

Set-Up

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import matplotlib as mpl
import statsmodels.tsa.api as tsa
from scipy.linalg import cholesky
from scipy.optimize import minimize, LinearConstraint, NonlinearConstraint
from scipy.linalg import orth

# Plot Setting
plt.rc('text', usetex=True)
plt.rc('font', family='serif')
plt.rc('font', serif='Computer Modern')
```

Transform data

```
In [2]: mydata = pd.read_csv("mydata.csv")
mydata['time'] = pd.to_datetime(mydata['time']).dt.to_period('Q')
mydata.set_index('time', inplace=True)
mydata.head()
```

C:\Users\Yutao\AppData\Local\Temp\ipykernel_5432\2964630960.py:2: UserWarning: Could not infer format, so each element will be parsed individually, falling back to `dateutil`. To ensure parsing is consistent and as-expected, please specify a format.

```
mydata['time'] = pd.to_datetime(mydata['time']).dt.to_period('Q')
```

Out[2]:

	Interest Rate	Real Wages	Adjusted Reserves	PPIC	GDP	private_consumption	gover
time							
1954Q3	1.03	42.832	8.191	31.4	390996000000	240303000000	
1954Q4	0.99	43.388	8.362	31.1	399734000000	245093000000	
1955Q1	1.34	43.629	8.339	30.9	413073000000	251398000000	
1955Q2	1.50	44.054	8.358	30.6	421532000000	256466000000	
1955Q3	1.94	44.802	8.320	30.3	430221000000	260651000000	

```
In [3]: denominator = mydata["gdp_deflator"] * mydata["Population"]
print(denominator.head())
names = [
    "GDP",
    "private_consumption",
    "government_expenditure",
    "government_revenue",
    "private_non_residential_investment",
]
```

```
for name in names:
    mydata[f"real_{name}"] = mydata[name] / denominator
mydata.iloc[:, 1:] = mydata.iloc[:, 1:].map(lambda x: np.log(x))
mydata.head()
```

```
time
1954Q3    2.208911e+09
1954Q4    2.225725e+09
1955Q1    2.245490e+09
1955Q2    2.263748e+09
1955Q3    2.289917e+09
Freq: Q-DEC, dtype: float64
```

```

C:\Users\Yutao\AppData\Local\Temp\ipykernel_5432\1810958852.py:12: FutureWarning: Setting an item of incompatible dtype is deprecated and will raise in a future error of pandas. Value 'time
1954Q3      26.691963
1954Q4      26.714065
1955Q1      26.746890
1955Q2      26.767162
1955Q3      26.787565
...
2014Q4      30.516496
2015Q1      30.524916
2015Q2      30.536817
2015Q3      30.543460
2015Q4      30.545280
Freq: Q-DEC, Name: GDP, Length: 246, dtype: float64' has dtype incompatible with int64, please explicitly cast to a compatible dtype first.
    mydata.iloc[:, 1:] = mydata.iloc[:, 1:].map(lambda x: np.log(x))
C:\Users\Yutao\AppData\Local\Temp\ipykernel_5432\1810958852.py:12: FutureWarning: Setting an item of incompatible dtype is deprecated and will raise in a future error of pandas. Value 'time
1954Q3      26.205166
1954Q4      26.224904
1955Q1      26.250303
1955Q2      26.270262
1955Q3      26.286448
...
2014Q4      30.122702
2015Q1      30.125859
2015Q2      30.137701
2015Q3      30.147305
2015Q4      30.150599
Freq: Q-DEC, Name: private_consumption, Length: 246, dtype: float64' has dtype incompatible with int64, please explicitly cast to a compatible dtype first.
    mydata.iloc[:, 1:] = mydata.iloc[:, 1:].map(lambda x: np.log(x))
C:\Users\Yutao\AppData\Local\Temp\ipykernel_5432\1810958852.py:12: FutureWarning: Setting an item of incompatible dtype is deprecated and will raise in a future error of pandas. Value 'time
1954Q3      25.235750
1954Q4      25.238030
1955Q1      25.243629
1955Q2      25.250831
1955Q3      25.267613
...
2014Q4      28.790808
2015Q1      28.790388
2015Q2      28.804762
2015Q3      28.810769
2015Q4      28.812178
Freq: Q-DEC, Name: government_expenditure, Length: 246, dtype: float64' has dtype incompatible with int64, please explicitly cast to a compatible dtype first.
    mydata.iloc[:, 1:] = mydata.iloc[:, 1:].map(lambda x: np.log(x))
C:\Users\Yutao\AppData\Local\Temp\ipykernel_5432\1810958852.py:12: FutureWarning: Setting an item of incompatible dtype is deprecated and will raise in a future error of pandas. Value 'time
1954Q3      24.328438
1954Q4      24.375844

```

```
1955Q1    24.491026
1955Q2    24.566646
1955Q3    24.617552
...
2014Q4    28.577965
2015Q1    28.614003
2015Q2    28.610021
2015Q3    28.605101
2015Q4    28.589439
Freq: Q-DEC, Name: private_non_residential_investment, Length: 246, dtype: float64'
has dtype incompatible with int64, please explicitly cast to a compatible dtype first.
mydata.iloc[:, 1:] = mydata.iloc[:, 1:].map(lambda x: np.log(x))
```

Out[3]:

	Interest Rate	Real Wages	Adjusted Reserves	PPIC	GDP	private_consumption	gov
time							
1954Q3	1.03	3.757285	2.103036	3.446808	26.691963	26.205166	
1954Q4	0.99	3.770183	2.123698	3.437208	26.714065	26.224904	
1955Q1	1.34	3.775722	2.120943	3.430756	26.746890	26.250303	
1955Q2	1.50	3.785416	2.123219	3.421000	26.767162	26.270262	
1955Q3	1.94	3.802253	2.118662	3.411148	26.787565	26.286448	

In [4]:

```
filter_data = mydata[(mydata.index >= "1955Q1") & (mydata.index <= "2000Q4")]
filter_data.head()
```

Out[4]:

	Interest Rate	Real Wages	Adjusted Reserves	PPIC	GDP	private_consumption	gov
time							
1955Q1	1.34	3.775722	2.120943	3.430756	26.746890	26.250303	
1955Q2	1.50	3.785416	2.123219	3.421000	26.767162	26.270262	
1955Q3	1.94	3.802253	2.118662	3.411148	26.787565	26.286448	
1955Q4	2.36	3.808261	2.119743	3.394508	26.803410	26.301632	
1956Q1	2.48	3.823585	2.124893	3.391147	26.809463	26.307348	

VAR Estimation

In [5]:

```
names = ["real_GDP", "real_private_consumption", "real_government_expenditure", "re
X = filter_data[names]
X.to_csv(r"D:\Program Files (x86)\Project\PythonProject\Python Time Series\Replicat

In [6]:
```

```
var_res = tsa.VAR(X).fit(6, trend="n")
```

```
print(var_res.is_stable())
```

False

Get Parameter

```
In [7]: sigma = var_res.sigma_u
        params = var_res.params
        B0 = cholesky(sigma, lower=True)
        dim = B0.shape[0]
```

Define f function

```
In [8]: def f(x):
        if x >= 0:
            res = 100 * x
        else:
            res = x
        return res
```

Define get_q

```
In [9]: def get_q(m):

        random_vector = np.random.randn(m)
        q = random_vector / np.linalg.norm(random_vector)

        return q
```

Define get_r_ja

```
In [10]: def get_r_ja(var_res, q, horizon=10):

        q = np.asarray(q)
        n = var_res.neqs

        irf = var_res.irf(horizon)
        orth_irfs = irf.orth_irfs

        r_a = np.zeros((horizon+1, n))

        for k in range(horizon + 1):
            r_a[k, :] = orth_irfs[k] @ q

        return r_a
```

Test function

```
In [11]: q = get_q(10)
        irf = var_res.irf(10)
        orth_irfs = irf.orth_irfs
```

```
r_a_by_func = get_r_ja(var_res=var_res, q=q)
print(orth_irfs[1,:,:@q])
```

```
[-0.0021544 -0.00223172  0.00033237 -0.00348191 -0.00279819 -0.00381137
  0.28617107 -0.0111038  0.0177741  0.00042011]
```

In [12]: `print(r_a_by_func[1,])`

```
[-0.0021544 -0.00223172  0.00033237 -0.00348191 -0.00279819 -0.00381137
  0.28617107 -0.0111038  0.0177741  0.00042011]
```

Define psi function

```
In [13]: def psi(q, positive, negative, periods):
    positive_part = 0
    negative_part = 0
    if len(positive) > 0:
        for positive_index in positive:
            for period in range(periods + 1):
                std = np.sqrt(np.diag(sigma)[positive_index])
                inner_part = get_r_ja(var_res, q=q, horizon=10)[period, positive_index]
                positive_part += f(-1 * inner_part)

    if len(negative) > 0:
        for negative_index in negative:
            for period in range(periods + 1):
                std = np.sqrt(np.diag(sigma)[negative_index])
                inner_part = get_r_ja(var_res, q=q, horizon=10)[period, negative_index]
                negative_part += f(inner_part)
    res = positive_part + negative_part
    return res
```

Business Cycle shocks

```
In [14]: from scipy.optimize import minimize
    positive = [0, 1, 3, 5]
    negative = []
    periods = 3

    q0 = get_q(10)

    norm_constraint = NonlinearConstraint(lambda q: np.sum(q**2), lb=1.0, ub=1.0)

    res = minimize(
        fun=lambda q: psi(q, positive, negative, periods),
        x0=q0,
        method='trust-constr',
        constraints=[norm_constraint],
        options={'verbose': 2, 'maxiter': 1000, 'xtol': 1e-12}
    )

    q_business_shocks = res.x
```

niter	f evals	CG iter	obj func	tr radius	opt	c viol
1	11	0	-4.1430e+00	1.00e+00	1.23e+01	2.22e-16
2	22	1	-1.7090e+01	7.00e+00	3.77e+00	1.00e+00
3	33	2	-1.9690e+01	8.11e+00	3.19e+00	1.34e+00
4	44	3	-1.6301e+01	8.11e+00	1.06e+00	4.55e-01
5	55	4	-1.3979e+01	8.11e+00	4.12e-01	5.94e-02
6	66	5	-1.3616e+01	8.11e+00	9.52e-02	3.42e-03
7	77	6	-1.3594e+01	8.11e+00	1.60e-03	9.77e-05
8	88	7	-1.3594e+01	8.11e+00	4.51e-06	2.84e-08
9	99	8	-1.3594e+01	8.11e+00	3.47e-07	2.19e-13
10	110	10	-1.3594e+01	8.11e+00	5.79e-07	8.22e-15
11	121	11	-1.3594e+01	8.11e+00	1.81e-07	5.11e-15
12	143	13	-1.3594e+01	8.11e+00	5.16e-07	2.22e-16
13	165	16	-1.3594e+01	8.11e+00	4.68e-07	1.11e-16
14	176	19	-1.3594e+01	8.11e+00	1.77e-07	4.44e-16
15	198	24	-1.3594e+01	8.11e-01	1.77e-07	4.44e-16
16	220	29	-1.3594e+01	8.11e-02	1.77e-07	4.44e-16
17	242	35	-1.3594e+01	8.11e-03	1.77e-07	4.44e-16
18	264	43	-1.3594e+01	8.11e-04	1.77e-07	4.44e-16
19	286	50	-1.3594e+01	8.11e-05	1.77e-07	4.44e-16
20	308	55	-1.3594e+01	8.11e-06	1.77e-07	4.44e-16
21	330	60	-1.3594e+01	8.11e-07	1.77e-07	4.44e-16
22	352	65	-1.3594e+01	8.11e-08	1.77e-07	4.44e-16
23	374	70	-1.3594e+01	8.11e-09	1.77e-07	4.44e-16

d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize_differentiable_functions.py:728: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.

