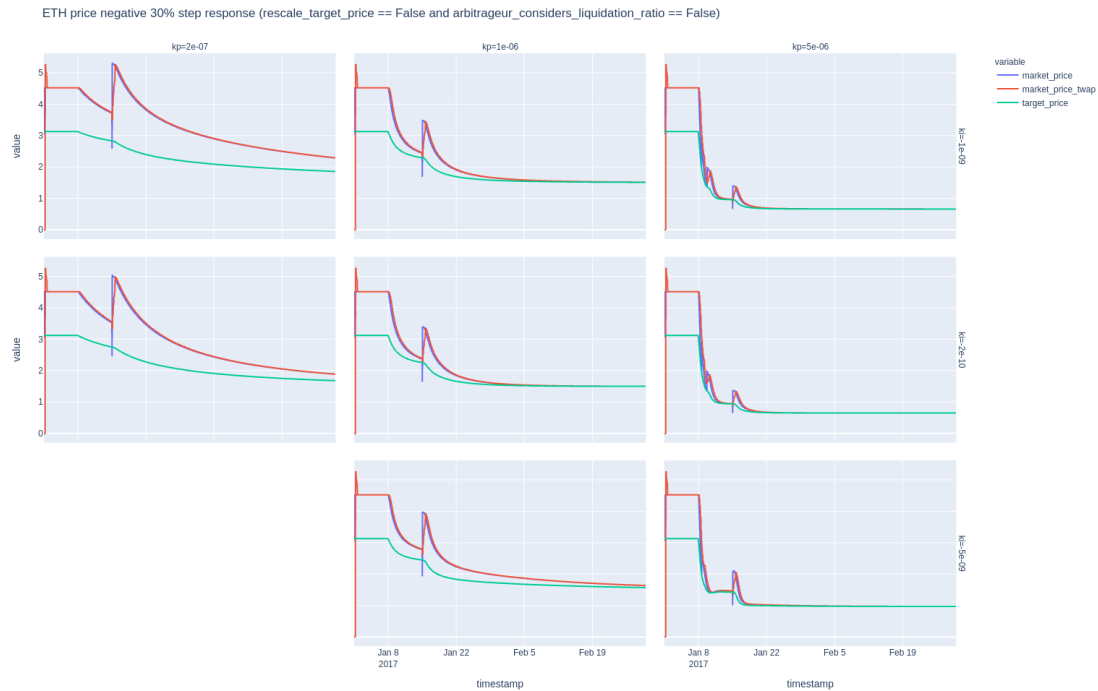
```

[36]: fig = px.line(
    df_stable_price.query('run == 2 and rescale_target_price == False and
    ↳ arbitrageur_considers_liquidation_ratio == False'),
    title="ETH price 30% impulse response (rescale_target_price == False and
    ↳ arbitrageur_considers_liquidation_ratio == False)",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
fig.show()

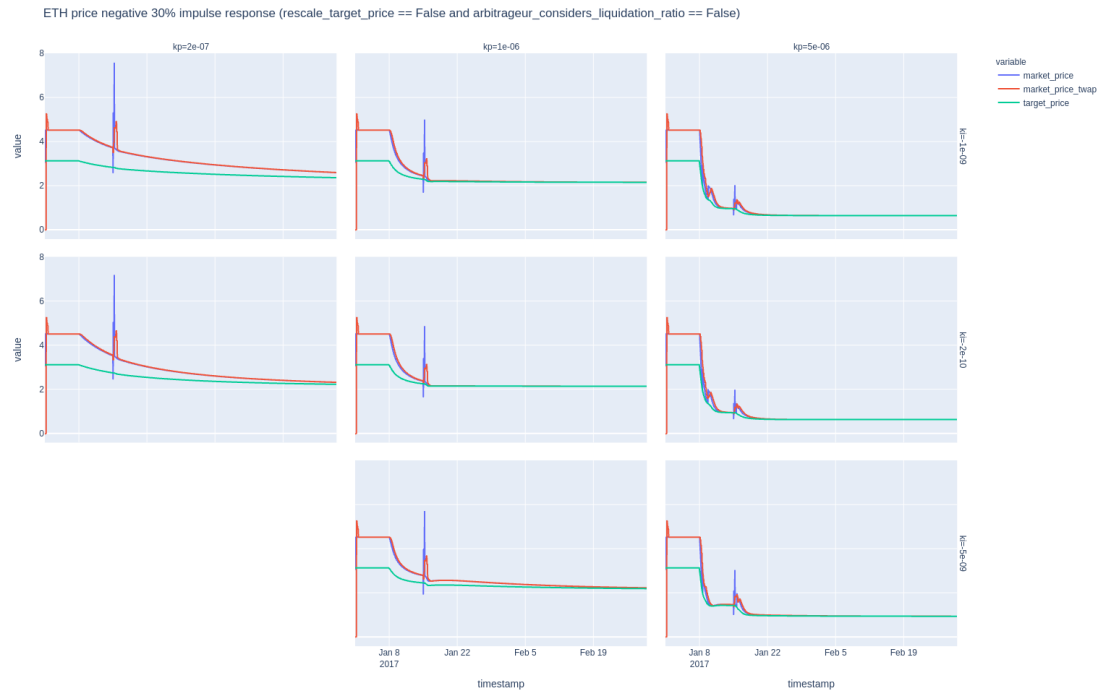
```



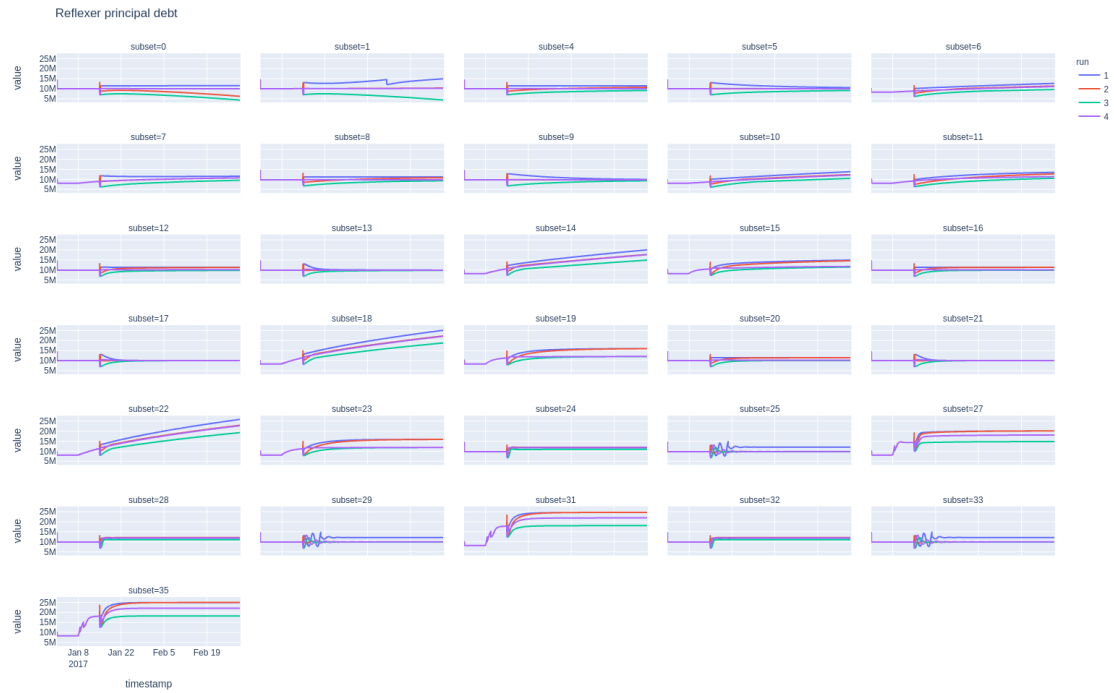
```
[45]: fig = px.line(
    df_stable_price.query('run == 3 and rescale_target_price == False and
