

# notebook-experiment-shocks

February 9, 2021

## 1 Experiment Analysis: Controller parameter stability search

Perform shocks of ETH price to test controller parameter stability, without stochastic processes.

- See `experiments/system_model_v3/experiment_shocks.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/bscholtz/workspace/reflexer/venv/lib/python3.8/site-packages  
Requires: ray, pandas, tables, boto3, pathos  
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_shocks/
    ↪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]: ['/processed_results_2021-02-08T19:20:25.053328',
      '/processed_results_2021-02-08T19:45:15.538863',
      '/processed_results_2021-02-08T20:05:35.827550',
      '/processed_results_2021-02-08T22:54:43.005952',
```

```

'/processed_results_2021-02-08T23:20:54.834451',
'/processed_results_2021-02-09T09:43:01.154807',
'/processed_results_2021-02-09T11:59:26.345803',
'/results_2021-02-08T19:11:14.829535',
'/results_2021-02-08T19:20:25.053328',
'/results_2021-02-08T19:45:15.538863',
'/results_2021-02-08T20:05:35.827550',
'/results_2021-02-08T22:54:43.005952',
'/results_2021-02-08T23:20:54.834451',
'/results_2021-02-09T09:43:01.154807',
'/results_2021-02-09T11:44:24.150805',
'/results_2021-02-09T11:48:54.447990',
'/results_2021-02-09T11:56:55.060726',
'/results_2021-02-09T11:59:26.345803',
'/results_2021-02-09T15:57:01.325820']

```

```

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

```

```

[7]: ['/exceptions_2021-02-08T19:11:14.829535',
      '/exceptions_2021-02-08T19:20:25.053328',
      '/exceptions_2021-02-08T19:45:15.538863',
      '/exceptions_2021-02-08T20:05:35.827550',
      '/exceptions_2021-02-08T22:54:43.005952',
      '/exceptions_2021-02-08T23:20:54.834451',
      '/exceptions_2021-02-09T09:43:01.154807',
      '/exceptions_2021-02-09T11:44:24.150805',
      '/exceptions_2021-02-09T11:48:54.447990',
      '/exceptions_2021-02-09T11:56:55.060726',
      '/exceptions_2021-02-09T11:59:26.345803',
      '/exceptions_2021-02-09T15:57:01.325820']

```

```

[8]: # Copy a results_ key from the above keys to select the experiment
experiment_results_key = 'results_2021-02-09T15:57:01.325820' # Or select last_
    ↪ result: experiment_results_keys[-1]
experiment_timestamp = experiment_results_key.strip('results_')
exceptions_key = 'exceptions_' + experiment_timestamp
experiment_timestamp

```