```
self.H.update(delta_x, delta_g)
```

24	396	73	-1.3594e+01	4.06e-09	1.77e-07	4.44e-16
25	418	76	-1.3594e+01	1.50e-09	1.77e-07	4.44e-16
26	429	79	-1.3594e+01	7.19e-10	1.77e-07	4.44e-16
27	451	82	-1.3594e+01	2.09e-10	1.77e-07	4.44e-16
28	462	85	-1.3594e+01	4.19e-11	1.77e-07	4.44e-16
29	473	87	-1.3594e+01	1.77e-11	1.77e-07	4.44e-16
30	484	89	-1.3594e+01	4.74e-12	1.77e-07	4.44e-16
31	495	91	-1.3594e+01	1.27e-12	1.77e-07	4.44e-16
32	517	93	-1.3594e+01	2.47e-13	1.77e-07	4.44e-16

`xtol` termination condition is satisfied.

Number of iterations: 32, function evaluations: 517, CG iterations: 93, optimality: 1.77e-07, constraint violation: 4.44e-16, execution time: 1.6e+01 s.

```
In [15]: r_business = get_r_ja(var_res,q_business_shocks, 25)
         print(r_business.shape)
```

(26, 10)

Plot Business shocks

```
In [16]: fig, axes = plt.subplots(5,2, figsize=(10, 10))
         for i in range(5):
             for j in range(2):
                 axes[i,j].plot(r_business[:, i*2+j], color = "black", lw=1.25)
                 axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
```

```
axes[i,j].set_xlim(0,25)
axes[i,j].set_title(names[i*2+j], fontsize=12)
axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

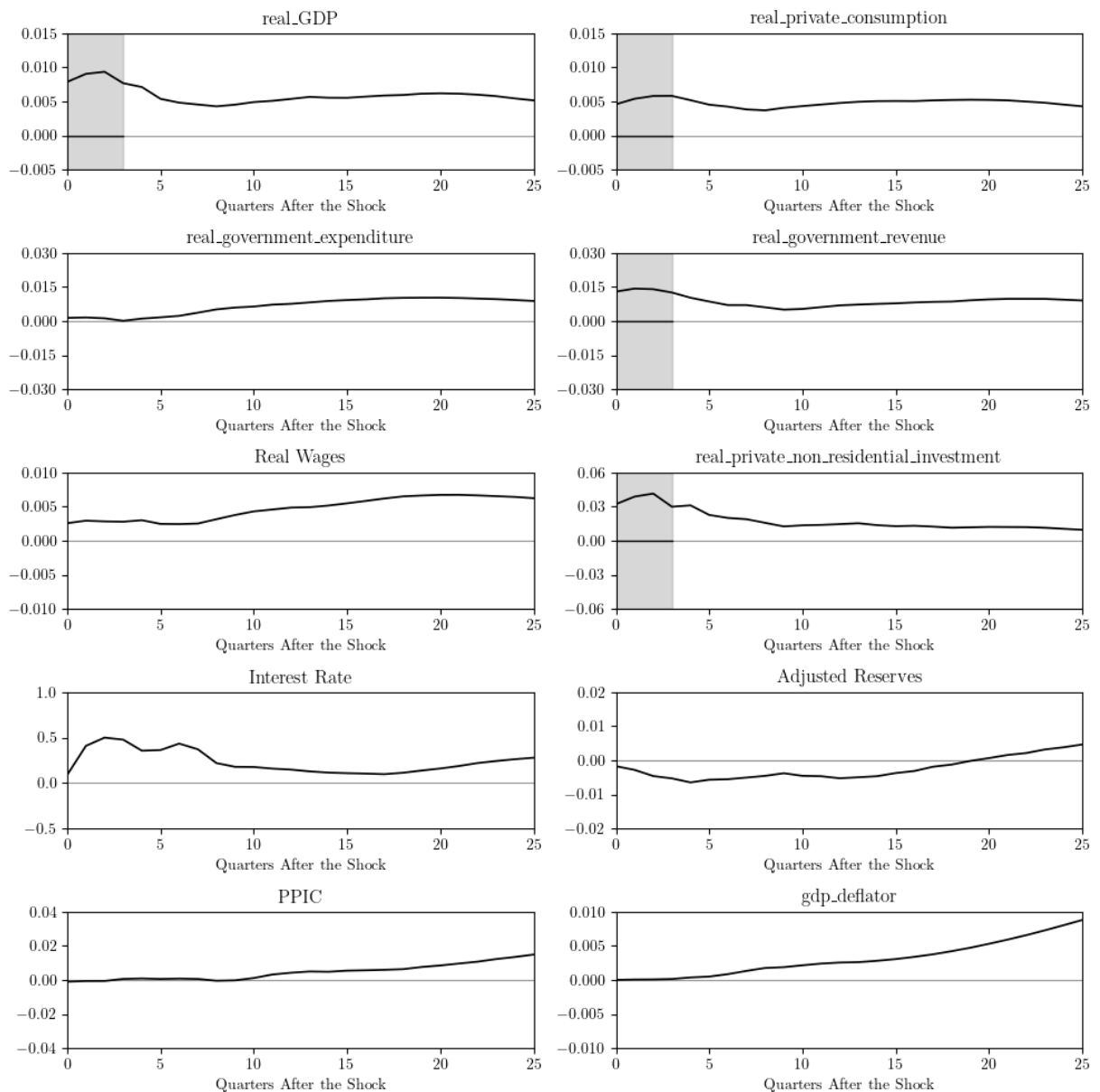
axes[0,0].axvspan(0, 3, color='gray', alpha=0.3)
axes[0,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[2,1].axvspan(0, 3, color='gray', alpha=0.3)

axes[0,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[0,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[2,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.005, 0.015)
axes[0,1].set_ylim(-0.005, 0.015)
axes[1,0].set_ylim(-0.03, 0.03)
axes[1,1].set_ylim(-0.03, 0.03)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.06, 0.06)
axes[3,0].set_ylim(-0.5, 1)
axes[3,1].set_ylim(-0.02, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("business.pdf", format="PDF")
plt.show()
```

Monetary shocks

```
In [17]: positive = [6]
negative = [7, 8, 9]
periods = 3

q0 = get_q(10)

q_business_shocks = q_business_shocks / np.linalg.norm(q_business_shocks)

norm_constraint = NonlinearConstraint(lambda q: np.sum(q**2), lb=1.0, ub=1.0)
orth_constraint = LinearConstraint(q_business_shocks.reshape(1, -1), lb=0.0, ub=0.0)

res = minimize(
    fun=lambda q: psi(q, positive, negative, periods),
    x0=q0,
    method='trust-constr',
    constraints=[norm_constraint, orth_constraint],
```

```
options={'verbose': 2, 'maxiter': 1000, 'xtol': 1e-12}  
)  
  
q_monetary_shocks = res.x
```

niter	f evals	CG iter	obj func	tr radius	opt	c viol
1	11	0	+4.5168e+02	1.00e+00	5.26e+02	5.34e-02
2	22	1	+6.5099e-01	2.00e+00	2.03e+02	1.00e+00
3	33	3	-5.7562e+00	2.64e+00	4.92e+00	1.42e-01
4	44	5	-5.5061e+00	2.64e+00	4.89e+00	4.79e-03
5	55	7	-5.4858e+00	2.64e+00	5.41e+01	2.17e-05
6	66	10	-5.7037e+00	2.64e+00	4.71e+00	8.85e-04
7	88	13	-6.4159e+00	2.64e+00	3.97e+00	6.81e-05
8	110	16	-6.9469e+00	2.64e+00	3.47e+00	3.28e-05
9	132	20	-6.9469e+00	2.64e-01	3.47e+00	3.28e-05
10	154	23	-7.9303e+00	1.85e+00	2.58e+00	1.22e-03
11	176	27	-8.3496e+00	1.85e+00	2.11e+00	4.86e-03
12	187	31	-8.3642e+00	1.85e+00	2.23e+00	4.18e-03
13	209	35	-8.3642e+00	1.85e-01	2.23e+00	4.18e-03
14	220	38	-8.3565e+00	1.85e-01	2.23e+00	2.45e-05
15	231	41	-8.3655e+00	1.85e-01	2.23e+00	1.68e-05
16	253	43	-8.3655e+00	1.85e-02	2.23e+00	1.68e-05
17	275	44	-8.3655e+00	1.85e-03	2.23e+00	1.68e-05
18	286	45	-8.3729e+00	1.29e-02	2.22e+00	3.42e-06
19	308	48	-8.3789e+00	9.06e-02	2.22e+00	7.30e-09
20	330	51	-8.3839e+00	1.64e-01	2.21e+00	7.57e-08
21	352	52	-8.3839e+00	1.64e-02	2.21e+00	7.57e-08
22	374	56	-8.3839e+00	1.64e-03	2.21e+00	7.57e-08
23	385	57	-8.3905e+00	1.15e-02	2.21e+00	2.70e-06
24	396	58	-8.3910e+00	1.15e-02	2.21e+00	1.91e-08
25	407	59	-8.3916e+00	1.15e-02	2.21e+00	1.91e-08
26	418	60	-8.3921e+00	1.15e-02	2.21e+00	1.91e-08
27	429	61	-8.3927e+00	1.15e-02	2.21e+00	1.91e-08
28	451	62	-8.3927e+00	1.15e-03	2.21e+00	1.91e-08
29	473	63	-8.3927e+00	1.15e-04	2.21e+00	1.91e-08
30	495	64	-8.3927e+00	1.21e-05	2.21e+00	1.91e-08
31	506	65	-8.3927e+00	8.45e-05	2.21e+00	1.46e-10
32	517	66	-8.3927e+00	8.45e-05	2.21e+00	1.67e-11
33	528	67	-8.3928e+00	8.45e-05	2.21e+00	1.67e-11
34	539	68	-8.3928e+00	8.45e-05	2.21e+00	1.67e-11
35	550	69	-8.3928e+00	8.45e-05	2.20e+00	1.67e-11
36	572	70	-8.3928e+00	8.45e-06	2.20e+00	1.67e-11
37	583	71	-8.3928e+00	2.35e-05	2.20e+00	1.13e-11
38	594	72	-8.3928e+00	2.35e-05	2.20e+00	1.04e-11
39	605	73	-8.3928e+00	2.35e-05	2.20e+00	1.04e-11
40	616	74	-8.3928e+00	2.35e-05	2.20e+00	1.04e-11
41	627	75	-8.3929e+00	2.35e-05	2.20e+00	1.04e-11
42	638	76	-8.3929e+00	1.64e-04	2.20e+00	5.52e-10
43	660	77	-8.3929e+00	1.64e-05	2.20e+00	5.52e-10
44	671	78	-8.3930e+00	4.70e-05	2.20e+00	4.51e-11
45	693	79	-8.3930e+00	4.70e-06	2.20e+00	4.51e-11
46	704	80	-8.3930e+00	4.70e-06	2.20e+00	6.37e-14
47	715	81	-8.3930e+00	4.70e-06	2.20e+00	5.88e-14
48	726	82	-8.3930e+00	4.70e-06	2.20e+00	5.97e-14
49	737	83	-8.3930e+00	4.70e-06	2.20e+00	5.86e-14
50	759	84	-8.3930e+00	4.70e-07	2.20e+00	5.86e-14
51	770	85	-8.3930e+00	1.30e-06	2.20e+00	3.82e-14
52	781	86	-8.3930e+00	1.30e-06	2.18e+01	3.33e-14
53	803	88	-8.3930e+00	1.30e-07	2.18e+01	3.33e-14
54	825	89	-8.3930e+00	6.52e-08	2.18e+01	3.33e-14