    ↳arbitrageur_considers_liquidation_ratio == False'),
    title="ETH price negative 30% step response (rescale_target_price == False
    ↳and arbitrageur_considers_liquidation_ratio == False)",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
fig.show()
```



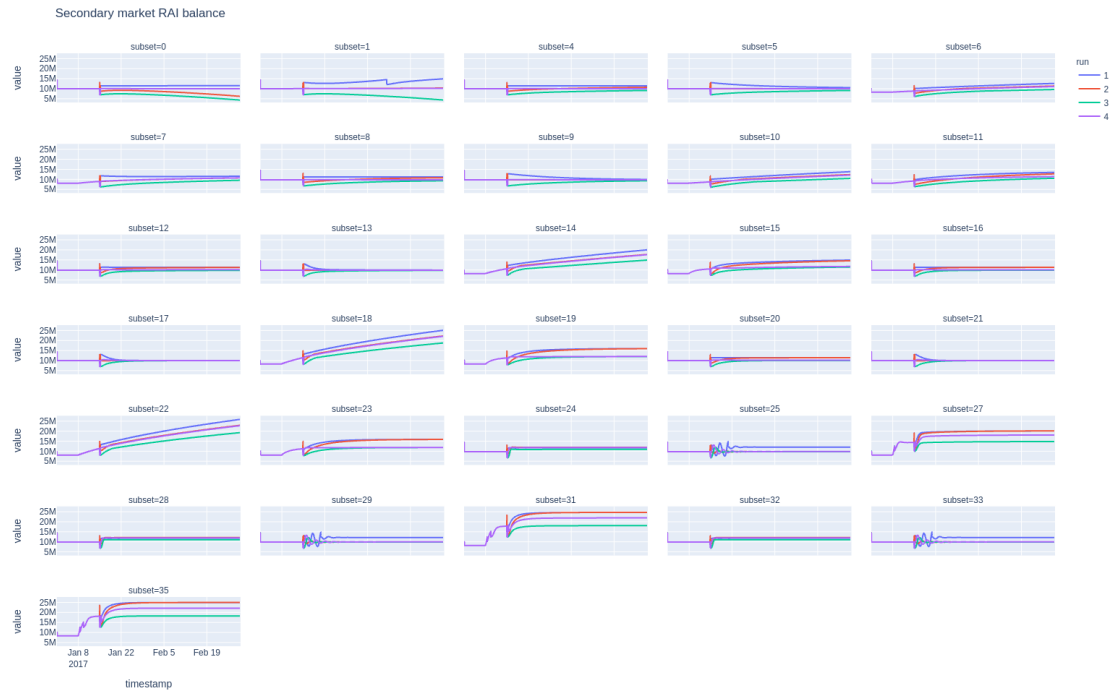
```
[44]: fig = px.line(
    df_stable_price.query('run == 4 and rescale_target_price == False and
    ↪arbitrageur_considers_liquidation_ratio == False'),
    title="ETH price negative 30% impulse response (rescale_target_price ==
    ↪False and arbitrageur_considers_liquidation_ratio == False)",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