```

[8]: '2021-02-09T15:57:01.325820'

```

```

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

```

```

[9]:
3240120  {'cdp_count': 1, 'open_cdp_count': 1, 'closed_...      {}
3240121  {'cdp_count': 1, 'open_cdp_count': 1, 'closed_...      {}

```

3240122	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...	{}
3240123	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...	{}
3240124	{'cdp_count': 1, 'open_cdp_count': 1, 'closed_...	{}

	sim_metrics	timedelta	cumulative_time	timestamp	blockheight	\
3240120	{}	3600	5184000	2017-03-02	0	
3240121	{}	3600	5184000	2017-03-02	0	
3240122	{}	3600	5184000	2017-03-02	0	
3240123	{}	3600	5184000	2017-03-02	0	
3240124	{}	3600	5184000	2017-03-02	0	

	eth_price	liquidity_demand	liquidity_demand_mean	\
3240120	300.0	0.0	0.0	
3240121	300.0	0.0	0.0	
3240122	300.0	0.0	0.0	
3240123	300.0	0.0	0.0	
3240124	300.0	0.0	0.0	

					cdps	eth_collateral	\
3240120	open	time	locked	drawn	...	105925.122117	
3240121	open	time	locked	drawn	...	105925.122117	
3240122	open	time	locked	drawn	...	105925.122117	
3240123	open	time	locked	drawn	...	105925.122117	
3240124	open	time	locked	drawn	...	105925.122117	

	eth_locked	eth_freed	eth_bitten	principal_debt	\
3240120	154827.528922	48902.406805	0.0	1.012023e+07	
3240121	154827.528922	48902.406805	0.0	1.012023e+07	
3240122	154827.528922	48902.406805	0.0	1.012023e+07	
3240123	154827.528922	48902.406805	0.0	1.012023e+07	
3240124	154827.528922	48902.406805	0.0	1.012023e+07	

	rai_drawn	rai_wiped	rai_bitten	accrued_interest	\
3240120	1.753607e+07	7.415836e+06	0.0	8310.947187	
3240121	1.753607e+07	7.415836e+06	0.0	8310.947187	
3240122	1.753607e+07	7.415836e+06	0.0	8310.947187	
3240123	1.753607e+07	7.415836e+06	0.0	8310.947187	
3240124	1.753607e+07	7.415836e+06	0.0	8310.947187	

	interest_dripped	interest_wiped	interest_bitten	w_1	w_2	w_3	\
3240120	0	0	0	0.0	0.0	0.0	
3240121	0	0	0	0.0	0.0	0.0	
3240122	0	0	0	0.0	0.0	0.0	
3240123	0	0	0	0.0	0.0	0.0	
3240124	0	0	0	0.0	0.0	0.0	

system_revenue	stability_fee	market_price	market_price_twap	\
----------------	---------------	--------------	-------------------	---

3240120	0.0	1.585490e-10	3.139079	3.139079
3240121	0.0	1.585490e-10	3.139079	3.139079
3240122	0.0	1.585490e-10	3.139079	3.139079
3240123	0.0	1.585490e-10	3.139079	3.139079
3240124	0.0	1.585490e-10	3.139079	3.139079

	target_price	target_rate	eth_return	eth_gross_return	\
3240120	2.165517	0.0	0.0	1.0	
3240121	2.165517	0.0	0.0	1.0	
3240122	2.165517	0.0	0.0	1.0	
3240123	2.165517	0.0	0.0	1.0	
3240124	2.165517	0.0	0.0	1.0	

	expected_market_price	expected_debt_price	error_star	\
3240120	3.230577	3.14	0.000921	
3240121	3.230577	3.14	0.000921	
3240122	3.230577	3.14	0.000921	
3240123	3.230577	3.14	0.000921	
3240124	3.230577	3.14	0.000921	

	error_star_integral	market_slippage	RAI_balance	ETH_balance	\
3240120	445.315323	0.0	1.012023e+07	105894.055592	
3240121	445.315323	0.0	1.012023e+07	105894.055592	
3240122	445.315323	0.0	1.012023e+07	105894.055592	
3240123	445.315323	0.0	1.012023e+07	105894.055592	
3240124	445.315323	0.0	1.012023e+07	105894.055592	

	UNI_supply	uniswap_oracle	\
3240120	10000000.0	<models.system_model_v3.model.parts.uniswap_or...	
3240121	10000000.0	<models.system_model_v3.model.parts.uniswap_or...	
3240122	10000000.0	<models.system_model_v3.model.parts.uniswap_or...	
3240123	10000000.0	<models.system_model_v3.model.parts.uniswap_or...	
3240124	10000000.0	<models.system_model_v3.model.parts.uniswap_or...	

	simulation	subset	run	substep	timestep	events
3240120	0	24	5	14	1440	NaN
3240121	0	24	5	15	1440	NaN
3240122	0	24	5	16	1440	NaN
3240123	0	24	5	17	1440	NaN
3240124	0	24	5	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_shocks import params, \
↳ SIMULATION_TIMESTEPS
```

```
* Number of timesteps: 1440 / 60.0 days
* Number of MC runs: 5
* Timestep duration: 0.004 seconds
* Control parameters: ['kp', 'ki']
* Number of parameter combinations: 25
* Expected experiment duration: 12.0 minutes / 0.2 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'])
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_...
...
180120 3240052 {'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
180121 3240070 {'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
180122 3240088 {'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
180123 3240106 {'cdp_count': 1, 'open_cdp_count': 1, 'closed_...
180124 3240124 {'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
...
180120                {}          {}          3600          5169600
180121                {}          {}          3600          5173200
180122                {}          {}          3600          5176800
180123                {}          {}          3600          5180400
180124                {}          {}          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
...
180120 2017-03-01 20:00:00          0  300.000000            -0.0
180121 2017-03-01 21:00:00          0  300.000000            -0.0
180122 2017-03-01 22:00:00          0  300.000000             0.0
180123 2017-03-01 23:00:00          0  300.000000             0.0
180124 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
...
180120                0.0000

```

180121	0.0000
180122	0.0000
180123	0.0000
180124	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	
...		...	...	
180120		None	105925.122117	
180121		None	105925.122117	
180122		None	105925.122117	
180123	open time	locked	drawn	... 105925.122117
180124	open time	locked	drawn	... 105925.122117