```
d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize\_differentiable_functions.py:728: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.
```

```
self.H.update(delta_x, delta_g)
```

55	847	90	-8.3930e+00	3.26e-08	2.18e+01	3.33e-14
56	869	91	-8.3930e+00	1.63e-08	2.18e+01	3.33e-14
57	891	92	-8.3930e+00	8.15e-09	2.18e+01	3.33e-14
58	913	93	-8.3930e+00	4.08e-09	2.18e+01	3.33e-14

```
d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize\_differentiable_functions.py:376: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.
```

```
self.H.update(self.x - self.x_prev, self.g - self.g_prev)
```

59	935	94	-8.3930e+00	2.04e-09	2.18e+01	3.33e-14
60	957	95	-8.3930e+00	1.02e-09	2.18e+01	3.33e-14
61	979	96	-8.3930e+00	5.10e-10	2.18e+01	3.33e-14
62	1001	97	-8.3930e+00	2.55e-10	2.18e+01	3.33e-14
63	1023	98	-8.3930e+00	1.27e-10	2.18e+01	3.33e-14
64	1045	99	-8.3930e+00	6.37e-11	2.18e+01	3.33e-14
65	1067	100	-8.3930e+00	3.18e-11	2.18e+01	3.33e-14
66	1089	101	-8.3930e+00	1.59e-11	2.18e+01	3.33e-14
67	1111	102	-8.3930e+00	7.96e-12	2.18e+01	3.33e-14
68	1133	103	-8.3930e+00	3.98e-12	2.18e+01	3.33e-14
69	1144	104	-8.3930e+00	1.99e-12	2.18e+01	3.33e-14
70	1166	105	-8.3930e+00	9.95e-13	2.18e+01	3.33e-14

```
`xtol` termination condition is satisfied.
```

```
Number of iterations: 70, function evaluations: 1166, CG iterations: 105, optimality: 2.18e+01, constraint violation: 3.33e-14, execution time: 3.7e+01 s.
```

```
In [18]: r_monetary = get_r_ja(var_res, q_monetary_shocks, 25)
         print(r_monetary.shape)
```

```
(26, 10)
```

Plot Monetary Shocks

```
In [19]: fig, axes = plt.subplots(5,2, figsize=(10, 10))
         for i in range(5):
             for j in range(2):
                 axes[i,j].plot(r_monetary[:, i*2+j], color = "black", lw=1.25)
                 axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
                 axes[i,j].set_xlim(0,25)
                 axes[i,j].set_title(names[i*2+j], fontsize=12)
                 axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

         axes[3,0].axvspan(0, 3, color='gray', alpha=0.3)
         axes[3,1].axvspan(0, 3, color='gray', alpha=0.3)
         axes[4,0].axvspan(0, 3, color='gray', alpha=0.3)
         axes[4,1].axvspan(0, 3, color='gray', alpha=0.3)

         axes[3,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
         axes[3,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
         axes[4,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
```

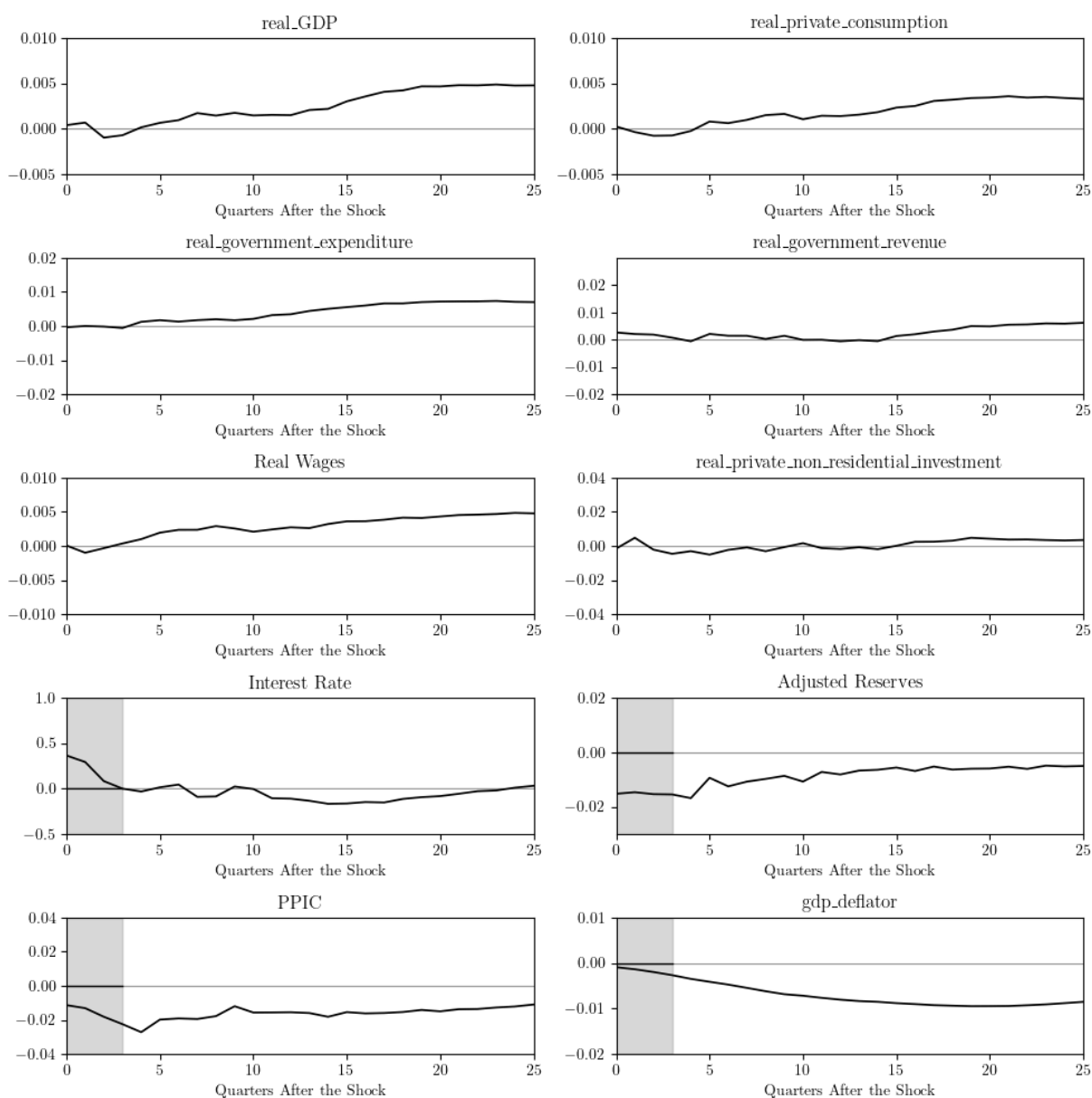
```

axes[4,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.005, 0.01)
axes[0,1].set_ylim(-0.005, 0.01)
axes[1,0].set_ylim(-0.02, 0.02)
axes[1,1].set_ylim(-0.02, 0.03)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 1)
axes[3,1].set_ylim(-0.03, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.02, 0.01)

axes[1,1].set_yticks([-0.02,-0.01,0,0.01,0.02])
plt.tight_layout()
plt.savefig("monetary.pdf", format="PDF")
plt.show()

```



Fiscal revenue shocks

```
In [20]: positive = [3]
negative = []
periods = 3

q0 = get_q(10)

q_business_shocks = q_business_shocks / np.linalg.norm(q_business_shocks)
q_monetary_shocks = q_monetary_shocks / np.linalg.norm(q_monetary_shocks)

norm_constraint = NonlinearConstraint(lambda q: np.sum(q**2), lb=1.0, ub=1.0)
orth_constraint1 = LinearConstraint(q_business_shocks.reshape(1, -1), lb=0.0, ub=0.0)
orth_constraint2 = LinearConstraint(q_monetary_shocks.reshape(1, -1), lb=0.0, ub=0.0)

res = minimize(
    fun=lambda q: psi(q, positive, negative, periods),
    x0=q0,
    method='trust-constr',
    constraints=[norm_constraint, orth_constraint1, orth_constraint2],
    options={'verbose': 2, 'maxiter': 1000, 'xtol': 1e-12}
)

q_revenue_shocks = res.x
```

niter	f evals	CG iter	obj func	tr radius	opt	c viol
1	11	0	+3.9525e+01	1.00e+00	1.64e+02	7.91e-02
2	22	1	-1.9198e+00	1.00e+00	1.04e+00	1.00e+00
3	33	2	-1.2287e+00	2.69e+00	1.30e+00	1.48e-01
4	44	3	-1.1056e+00	2.69e+00	1.34e+00	5.55e-03
5	55	4	-1.1099e+00	2.69e+00	1.34e+00	2.83e-05
6	66	5	-1.1194e+00	2.69e+00	1.33e+00	3.53e-05
7	77	6	-1.1287e+00	2.69e+00	1.32e+00	3.46e-05
8	88	7	-1.1380e+00	2.69e+00	1.31e+00	3.39e-05
9	99	8	-1.1471e+00	2.69e+00	1.30e+00	3.33e-05
10	121	9	-1.1471e+00	2.69e-01	1.30e+00	3.33e-05
11	143	10	-1.5298e+00	1.89e+00	8.08e-01	1.32e-03
12	154	11	-1.6022e+00	1.89e+00	6.99e-01	3.70e-03
13	165	13	-1.6379e+00	1.89e+00	6.38e-01	1.20e-03
14	176	14	-1.6686e+00	1.89e+00	6.20e-01	8.98e-04
15	187	16	-1.6930e+00	1.89e+00	6.07e-01	6.26e-04
16	209	18	-1.6930e+00	1.89e-01	6.07e-01	6.26e-04
17	231	19	-1.8446e+00	1.32e+00	4.53e-01	3.27e-04
18	253	21	-1.9509e+00	1.32e+00	7.97e-02	4.75e-03
19	264	23	-1.9515e+00	1.32e+00	1.11e-02	2.77e-03
20	275	24	-1.9488e+00	1.32e+00	1.08e-02	1.97e-06
21	286	25	-1.9488e+00	1.32e+00	1.04e-02	7.64e-08
22	297	26	-1.9488e+00	1.32e+00	1.00e-02	1.12e-07
23	308	27	-1.9488e+00	1.32e+00	9.62e-03	1.03e-07
24	330	28	-1.9488e+00	1.32e+00	1.31e-07	8.92e-10
25	341	30	-1.9488e+00	1.32e+00	1.78e-08	1.51e-14
26	352	33	-1.9488e+00	1.32e+00	2.41e-08	6.94e-18
27	374	37	-1.9488e+00	1.32e-01	2.41e-08	6.94e-18
28	396	44	-1.9488e+00	1.32e-02	2.41e-08	6.94e-18

d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize_differentiable_functions.py:376: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.

self.H.update(self.x - self.x_prev, self.g - self.g_prev)

d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize_differentiable_functions.py:728: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.

self.H.update(delta_x, delta_g)

29	418	50	-1.9488e+00	1.32e-03	2.41e-08	6.94e-18
30	440	55	-1.9488e+00	1.32e-03	3.16e-08	0.00e+00
31	462	60	-1.9488e+00	1.32e-04	3.16e-08	0.00e+00
32	484	65	-1.9488e+00	1.32e-05	3.16e-08	0.00e+00
33	506	70	-1.9488e+00	1.32e-06	3.16e-08	0.00e+00
34	528	75	-1.9488e+00	1.32e-07	3.16e-08	0.00e+00
35	550	80	-1.9488e+00	1.32e-08	3.16e-08	0.00e+00
36	572	85	-1.9488e+00	1.74e-09	3.16e-08	0.00e+00
37	594	87	-1.9488e+00	7.98e-10	3.16e-08	0.00e+00
38	616	90	-1.9488e+00	3.99e-10	3.16e-08	0.00e+00
39	627	92	-1.9488e+00	2.00e-10	3.16e-08	0.00e+00
40	649	94	-1.9488e+00	9.98e-11	3.16e-08	0.00e+00
41	671	96	-1.9488e+00	4.55e-11	3.16e-08	0.00e+00
42	693	98	-1.9488e+00	2.18e-11	3.16e-08	0.00e+00
43	715	99	-1.9488e+00	1.09e-11	3.16e-08	0.00e+00
44	737	100	-1.9488e+00	5.45e-12	3.16e-08	0.00e+00
45	759	101	-1.9488e+00	2.73e-12	3.16e-08	0.00e+00
46	781	104	-1.9488e+00	1.36e-12	3.16e-08	0.00e+00
47	792	107	-1.9488e+00	6.81e-13	3.16e-08	0.00e+00

`xtol` termination condition is satisfied.

Number of iterations: 47, function evaluations: 792, CG iterations: 107, optimality: 3.16e-08, constraint violation: 0.00e+00, execution time: 6.3 s.