fig.show()
```



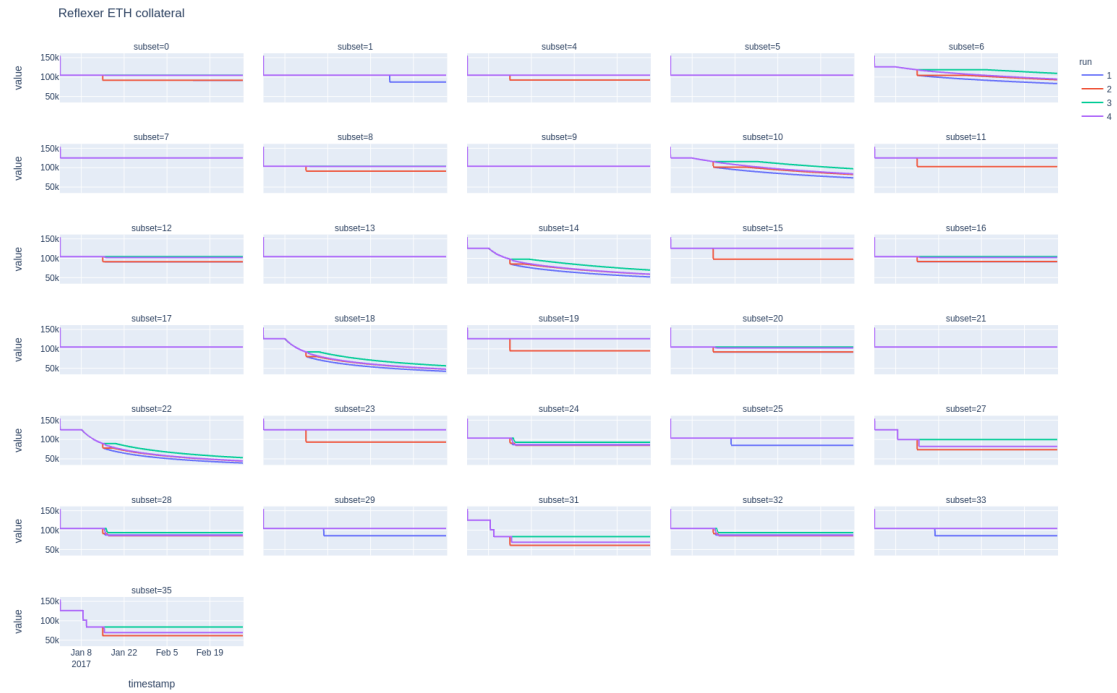
```
[39]: fig = px.line(
    df_stable_price,
    title="Reflexer principal debt",
    x="timestamp",
    y=["principal_debt"],
    color='run',
    facet_col="subset",
    facet_col_wrap=5,
    height=1000
)
fig.show()
```



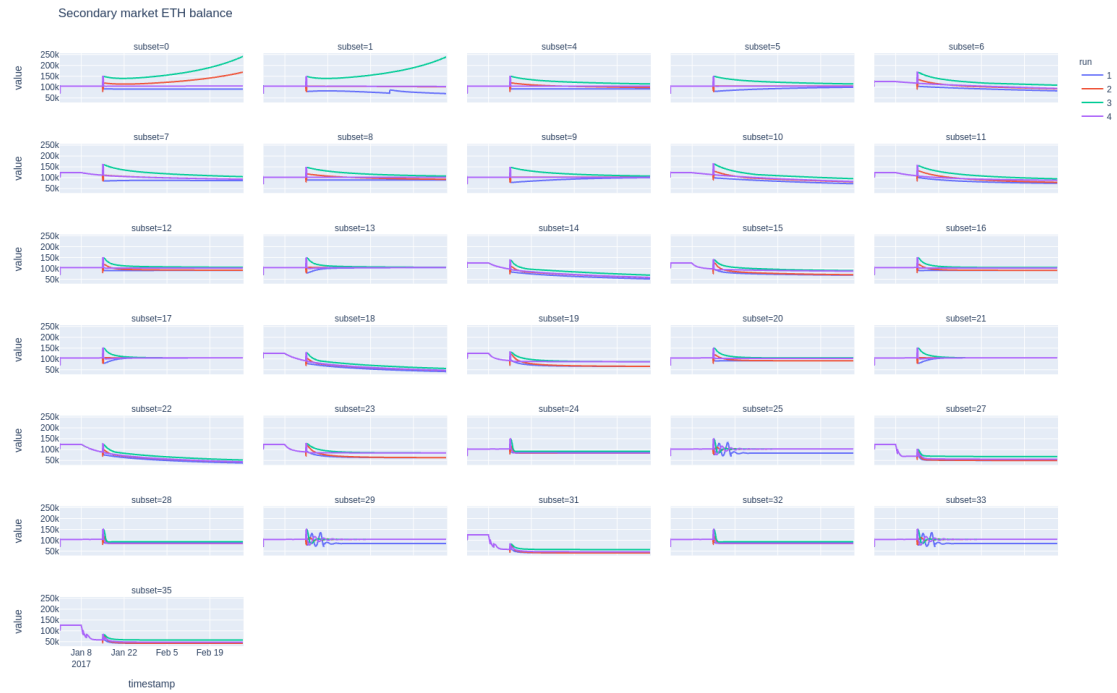
```
[40]: fig = px.line(
    df_stable_price,
    title="Secondary market RAI balance",
    x="timestamp",
    y=["RAI_balance"],
    color='run',
    facet_col="subset",
    facet_col_wrap=5,
    height=1000
)
fig.show()
```



```
[41]: fig = px.line(
    df_stable_price,
    title="Reflexer ETH collateral",
    x="timestamp",
    y=["eth_collateral"],
    color='run',
    facet_col="subset",
    facet_col_wrap=5,
    height=1000
)
fig.show()
```



```
[42]: fig = px.line(
    df_stable_price,
    title="Secondary market ETH balance",
    x="timestamp",
    y=["ETH_balance"],
    color='run',
    facet_col="subset",
    facet_col_wrap=5,
    height=1000
)
fig.show()
```



```
[43]: df_stable_price.plot(
    x='timestamp',
    y=['collateralization_ratio'],
    title='Collateralization ratio',
    facet_col="subset",
    facet_col_wrap=5,
    height=1000
)
```