	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	
...	...	...	...	...	...	
180120	154827.528922	48902.406805	0.0	1.012023e+07	1.753607e+07	
180121	154827.528922	48902.406805	0.0	1.012023e+07	1.753607e+07	
180122	154827.528922	48902.406805	0.0	1.012023e+07	1.753607e+07	
180123	154827.528922	48902.406805	0.0	1.012023e+07	1.753607e+07	
180124	154827.528922	48902.406805	0.0	1.012023e+07	1.753607e+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	
...	...	...	...	...	
180120	7.415836e+06	0.0	8287.822681	0	
180121	7.415836e+06	0.0	8293.603802	0	
180122	7.415836e+06	0.0	8299.384927	0	
180123	7.415836e+06	0.0	8305.166055	0	
180124	7.415836e+06	0.0	8310.947187	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
...	...	...	...	...	...
180120	0	0	0.000000e+00	0.0	0.0
180121	0	0	0.000000e+00	0.0	0.0
180122	0	0	0.000000e+00	0.0	0.0
180123	0	0	0.000000e+00	0.0	0.0
180124	0	0	0.000000e+00	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	
...	...	...	...	...	
180120	0.0	1.585490e-10	3.139079	3.139079	
180121	0.0	1.585490e-10	3.139079	3.139079	
180122	0.0	1.585490e-10	3.139079	3.139079	
180123	0.0	1.585490e-10	3.139079	3.139079	
180124	0.0	1.585490e-10	3.139079	3.139079	

	target_price	target_rate	eth_return	eth_gross_return	\
0	3.140000	0.0	0.000000	0.0	
1	2.165517	0.0	0.020168	1.0	
2	2.165517	0.0	0.000000	1.0	
3	2.165517	0.0	0.000000	1.0	
4	2.165517	0.0	0.000000	1.0	
...	...	...	...	...	
180120	2.165517	0.0	0.000000	1.0	
180121	2.165517	0.0	0.000000	1.0	
180122	2.165517	0.0	0.000000	1.0	
180123	2.165517	0.0	0.000000	1.0	
180124	2.165517	0.0	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	
...	...	...	...	
180120	3.230576	3.14	0.000921	
180121	3.230576	3.14	0.000921	
180122	3.230577	3.14	0.000921	
180123	3.230577	3.14	0.000921	

180124 3.230577 3.14 0.000921

	error_star_integral	market_slippage	RAI_balance	ETH_balance \
0	0.000000	0.000000	1.000000e+07	106777.606153
1	5652.000000	0.000000	1.450000e+07	73708.353400
2	16932.000000	0.514331	1.012023e+07	105703.331685
3	28166.000000	-1.054705	1.012023e+07	105703.331685
4	39354.000000	0.000000	1.012023e+07	105703.331685
...	...	...	...	...
180120	441.315323	0.000000	1.012023e+07	105894.055592
180121	442.315323	0.000000	1.012023e+07	105894.055592
180122	443.315323	0.000000	1.012023e+07	105894.055592
180123	444.315323	0.000000	1.012023e+07	105894.055592
180124	445.315323	0.000000	1.012023e+07	105894.055592

	UNI_supply	uniswap_oracle \
0	10000000.0	None
1	10000000.0	None
2	10000000.0	None
3	10000000.0	None
4	10000000.0	None
...	...	...
180120	10000000.0	None
180121	10000000.0	None
180122	10000000.0	None
180123	10000000.0	<models.system_model_v3.model.parts.uniswap_or...
180124	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
...	...	...	...	...	...	...
180120	0	24	5	18	1436	NaN
180121	0	24	5	18	1437	NaN
180122	0	24	5	18	1438	NaN
180123	0	24	5	18	1439	NaN
180124	0	24	5	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

```

...
180120      3.177754e+07      1.450000 -2.000000e-10
180121      3.177754e+07      1.450000 -2.000000e-10
180122      3.177754e+07      1.450000 -2.000000e-10
180123      3.177754e+07      1.450000 -2.000000e-10
180124      3.177754e+07      1.450000 -2.000000e-10

```

```

      kp  liquidation_ratio  target_price_scaled
0      2.000000e-07          1.45              4.553
1      2.000000e-07          1.45              3.140
2      2.000000e-07          1.45              3.140
3      2.000000e-07          1.45              3.140
4      2.000000e-07          1.45              3.140

```

```

...
180120  5.000000e-06          1.45              3.140
180121  5.000000e-06          1.45              3.140
180122  5.000000e-06          1.45              3.140
180123  5.000000e-06          1.45              3.140
180124  5.000000e-06          1.45              3.140