```
In [21]: r_revenue = get_r_ja(var_res, q_revenue_shocks, 25)
         print(r_revenue.shape)
```

(26, 10)

```
In [22]: fig, axes = plt.subplots(5,2, figsize=(10, 10))
         for i in range(5):
             for j in range(2):
                 axes[i,j].plot(r_revenue[:, i*2+j], color = "black", lw=1.25)
                 axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
                 axes[i,j].set_xlim(0,25)
                 axes[i,j].set_title(names[i*2+j], fontsize=12)
                 axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

         axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)

         axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

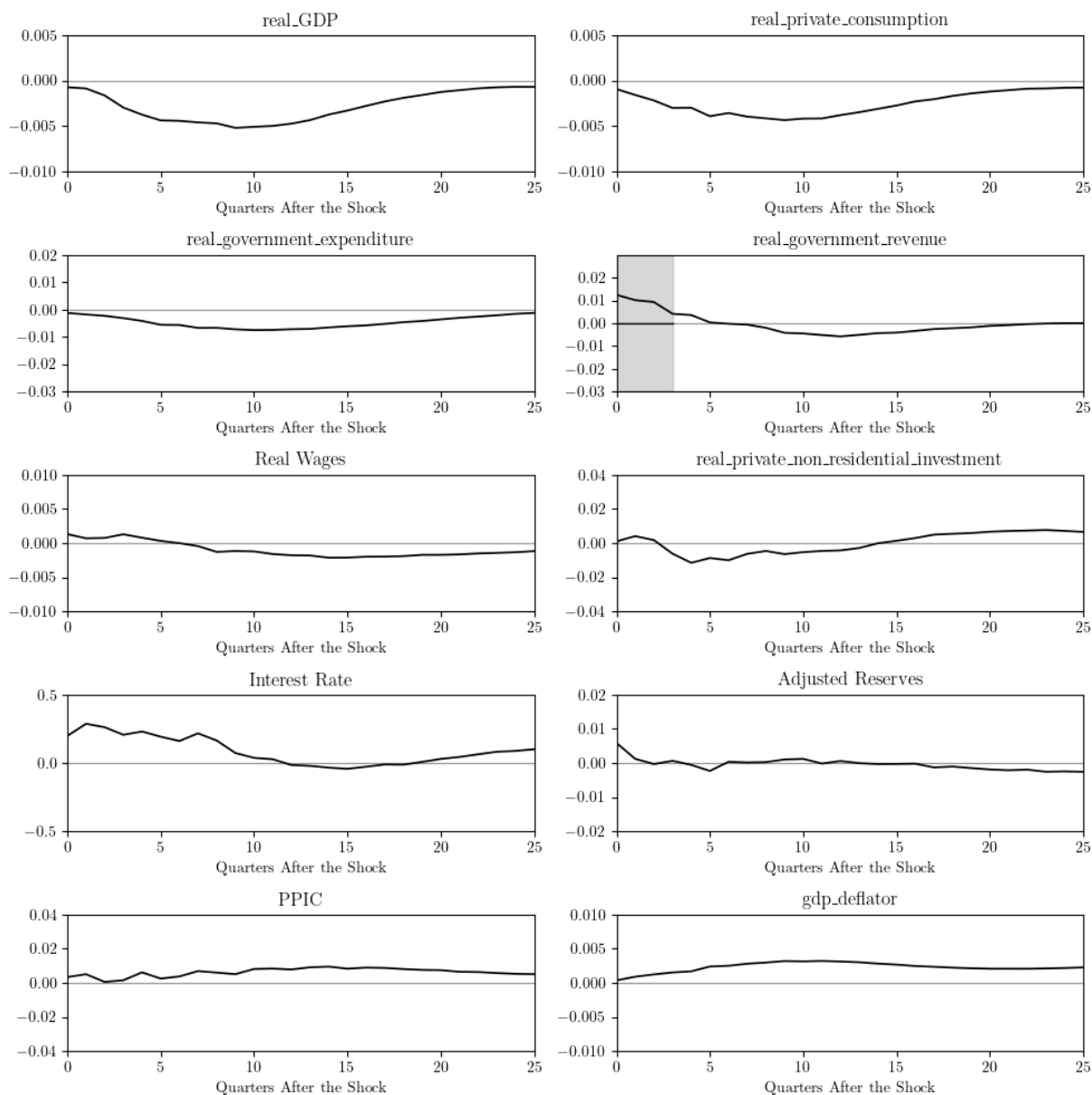
         axes[0,0].set_ylim(-0.01, 0.005)
         axes[0,1].set_ylim(-0.01, 0.005)
         axes[1,0].set_ylim(-0.03, 0.02)
         axes[1,1].set_ylim(-0.03, 0.03)
         axes[2,0].set_ylim(-0.01, 0.01)
         axes[2,1].set_ylim(-0.04, 0.04)
         axes[3,0].set_ylim(-0.5, 0.5)
         axes[3,1].set_ylim(-0.02, 0.02)
         axes[4,0].set_ylim(-0.04, 0.04)
         axes[4,1].set_ylim(-0.01, 0.01)

         axes[1,0].set_yticks([-0.03,-0.02,-0.01,0,0.01,0.02])
         axes[1,1].set_yticks([-0.03,-0.02,-0.01,0,0.01,0.02])
```



```
axes[3,0].set_yticks([-0.5,0,0.5])

plt.tight_layout()
plt.savefig("revenue.pdf", format="PDF")
plt.show()
```



Fiscal spending shocks

```
In [23]: positive = [2]
negative = []
periods = 3

q_business_shocks = q_business_shocks / np.linalg.norm(q_business_shocks)
q_monetary_shocks = q_monetary_shocks / np.linalg.norm(q_monetary_shocks)

norm_constraint = NonlinearConstraint(lambda q: np.sum(q**2), lb=1.0, ub=1.0)
orth_constraint1 = LinearConstraint(q_business_shocks.reshape(1, -1), lb=0.0, ub=0.0)
orth_constraint2 = LinearConstraint(q_monetary_shocks.reshape(1, -1), lb=0.0, ub=0.0)
```

```

q0 = get_q(10)

res = minimize(
    fun=lambda q: psi(q, positive, negative, periods),
    x0=q0,
    method='trust-constr',
    constraints=[norm_constraint, orth_constraint1, orth_constraint2],
    options={'verbose': 2, 'maxiter': 1000, 'xtol': 1e-12}
)

q_spending_shocks = res.x

```

niter	f evals	CG iter	obj func	tr radius	opt	c viol
1	11	0	-1.3921e+00	1.00e+00	2.28e+00	4.55e-01
2	22	1	-3.6936e+00	7.00e+00	9.77e-01	1.00e+00
3	33	2	-4.9084e+00	7.77e+00	2.35e-01	1.23e+00
4	44	3	-3.6163e+00	7.77e+00	5.43e-02	1.96e-01
5	55	4	-3.3251e+00	7.77e+00	5.87e-02	1.14e-02
6	66	5	-3.3098e+00	7.77e+00	8.68e-03	1.20e-03
7	77	6	-3.3079e+00	7.77e+00	1.85e-05	1.99e-05
8	88	7	-3.3079e+00	7.77e+00	1.43e-07	1.89e-10
9	99	8	-3.3079e+00	7.77e+00	3.67e-08	6.44e-15
10	110	9	-3.3079e+00	7.77e+00	5.59e-08	6.66e-16
11	121	11	-3.3079e+00	7.77e+00	7.00e-08	1.33e-15
12	132	13	-3.3079e+00	7.77e+00	4.05e-08	1.11e-15
13	154	15	-3.3079e+00	7.77e-01	4.05e-08	1.11e-15
14	176	17	-3.3079e+00	7.77e-01	5.70e-08	2.22e-16
15	198	20	-3.3079e+00	7.77e-02	5.70e-08	2.22e-16
16	220	23	-3.3079e+00	7.77e-03	5.70e-08	2.22e-16
17	242	27	-3.3079e+00	7.77e-04	5.70e-08	2.22e-16
18	264	30	-3.3079e+00	7.77e-05	5.70e-08	2.22e-16

d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize_differentiable_functions.py:728: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.

```
self.H.update(delta_x, delta_g)
```

19	286	33	-3.3079e+00	7.77e-06	5.70e-08	2.22e-16
20	308	38	-3.3079e+00	7.77e-07	5.70e-08	2.22e-16
21	330	43	-3.3079e+00	7.77e-08	5.70e-08	2.22e-16
22	352	48	-3.3079e+00	1.57e-08	5.70e-08	2.22e-16
23	374	50	-3.3079e+00	7.62e-09	5.70e-08	2.22e-16
24	385	52	-3.3079e+00	1.52e-08	8.82e-08	2.22e-16
25	407	54	-3.3079e+00	4.23e-09	8.82e-08	2.22e-16
26	429	55	-3.3079e+00	2.12e-09	8.82e-08	2.22e-16

d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize_differentiable_functions.py:376: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.

```
self.H.update(self.x - self.x_prev, self.g - self.g_prev)
```

27	451	57	-3.3079e+00	6.05e-10	8.82e-08	2.22e-16
28	473	59	-3.3079e+00	1.22e-10	8.82e-08	2.22e-16
29	495	61	-3.3079e+00	1.97e-11	8.82e-08	2.22e-16
30	517	63	-3.3079e+00	5.21e-12	8.82e-08	2.22e-16
31	539	65	-3.3079e+00	5.52e-13	8.82e-08	2.22e-16

`xtol` termination condition is satisfied.

Number of iterations: 31, function evaluations: 539, CG iterations: 65, optimality: 8.82e-08, constraint violation: 2.22e-16, execution time: 4.3 s.

```
In [24]: r_spending = get_r_ja(var_res, q_spending_shocks, 25)
         print(r_spending.shape)
```

(26, 10)

```
In [25]: fig, axes = plt.subplots(5,2, figsize=(10, 10))
         for i in range(5):
             for j in range(2):
                 axes[i,j].plot(r_spending[:, i*2+j], color = "black", lw=1.25)
                 axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
                 axes[i,j].set_xlim(0,25)
                 axes[i,j].set_title(names[i*2+j], fontsize=12)
                 axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

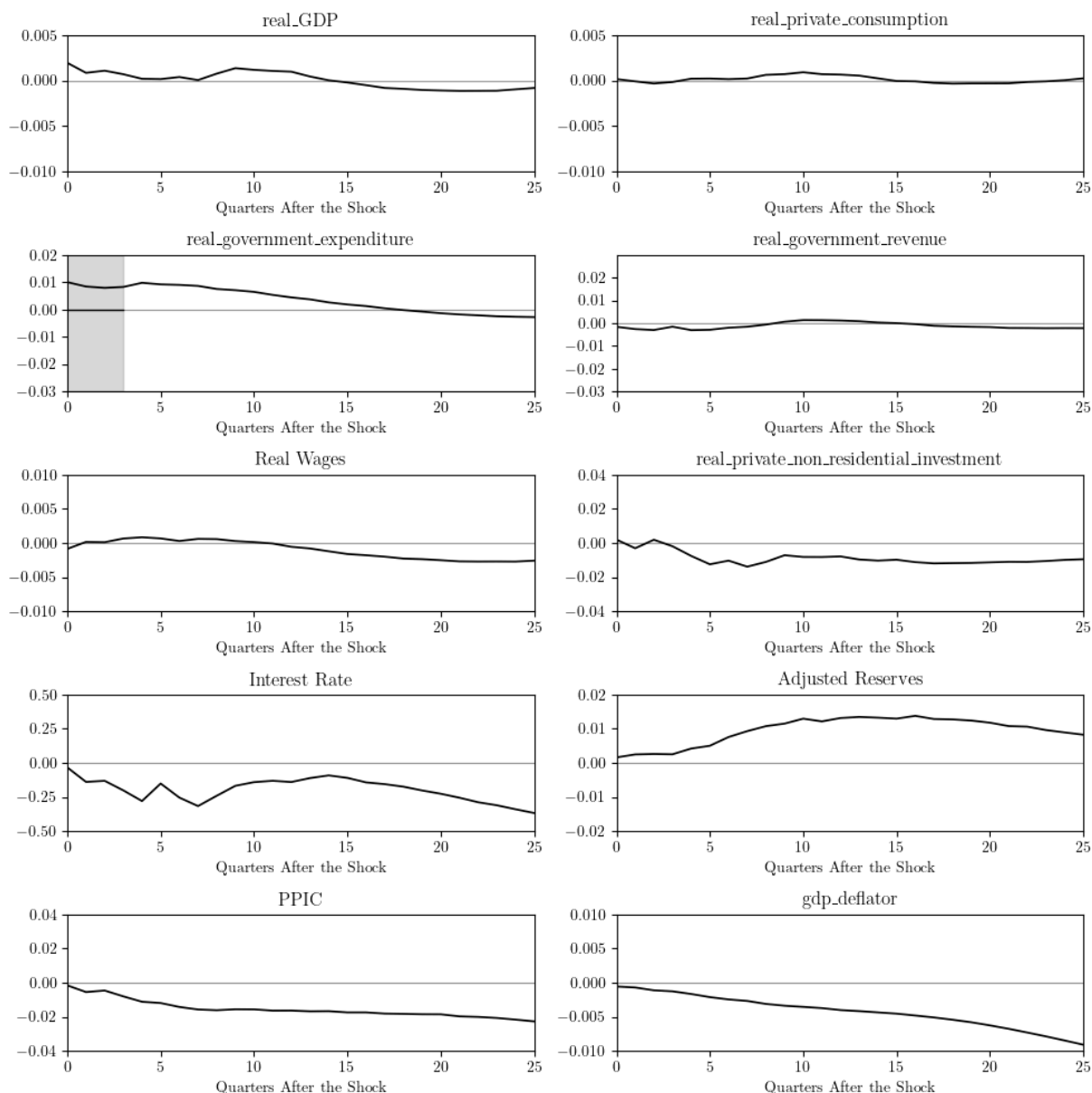
         axes[1,0].axvspan(0, 3, color='gray', alpha=0.3)

         axes[1,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

         axes[0,0].set_ylim(-0.01, 0.005)
         axes[0,1].set_ylim(-0.01, 0.005)
         axes[1,0].set_ylim(-0.03, 0.02)
         axes[1,1].set_ylim(-0.03, 0.03)
         axes[2,0].set_ylim(-0.01, 0.01)
         axes[2,1].set_ylim(-0.04, 0.04)
         axes[3,0].set_ylim(-0.5, 0.5)
         axes[3,1].set_ylim(-0.02, 0.02)
         axes[4,0].set_ylim(-0.04, 0.04)
         axes[4,1].set_ylim(-0.01, 0.01)

         axes[1,0].set_yticks([-0.03,-0.02,-0.01,0, 0.01 ,0.02])
         axes[1,1].set_yticks([-0.03,-0.02,-0.01,0, 0.01 ,0.02])

         plt.tight_layout()
         plt.savefig("spending.pdf", format="PDF")
         plt.show()
```



Extension-1 Zero-restriction

Define Zero-restriction psi function

```
In [26]: def zero_psi(q, positive, negative, periods_start, periods_end):
    positive_part = 0
    negative_part = 0
    if len(positive) > 0:
        for positive_index in positive:
            for period in range(periods_start, periods_end + 1):
                std = np.sqrt(np.diag(sigma)[positive_index])
                inner_part = get_r_ja(var_res, q=q, horizon=10)[period, positive_index]
                positive_part += f(-1 * inner_part)

    if len(negative) > 0:
        for negative_index in negative:
            for period in range(periods_start, periods_end + 1):
```

```

        std = np.sqrt(np.diag(sigma)[negative_index])
        inner_part = get_r_ja(var_res, q=q, horizon=10)[period, negative_index]
        negative_part += f(inner_part)
    res = positive_part + negative_part
    return res

```

Zero-restriction for revenue shocks

```

In [27]: j = 3
         h_zero = 4
         R = np.vstack([var_res.irf(10).orth_irfs[k][j, :] for k in range(h_zero+1)])

         norm_constraint = NonlinearConstraint(lambda q: np.sum(q**2), lb=1.0, ub=1.0)

         orth_constraint1 = LinearConstraint(q_business_shocks.reshape(1, -1), lb=0.0, ub=0.0)
         orth_constraint2 = LinearConstraint(q_monetary_shocks.reshape(1, -1), lb=0.0, ub=0.0)

         zero_constraints = [LinearConstraint(R[i].reshape(1, -1), lb=0.0, ub=0.0) for i in range(h_zero)]

         q0 = get_q(10)

         positive = [3]
         negative = []
         periods_start = 4
         periods_end = 7

         res = minimize(
             fun=lambda q: zero_psi(q, positive, negative, periods_start, periods_end),
             x0=q0,
             method='trust-constr',
             constraints=[
                 norm_constraint,
                 orth_constraint1,
                 orth_constraint2,
                 *zero_constraints
             ],
             options={'verbose': 2, 'maxiter': 1000, 'xtol': 1e-12}
         )

         q_anticipated_revenue_shocks = res.x

```

niter	f evals	CG iter	obj func	tr radius	opt	c viol
1	11	0	-5.1372e-01	1.00e+00	4.91e-01	3.09e-01
2	22	1	-5.1372e-01	2.53e-01	4.91e-01	3.09e-01
3	33	2	-4.5711e-01	1.77e+00	4.06e-01	1.49e-01
4	44	3	-4.5711e-01	2.16e-01	4.06e-01	1.49e-01
5	55	4	-3.6510e-01	1.51e+00	4.65e-01	9.97e-02
6	66	5	-3.6510e-01	1.51e-01	4.65e-01	9.97e-02
7	77	6	-4.2366e-01	1.03e+00	4.28e-01	4.63e-02
8	88	7	-4.2366e-01	1.79e-01	4.28e-01	4.63e-02
9	99	8	-5.0566e-01	1.25e+00	3.81e-01	1.48e-02
10	110	9	-5.0566e-01	1.25e-01	3.81e-01	1.48e-02
11	121	11	-5.1731e-01	2.51e-01	3.86e-01	1.46e-02
12	132	13	-5.1731e-01	3.77e-02	3.86e-01	1.46e-02
13	143	14	-5.3637e-01	2.64e-01	3.74e-01	3.44e-03
14	154	16	-5.3637e-01	2.64e-02	3.74e-01	3.44e-03
15	165	17	-5.4999e-01	1.85e-01	3.66e-01	1.53e-03
16	176	19	-5.4999e-01	1.85e-02	3.66e-01	1.53e-03
17	187	20	-5.6161e-01	1.29e-01	3.60e-01	1.48e-03
18	198	21	-5.6161e-01	5.39e-02	3.60e-01	1.48e-03
19	209	22	-5.9502e-01	1.08e-01	3.44e-01	2.75e-03
20	220	23	-5.9502e-01	5.39e-02	3.44e-01	2.75e-03
21	231	24	-6.2651e-01	1.08e-01	3.27e-01	2.89e-03
22	242	25	-6.2651e-01	5.39e-02	3.27e-01	2.89e-03
23	253	26	-6.5639e-01	1.08e-01	3.14e-01	2.90e-03
24	264	27	-6.5639e-01	5.39e-02	3.14e-01	2.90e-03
25	275	28	-6.8456e-01	1.08e-01	3.04e-01	2.91e-03
26	286	29	-6.8456e-01	5.39e-02	3.04e-01	2.91e-03
27	297	30	-7.1090e-01	1.08e-01	2.92e-01	2.91e-03
28	308	31	-7.1090e-01	5.39e-02	2.92e-01	2.91e-03
29	319	32	-7.3531e-01	1.08e-01	2.80e-01	2.91e-03
30	330	33	-7.3531e-01	5.39e-02	2.80e-01	2.91e-03
31	341	34	-7.5766e-01	1.08e-01	2.67e-01	2.91e-03
32	352	35	-7.5766e-01	5.39e-02	2.67e-01	2.91e-03
33	363	36	-7.5766e-01	2.70e-02	2.67e-01	2.91e-03
34	374	38	-7.5315e-01	5.39e-02	2.71e-01	7.27e-04
35	385	39	-7.5315e-01	2.70e-02	2.71e-01	7.27e-04
36	396	40	-7.6333e-01	5.39e-02	2.65e-01	7.27e-04
37	407	41	-7.6333e-01	2.70e-02	2.65e-01	7.27e-04
38	418	42	-7.6333e-01	1.01e-02	2.65e-01	7.27e-04
39	429	44	-7.6210e-01	7.04e-02	2.66e-01	1.01e-04
40	440	45	-7.6210e-01	7.04e-03	2.66e-01	1.01e-04
41	451	46	-7.6462e-01	4.93e-02	2.64e-01	6.82e-05
42	462	47	-7.7277e-01	4.93e-02	4.03e+00	4.81e-04
43	484	49	-7.9129e-01	9.86e-02	4.83e+00	2.06e-06
44	506	51	-7.9129e-01	4.93e-02	4.83e+00	2.06e-06
45	528	52	-7.9129e-01	2.47e-02	4.83e+00	2.06e-06
46	550	53	-7.9129e-01	1.23e-02	4.83e+00	2.06e-06
47	572	54	-7.9129e-01	6.16e-03	4.83e+00	2.06e-06
48	594	55	-7.9129e-01	3.08e-03	4.83e+00	2.06e-06
49	616	56	-7.9129e-01	1.54e-03	4.83e+00	2.06e-06
50	638	57	-7.9129e-01	7.70e-04	4.83e+00	2.06e-06
51	660	58	-7.9129e-01	3.85e-04	4.83e+00	2.06e-06
52	682	59	-7.9129e-01	1.93e-04	4.83e+00	2.06e-06
53	704	60	-7.9129e-01	9.63e-05	4.83e+00	2.06e-06
54	726	61	-7.9129e-01	4.81e-05	4.83e+00	2.06e-06

55	737	62	-7.9127e-01	4.81e-05	4.83e+00	2.32e-09
56	759	63	-7.9127e-01	2.41e-05	4.83e+00	2.32e-09
57	781	64	-7.9127e-01	1.20e-05	4.83e+00	2.32e-09
58	803	65	-7.9127e-01	6.02e-06	4.83e+00	2.32e-09
59	825	66	-7.9127e-01	3.01e-06	4.83e+00	2.32e-09
60	847	67	-7.9127e-01	1.50e-06	4.83e+00	2.32e-09
61	869	68	-7.9127e-01	7.52e-07	4.83e+00	2.32e-09
62	891	69	-7.9127e-01	3.76e-07	4.83e+00	2.32e-09
63	913	70	-7.9127e-01	1.88e-07	4.83e+00	2.32e-09
64	935	71	-7.9127e-01	9.40e-08	4.83e+00	2.32e-09
65	957	72	-7.9127e-01	4.70e-08	4.83e+00	2.32e-09
66	968	73	-7.9127e-01	4.70e-08	4.83e+00	4.00e-15
67	990	74	-7.9127e-01	2.35e-08	4.83e+00	4.00e-15
68	1012	75	-7.9127e-01	1.18e-08	4.83e+00	4.00e-15

d:\Anaconda\envs\TimeSeriesCode\Lib\site-packages\scipy\optimize_differentiable_functions.py:728: UserWarning: delta_grad == 0.0. Check if the approximated function is linear. If the function is linear better results can be obtained by defining the Hessian as zero instead of using quasi-Newton approximations.