```

[180125 rows x 56 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

```

```

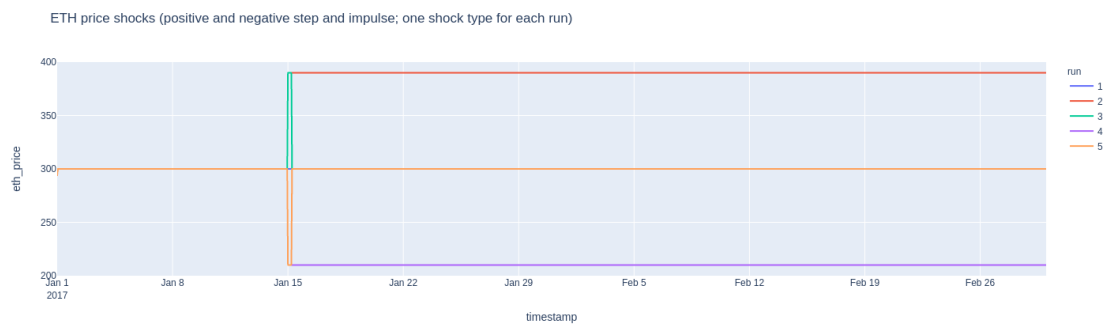
[16]:      subset      kp      ki
0         0  2.000000e-07 -5.000000e-09
1441        1  2.000000e-07 -3.000000e-09
2882        2  2.000000e-07 -1.000000e-09
4323        3  2.000000e-07 -6.000000e-10
5764        4  2.000000e-07 -2.000000e-10
7205        5  6.000000e-07 -5.000000e-09
8646        6  6.000000e-07 -3.000000e-09

```

10087	7	6.000000e-07	-1.000000e-09
11528	8	6.000000e-07	-6.000000e-10
12969	9	6.000000e-07	-2.000000e-10
14410	10	1.000000e-06	-5.000000e-09
15851	11	1.000000e-06	-3.000000e-09
17292	12	1.000000e-06	-1.000000e-09
18733	13	1.000000e-06	-6.000000e-10
20174	14	1.000000e-06	-2.000000e-10
21615	15	3.000000e-06	-5.000000e-09
23056	16	3.000000e-06	-3.000000e-09
24497	17	3.000000e-06	-1.000000e-09
25938	18	3.000000e-06	-6.000000e-10
27379	19	3.000000e-06	-2.000000e-10
28820	20	5.000000e-06	-5.000000e-09
30261	21	5.000000e-06	-3.000000e-09
31702	22	5.000000e-06	-1.000000e-09
33143	23	5.000000e-06	-6.000000e-10
34584	24	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'
    )
```

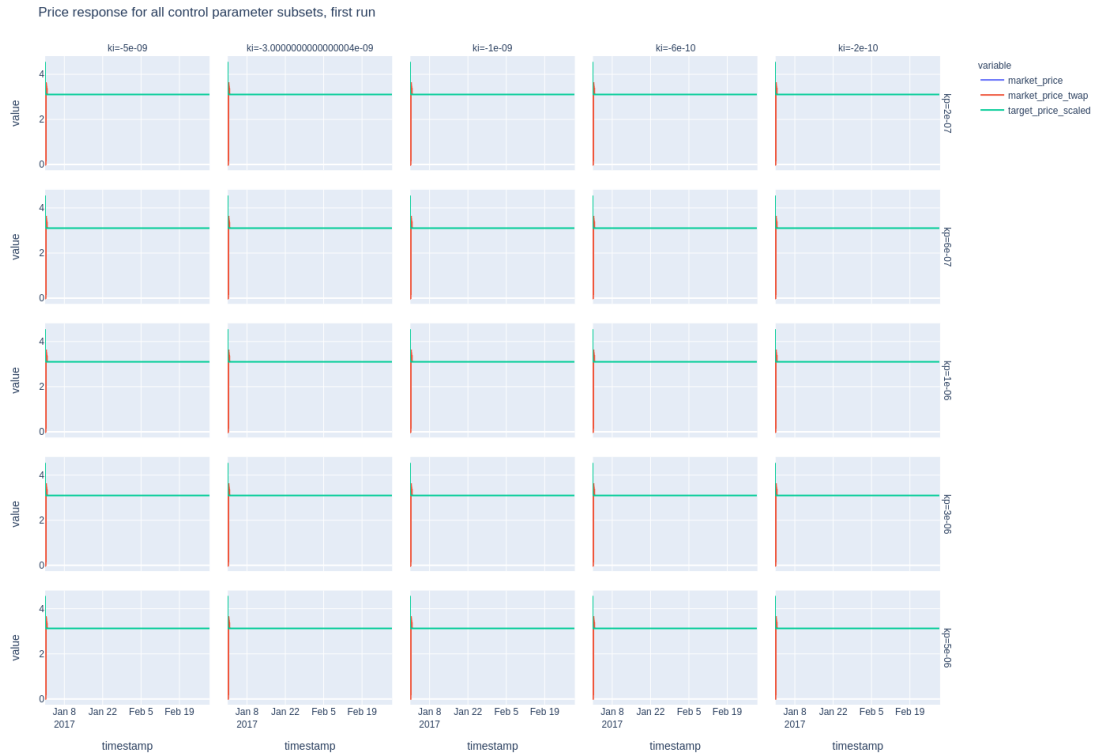


```
[18]: fig = px.line(
        df.query('run == 1'),
        title="Price response for all control parameter subsets, first run",
        x="timestamp",
```

```

y=["market_price", "market_price_twap", "target_price_scaled"],
facet_col="ki",
facet_row="kp",
height=1000
)
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]: Empty DataFrame
      Columns: [subset, kp, ki]
      Index: []