```
self.H.update(delta_x, delta_g)
```

69	1034	76	-7.9127e-01	5.88e-09	4.83e+00	4.00e-15
70	1056	77	-7.9127e-01	2.94e-09	4.83e+00	4.00e-15
71	1078	78	-7.9127e-01	1.47e-09	4.83e+00	4.00e-15
72	1100	79	-7.9127e-01	7.35e-10	4.83e+00	4.00e-15
73	1122	80	-7.9127e-01	3.67e-10	4.83e+00	4.00e-15
74	1144	81	-7.9127e-01	1.84e-10	4.83e+00	4.00e-15
75	1166	82	-7.9127e-01	9.18e-11	4.83e+00	4.00e-15
76	1188	83	-7.9127e-01	4.59e-11	4.83e+00	4.00e-15
77	1210	84	-7.9127e-01	2.30e-11	4.83e+00	4.00e-15
78	1232	85	-7.9127e-01	1.15e-11	4.83e+00	4.00e-15
79	1254	86	-7.9127e-01	5.74e-12	4.83e+00	4.00e-15
80	1276	87	-7.9127e-01	2.87e-12	4.83e+00	4.00e-15
81	1298	88	-7.9127e-01	1.43e-12	4.83e+00	4.00e-15
82	1320	89	-7.9127e-01	7.17e-13	4.83e+00	4.00e-15

`xtol` termination condition is satisfied.

Number of iterations: 82, function evaluations: 1320, CG iterations: 89, optimality: 4.83e+00, constraint violation: 4.00e-15, execution time: 1e+01 s.

```
In [28]: r_anticipated_revenue = get_r_ja(var_res, q_anticipated_revenue_shocks, 25)
print(r_anticipated_revenue.shape)
```

(26, 10)

```
In [29]: fig, axes = plt.subplots(5,2, figsize=(10, 10))
for i in range(5):
    for j in range(2):
        axes[i,j].plot(r_anticipated_revenue[:, i*2+j], color = "black", lw=1.25)
        axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
        axes[i,j].set_xlim(0,25)
        axes[i,j].set_title(names[i*2+j], fontsize=12)
        axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[1,1].axvspan(0, 7, color='gray', alpha=0.3)

axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
```

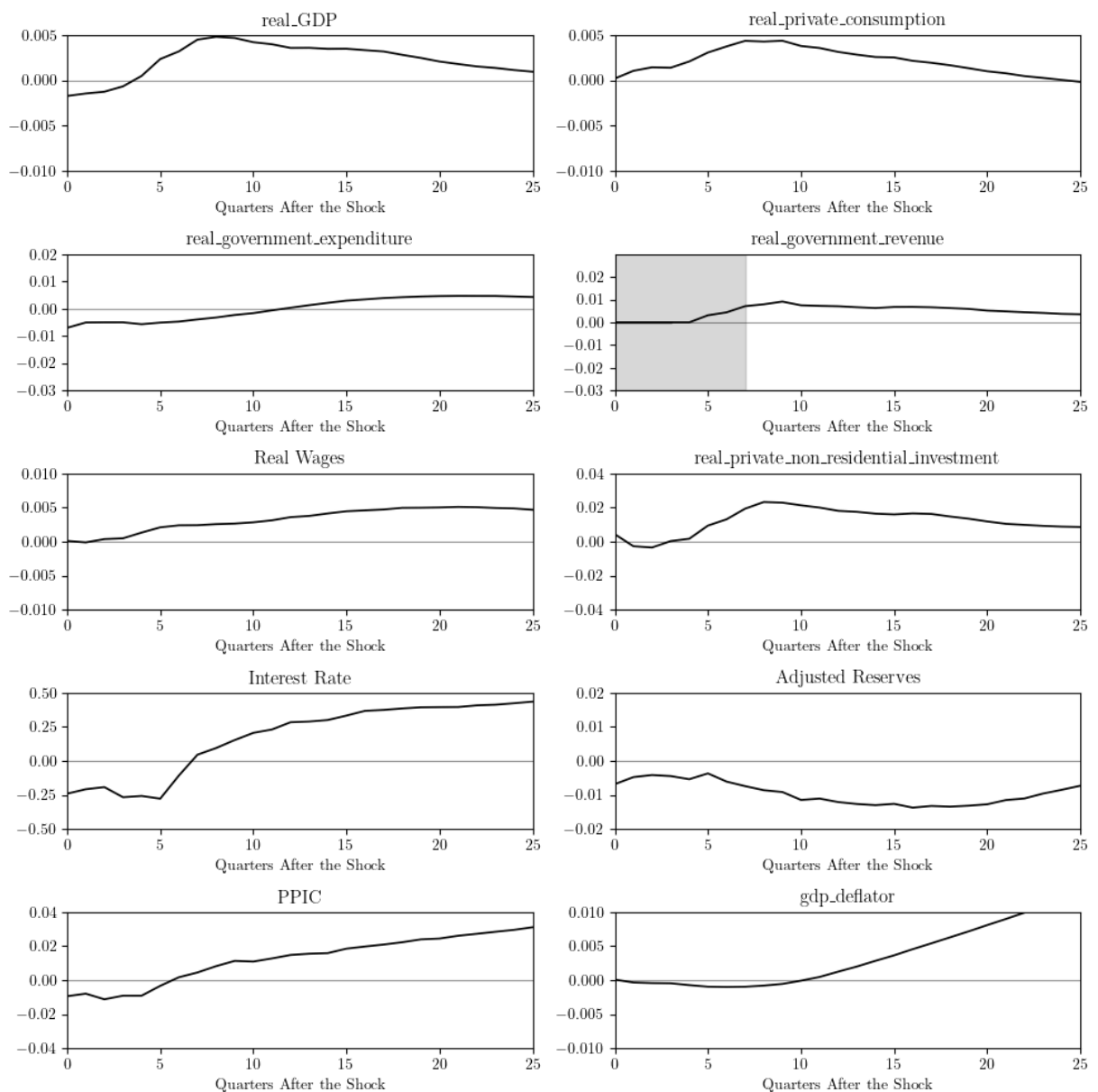
```

axes[0,0].set_ylim(-0.01, 0.005)
axes[0,1].set_ylim(-0.01, 0.005)
axes[1,0].set_ylim(-0.03, 0.02)
axes[1,1].set_ylim(-0.03, 0.03)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 0.5)
axes[3,1].set_ylim(-0.02, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.02,-0.01,0, 0.01 ,0.02])
axes[1,1].set_yticks([-0.03,-0.02,-0.01,0, 0.01 ,0.02])

plt.tight_layout()
plt.savefig("anticipated_revenue.pdf", format="PDF")
plt.show()

```



Zero-restriction for spending shocks

```
In [30]: j = 2
h_zero = 4
R = np.vstack([var_res.irf(10).orth_irfs[k][j, :] for k in range(h_zero+1)])

norm_constraint = NonlinearConstraint(lambda q: np.sum(q**2), lb=1.0, ub=1.0)

orth_constraint1 = LinearConstraint(q_business_shocks.reshape(1, -1), lb=0.0, ub=0.0)
orth_constraint2 = LinearConstraint(q_monetary_shocks.reshape(1, -1), lb=0.0, ub=0.0)

zero_constraints = [LinearConstraint(R[i].reshape(1, -1), lb=0.0, ub=0.0) for i in range(h_zero)]

q0 = get_q(10)

positive = [2]
negative = []
periods_start = 5
periods_end = 8

res = minimize(
    fun=lambda q: zero_psi(q, positive, negative, periods_start, periods_end),
    x0=q0,
    method='trust-constr',
    constraints=[
        norm_constraint,
        orth_constraint1,
        orth_constraint2,
        *zero_constraints
    ],
    options={'verbose': 2, 'maxiter': 1000, 'xtol': 1e-12, 'gtol': 1e-12}
)

q_anticipated_spending_shocks = res.x
```

niter	f evals	CG iter	obj func	tr radius	opt	c viol
1	11	0	+1.5483e+02	1.00e+00	5.65e+01	4.73e-01
2	22	1	+1.3207e+02	7.00e+00	6.62e+01	7.77e-01
3	33	2	+2.3134e+00	7.00e+00	2.39e+01	5.21e-01
4	44	4	-1.6783e-01	7.00e+00	8.63e-01	5.43e-02
5	55	5	-1.7630e-01	7.00e+00	8.65e-01	7.87e-04
6	66	6	-1.8173e-01	7.00e+00	8.66e-01	1.63e-05
7	77	8	-1.8757e-01	7.00e+00	8.68e-01	2.01e-05
8	99	10	-2.0192e-01	7.00e+00	8.76e-01	8.10e-06
9	121	11	-2.0192e-01	7.00e-01	8.76e-01	8.10e-06
10	143	12	-1.1093e+00	7.00e-01	5.81e-01	6.00e-02
11	165	13	-1.4265e+00	7.00e-01	4.90e-02	7.05e-02
12	176	14	-1.3869e+00	7.00e-01	1.09e-02	7.80e-03
13	187	15	-1.3819e+00	7.00e-01	1.60e-04	2.61e-04
14	198	16	-1.3817e+00	7.00e-01	1.41e-06	6.79e-08
15	209	17	-1.3817e+00	7.00e-01	7.37e-09	3.96e-12
16	220	18	-1.3817e+00	7.00e-01	1.07e-08	2.22e-16
17	231	19	-1.3817e+00	7.00e-01	2.08e-09	2.78e-17
18	253	21	-1.3817e+00	7.00e-02	2.08e-09	2.78e-17
19	264	23	-1.3817e+00	7.00e-02	9.50e-09	2.78e-17
20	275	25	-1.3817e+00	7.00e-02	5.79e-10	2.78e-17
21	297	27	-1.3817e+00	7.00e-03	5.79e-10	2.78e-17
22	308	28	-1.3817e+00	7.00e-04	5.79e-10	2.78e-17
23	330	30	-1.3817e+00	7.00e-05	5.79e-10	2.78e-17
24	352	32	-1.3817e+00	7.00e-06	5.79e-10	2.78e-17
25	374	33	-1.3817e+00	7.00e-07	5.79e-10	2.78e-17
26	396	34	-1.3817e+00	7.00e-08	5.79e-10	2.78e-17
27	418	35	-1.3817e+00	7.00e-09	5.79e-10	2.78e-17
28	440	36	-1.3817e+00	7.00e-10	5.79e-10	2.78e-17
29	462	37	-1.3817e+00	7.00e-11	5.79e-10	2.78e-17
30	484	38	-1.3817e+00	7.00e-12	5.79e-10	2.78e-17
31	506	39	-1.3817e+00	7.00e-13	5.79e-10	2.78e-17

`xtol` termination condition is satisfied.

Number of iterations: 31, function evaluations: 506, CG iterations: 39, optimality: 5.79e-10, constraint violation: 2.78e-17, execution time: 4.0 s.