```

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]: Empty DataFrame
Columns: [subset, kp, ki]
Index: []
```

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

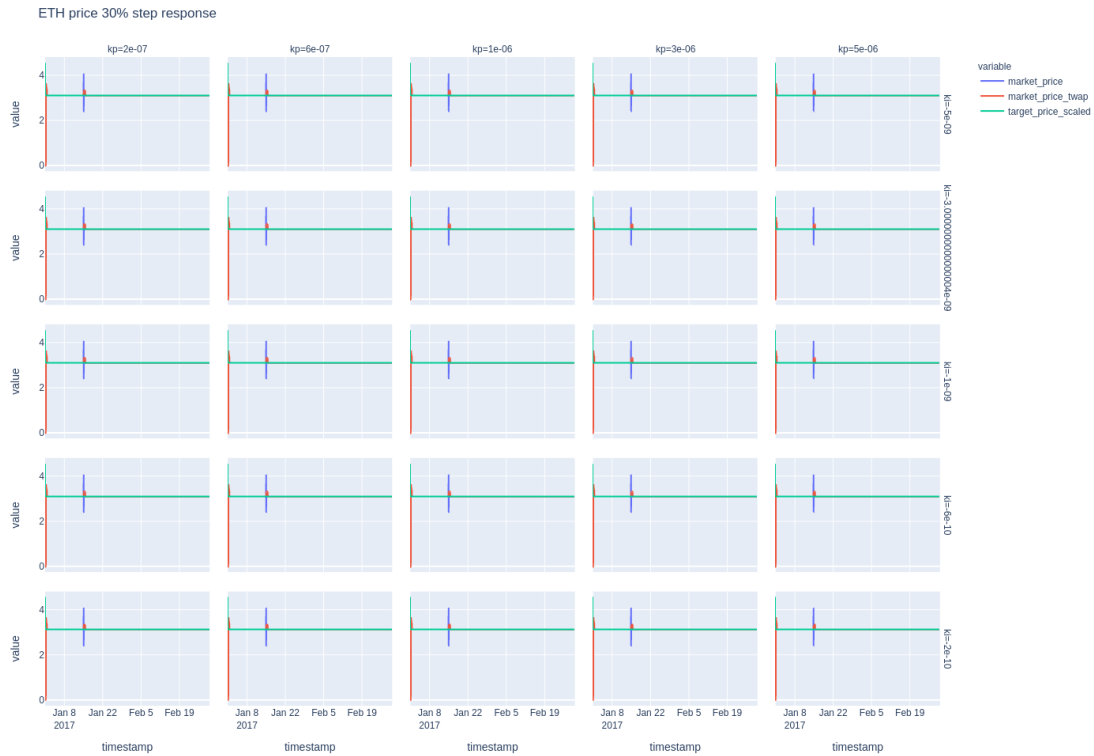
```
[22]: df['stable_price'] = False
df.loc[df.eval("""
0.1*@initial_target_price < market_price <= 10*@initial_target_price and 0.
↳ 1*@initial_target_price < target_price_scaled <= 10*@initial_target_price
"""), 'stable_price'] = True
df_stable_price = df.groupby("subset").filter(lambda x: all(x.query('timestep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

```
[22]: array([ 0,  1,  2,  3,  4,  5,  6,  7,  8,  9, 10, 11, 12, 13, 14, 15, 16,
        17, 18, 19, 20, 21, 22, 23, 24])
```

```
[23]: fig = px.line(
    df_stable_price.query('run == 1'),
    title="Base case: Stable ETH price response",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
# fig.for_each_annotation(lambda a: a.update(text = f"kp={param_sweep[int(a.
↳ text.split('=')[-1])]['kp']} ki={param_sweep[int(a.text.
↳ split('=')[-1])]['ki']}"))
fig.show()
```

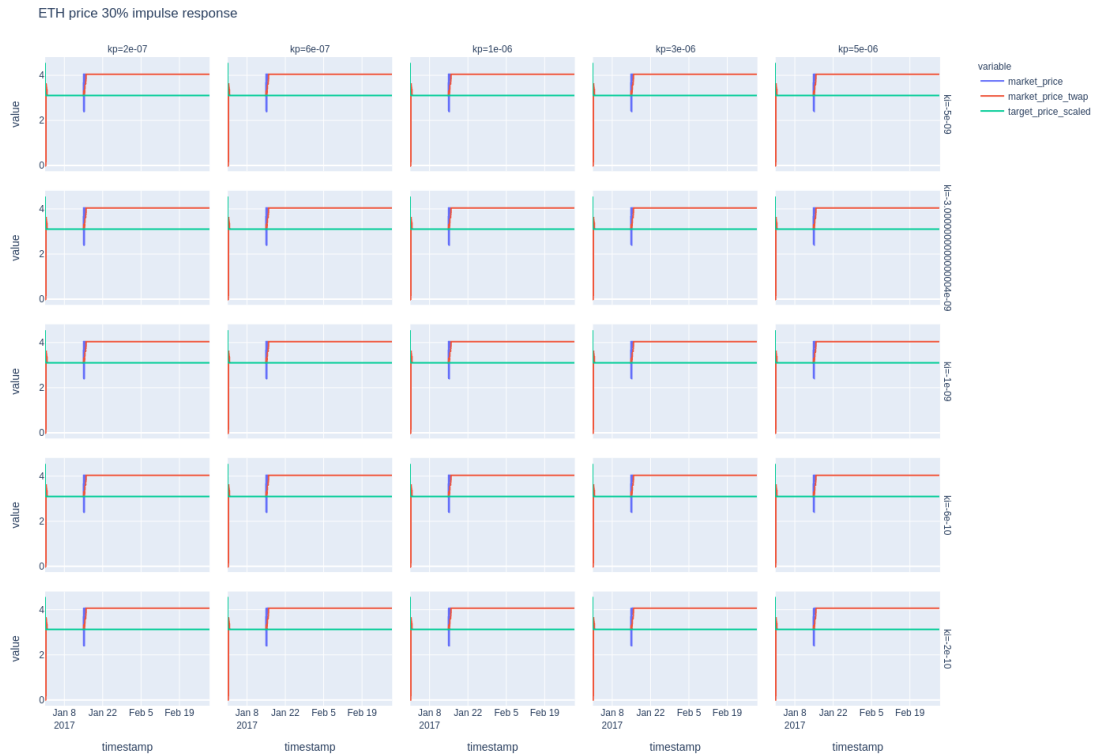


```
[24]: fig = px.line(
    df_stable_price.query('run == 2'),
    title="ETH price 30% step response",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price_scaled"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
fig.show()
```

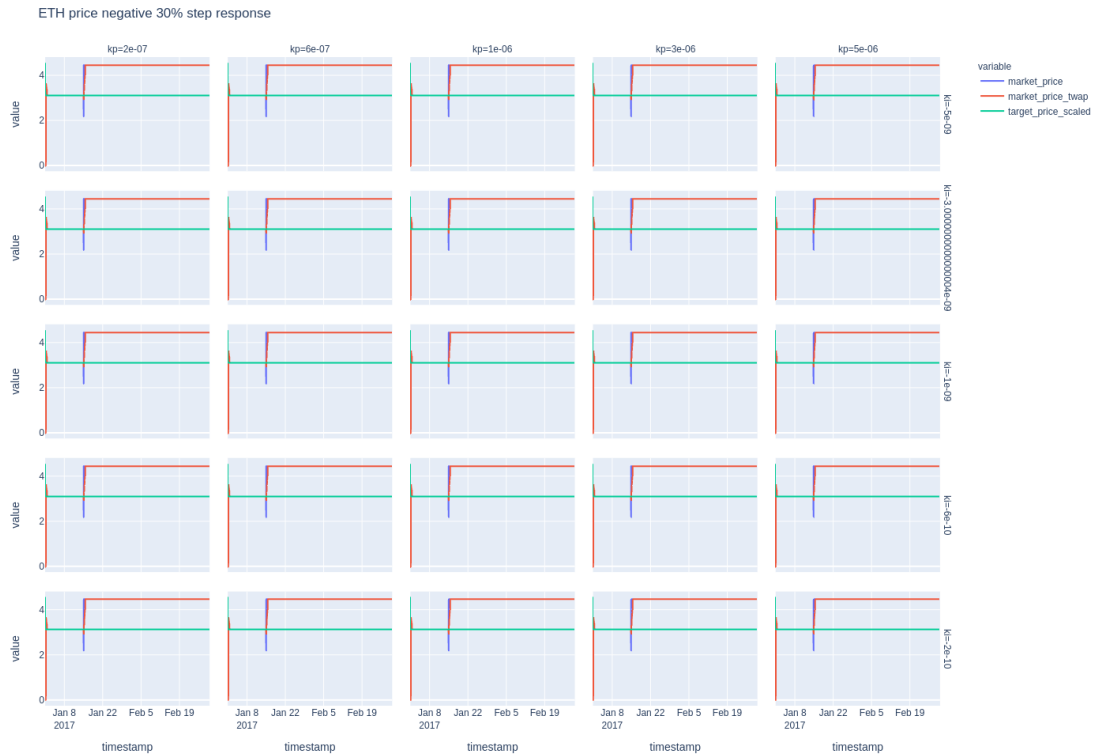


```
[25]: fig = px.line(
    df_stable_price.query('run == 3'),
    title="ETH price 30% impulse response",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price_scaled"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
fig.show()
```