```
In [31]: r_anticipated_spending = get_r_ja(var_res, q_anticipated_spending_shocks, 25)
         print(r_anticipated_spending.shape)
```

(26, 10)

```
In [32]: fig, axes = plt.subplots(5,2, figsize=(10, 10))
         for i in range(5):
             for j in range(2):
                 axes[i,j].plot(r_anticipated_spending[:, i*2+j], color = "black", lw=1.25)
                 axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
                 axes[i,j].set_xlim(0,25)
                 axes[i,j].set_title(names[i*2+j], fontsize=12)
                 axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

         axes[1,0].axvspan(0, 7, color='gray', alpha=0.3)

         axes[1,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
```

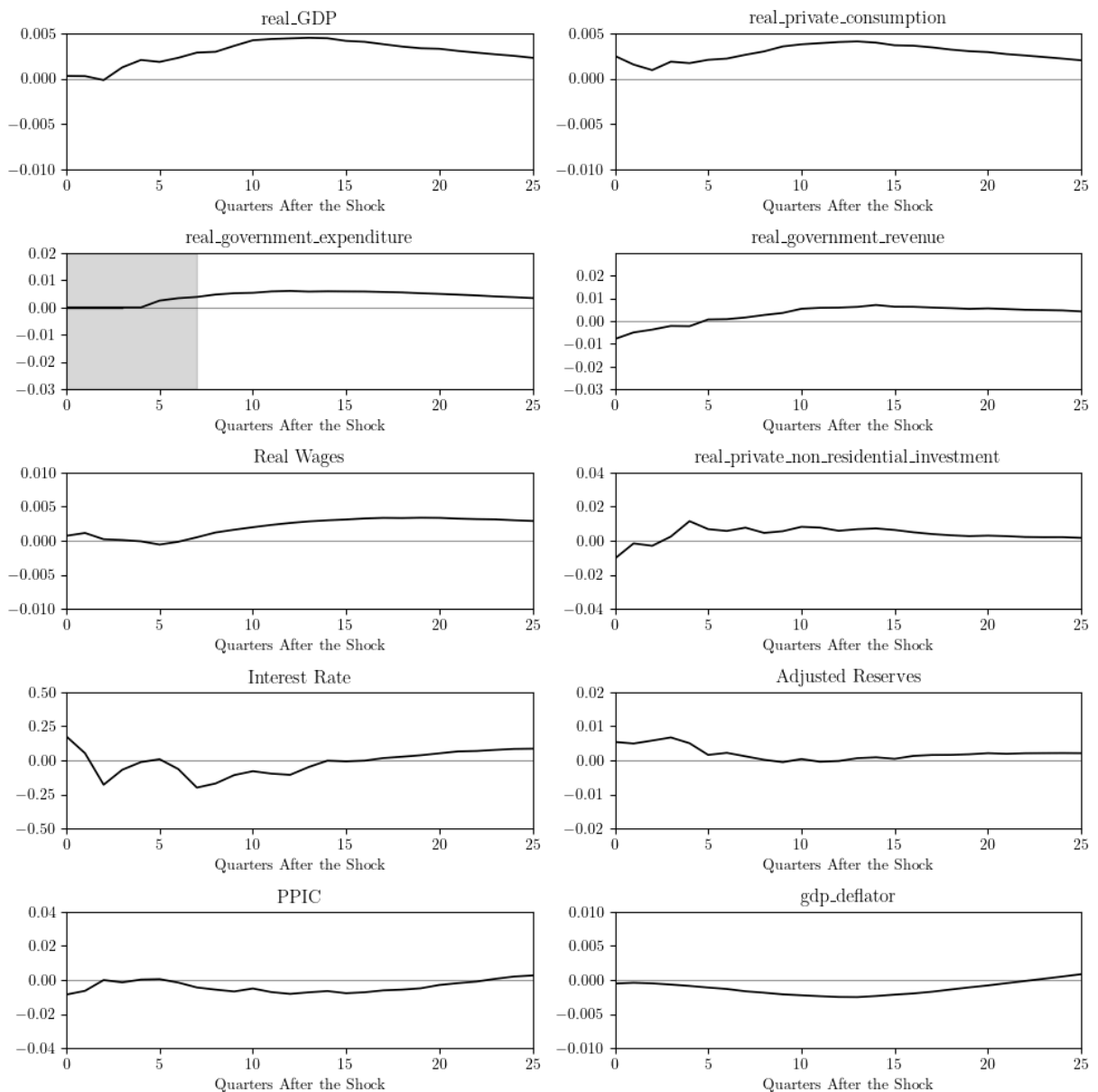
```

axes[0,0].set_ylim(-0.01, 0.005)
axes[0,1].set_ylim(-0.01, 0.005)
axes[1,0].set_ylim(-0.03, 0.02)
axes[1,1].set_ylim(-0.03, 0.03)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 0.5)
axes[3,1].set_ylim(-0.02, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.02,-0.01,0, 0.01 ,0.02])
axes[1,1].set_yticks([-0.03,-0.02,-0.01,0, 0.01 ,0.02])

plt.tight_layout()
plt.savefig("anticipated_spending.pdf", format="PDF")
plt.show()

```



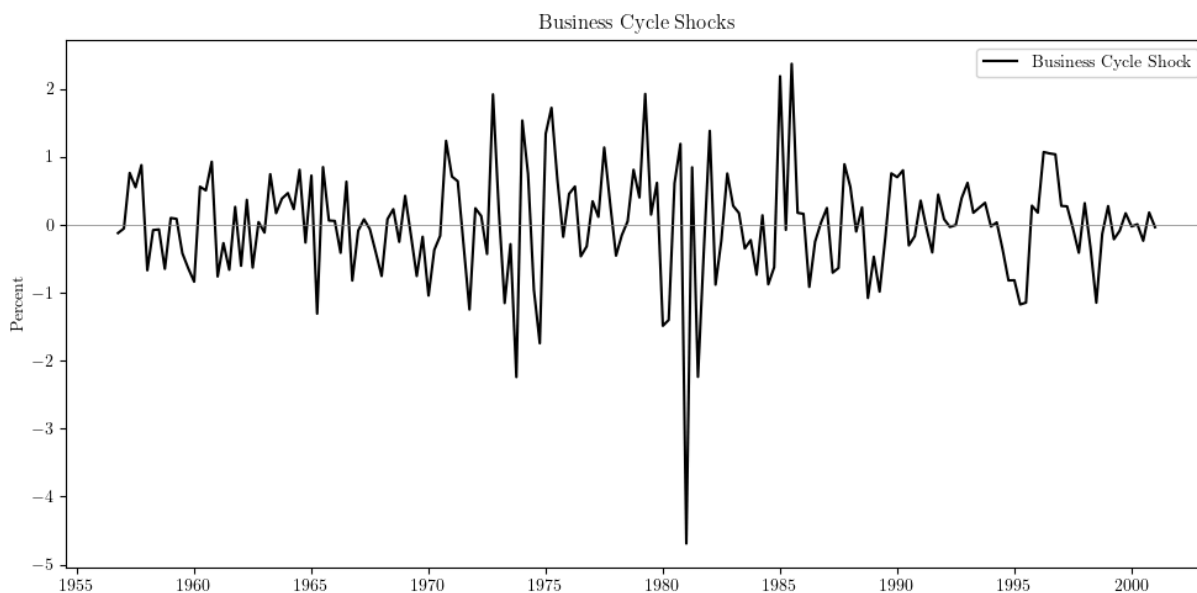
Extension 2 - Time trend of Shocks

```
In [33]: resid = var_res.resid
print(resid.shape)
```

(178, 10)

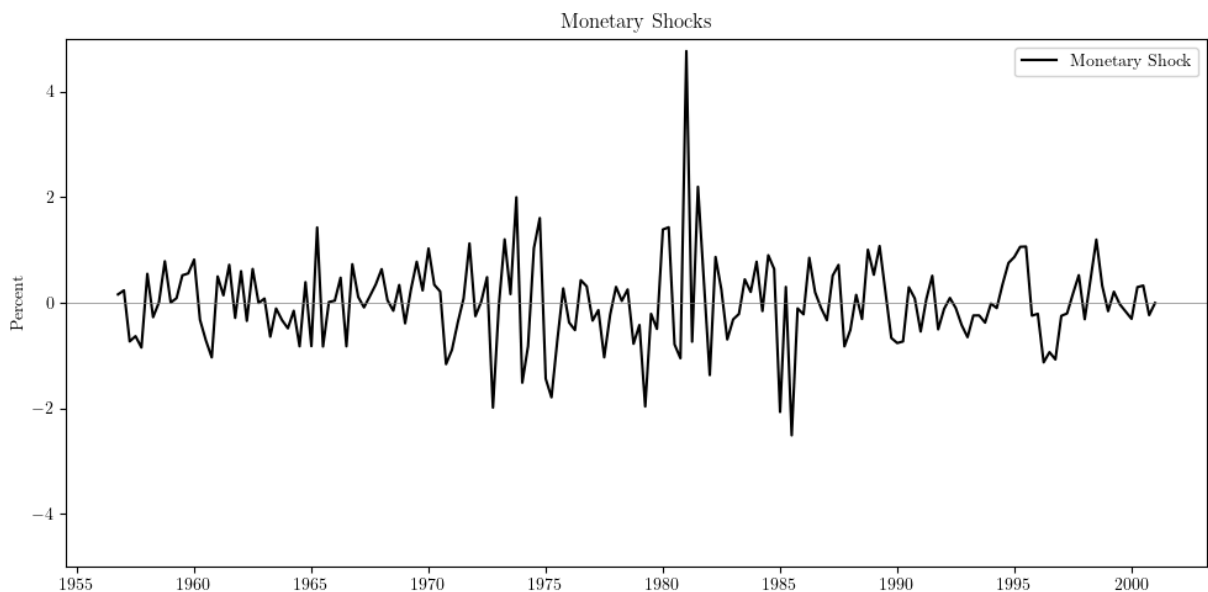
```
In [34]: q = q_business_shocks / np.sqrt(q_business_shocks.T @ sigma @ q_business_shocks)
eps_biz = resid @ q
```