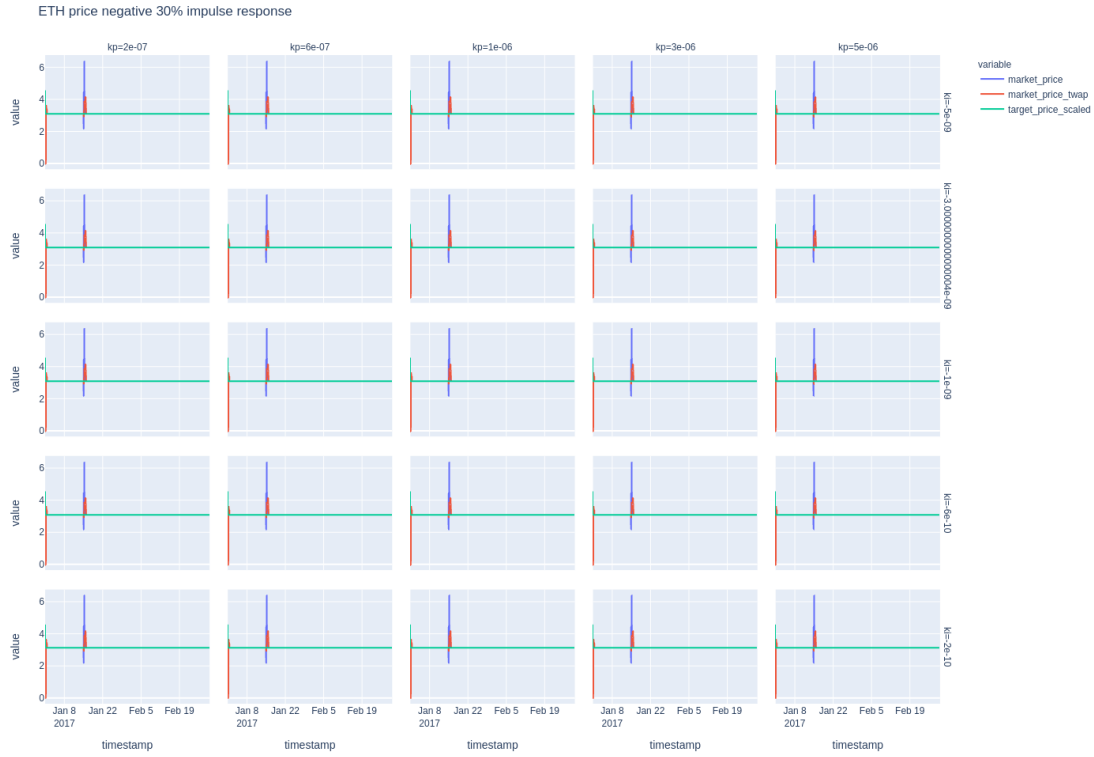




```
[28]: fig = px.line(
    df_stable_price.query('run == 4'),
    title="ETH price negative 30% step response",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price_scaled"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
fig.show()
```



```
[29]: fig = px.line(
    df_stable_price.query('run == 5'),
    title="ETH price negative 30% impulse response",
    x="timestamp",
    y=["market_price", "market_price_twap", "target_price_scaled"],
    facet_col="kp",
    facet_row="ki",
    facet_col_wrap=2,
    height=1000
)
fig.show()
```

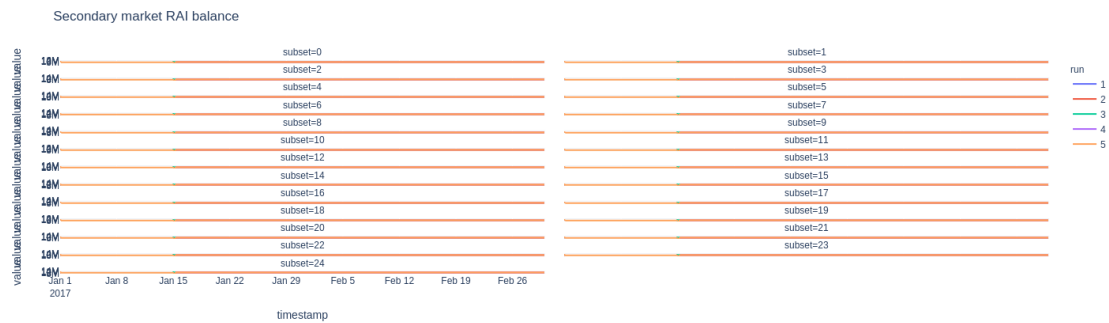


```
[35]: fig = px.line(
    df_stable_price,
    title="Reflexer principal debt",
    x="timestamp",
    y=["principal_debt"],
    color='run',
    facet_col="subset",
    facet_col_wrap=2
)
fig.show()
```

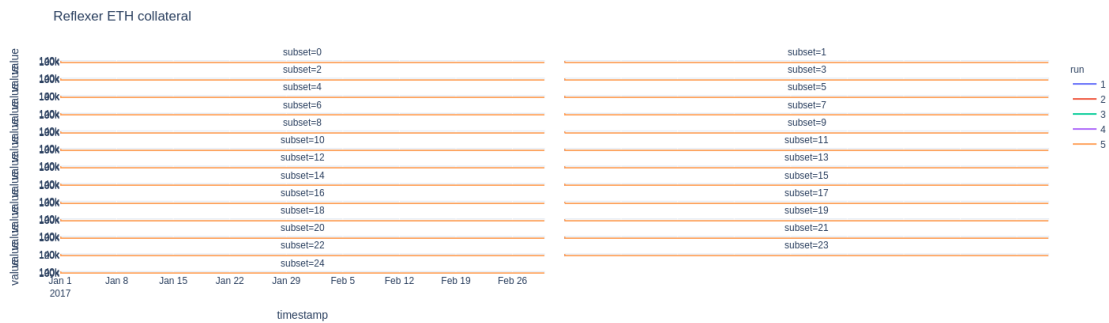
- \* Number of timesteps: 1440 / 60.0 days
- \* Number of MC runs: 5
- \* Timestep duration: 0.004 seconds
- \* Control parameters: ['kp', 'ki']
- \* Number of parameter combinations: 25
- \* Expected experiment duration: 12.0 minutes / 0.2 hours



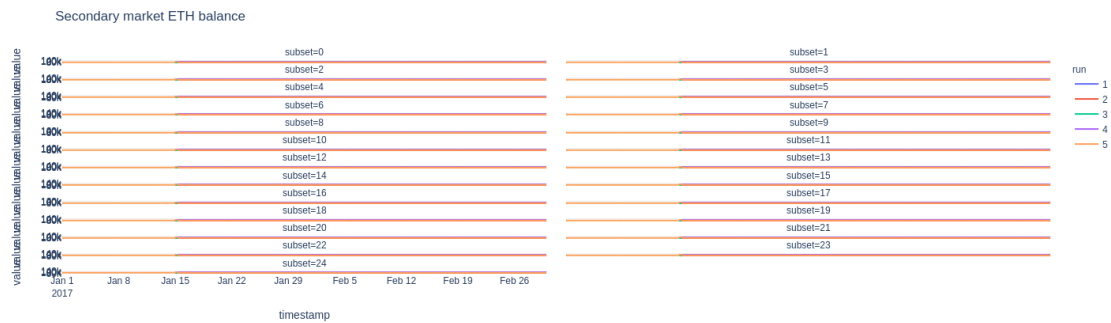
```
[31]: fig = px.line(
    df_stable_price,
    title="Secondary market RAI balance",
    x="timestamp",
    y=["RAI_balance"],
    color='run',
    facet_col="subset",
    facet_col_wrap=2
)
fig.show()
```



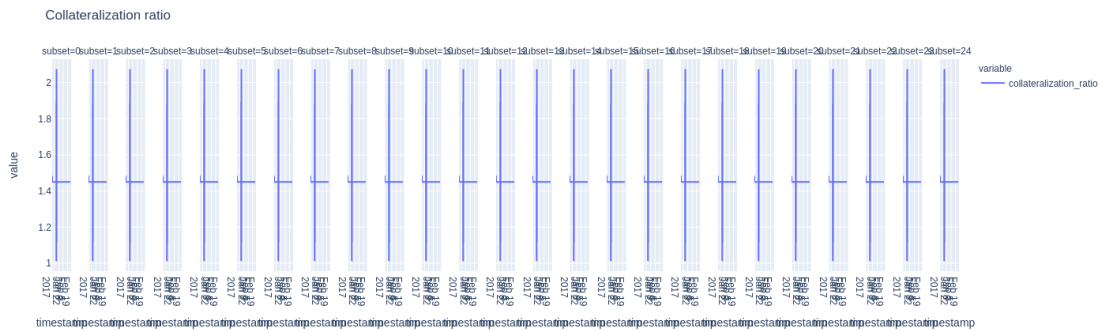
```
[32]: fig = px.line(
    df_stable_price,
    title="Reflexer ETH collateral",
    x="timestamp",
    y=["eth_collateral"],
    color='run',
    facet_col="subset",
    facet_col_wrap=2
)
fig.show()
```



```
[33]: fig = px.line(
    df_stable_price,
    title="Secondary market ETH balance",
    x="timestamp",
    y=["ETH_balance"],
    color='run',
    facet_col="subset",
    facet_col_wrap=2
)
fig.show()
```



```
[34]: df_stable_price.plot(
    x='timestamp',
    y=['collateralization_ratio'],
    title='Collateralization ratio',
    facet_col="subset",
)
```



```
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```

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