```
In [35]: T = resid.shape[0]
dates = pd.date_range(start="1956-10-01", periods=T, freq="QS")
plt.figure(figsize=(10, 5))
plt.plot(dates, eps_biz, label="Business Cycle Shock", color='black')
plt.axhline(0, color='gray', lw=0.5)
#plt.ylim(-0.5,0.5)
plt.ylabel("Percent")
plt.title("Business Cycle Shocks")
plt.legend()
plt.tight_layout()
plt.show()
```



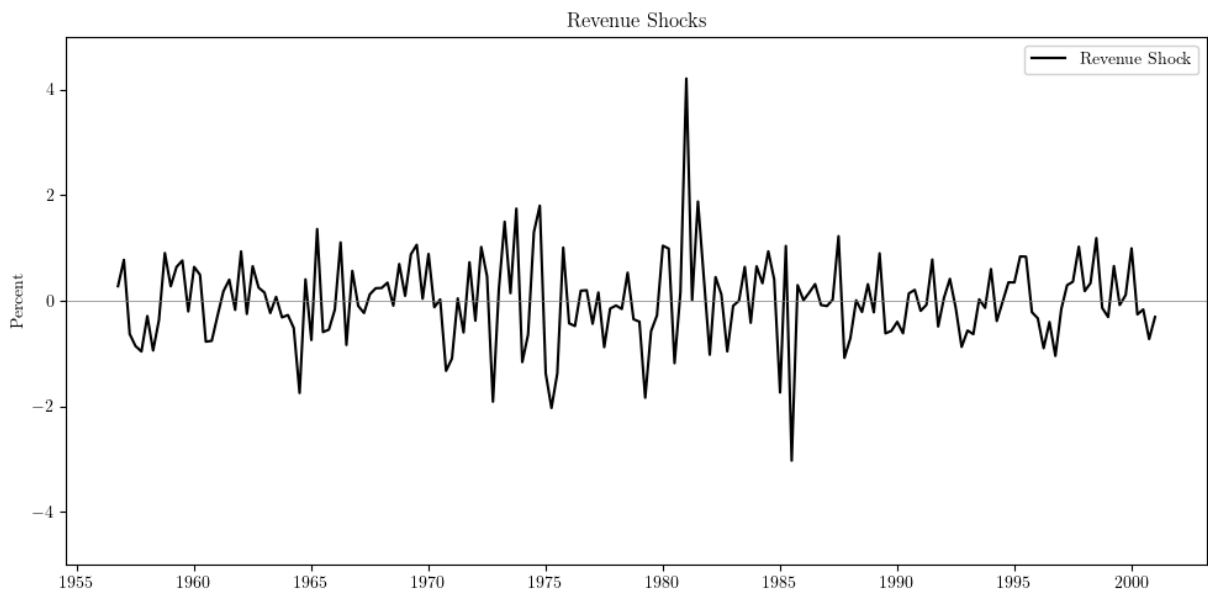
```
In [36]: q = q_monetary_shocks / np.sqrt(q_monetary_shocks.T @ sigma @ q_monetary_shocks)
eps_monetary= resid @ q
```

```
In [37]: T = resid.shape[0]
dates = pd.date_range(start="1956-10-01", periods=T, freq="QS")
plt.figure(figsize=(10, 5))
plt.plot(dates, eps_monetary, label="Monetary Shock", color='black')
plt.axhline(0, color='gray', lw=0.5)
plt.ylim(-5,5)
plt.ylabel("Percent")
plt.title("Monetary Shocks")
plt.legend()
plt.tight_layout()
plt.show()
```



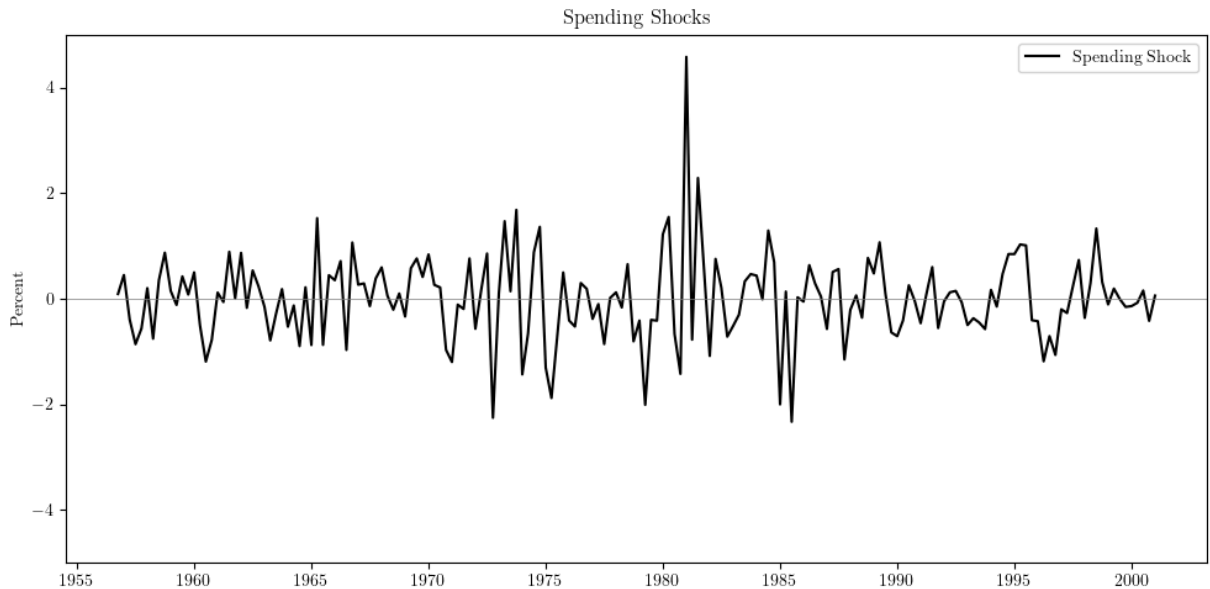
```
In [38]: q = q_revenue_shocks / np.sqrt(q_revenue_shocks.T @ sigma @ q_revenue_shocks)
eps_revenue = resid @ q
```

```
In [39]: T = resid.shape[0]
dates = pd.date_range(start="1956-10-01", periods=T, freq="QS")
plt.figure(figsize=(10, 5))
plt.plot(dates, eps_revenue, label="Revenue Shock", color='black')
plt.axhline(0, color='gray', lw=0.5)
plt.ylim(-5,5)
plt.ylabel("Percent")
plt.title("Revenue Shocks")
plt.legend()
plt.tight_layout()
plt.show()
```



```
In [40]: q = q_spending_shocks / np.sqrt(q_spending_shocks.T @ sigma @ q_spending_shocks)
eps_spending = resid @ q
```

```
In [41]: T = resid.shape[0]
dates = pd.date_range(start="1956-10-01", periods=T, freq="QS")
plt.figure(figsize=(10, 5))
plt.plot(dates, eps_spending, label="Spending Shock", color='black')
plt.axhline(0, color='gray', lw=0.5)
plt.ylim(-5,5)
plt.ylabel("Percent")
plt.title("Spending Shocks")
plt.legend()
plt.tight_layout()
plt.show()
```



Extension 3 - Confidence Interval

```
In [42]: lower, upper = var_res.irf_errband_mc(orth=True, steps=30, repl=3000, signif=0.32,
```

Business Cycle Shocks

```
In [43]: irf_lower = np.zeros((25 + 1, 10))
irf_upper = np.zeros((25 + 1, 10))

for k in range(25 + 1):
    for i in range(10):
        cholesky_irf_lower = lower[k, i, :]
        cholesky_irf_upper = upper[k, i, :]

        irf_lower[k, i] = np.dot(cholesky_irf_lower, q_business_shocks)
        irf_upper[k, i] = np.dot(cholesky_irf_upper, q_business_shocks)
print(irf_lower.shape, irf_upper.shape)
```

(26, 10) (26, 10)

```
In [44]: fig, axes = plt.subplots(5,2, figsize=(10, 10))
for i in range(5):
    for j in range(2):
```

```

axes[i,j].plot(r_business[:, i*2+j], color = "black", lw=1.25)
axes[i,j].plot(irf_lower[:, i*2+j], color = "red", ls="dashed", lw=1.25)
axes[i,j].plot(irf_upper[:, i*2+j], color = "red", ls="dashed", lw=1.25)
axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
axes[i,j].set_xlim(0,25)
axes[i,j].set_title(names[i*2+j], fontsize=12)
axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[0,0].axvspan(0, 3, color='gray', alpha=0.3)
axes[0,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[2,1].axvspan(0, 3, color='gray', alpha=0.3)

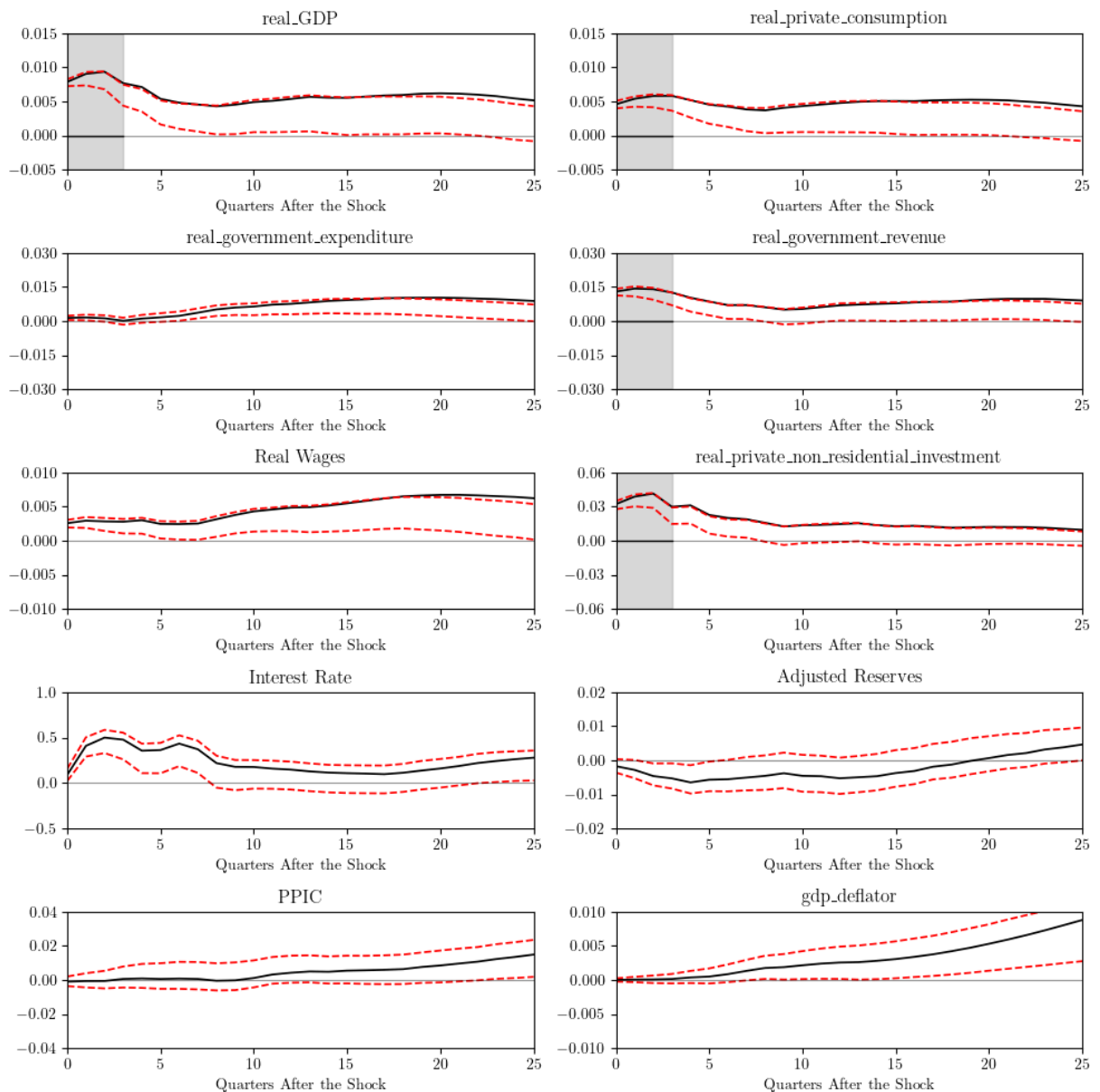
axes[0,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[0,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[2,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.005, 0.015)
axes[0,1].set_ylim(-0.005, 0.015)
axes[1,0].set_ylim(-0.03, 0.03)
axes[1,1].set_ylim(-0.03, 0.03)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.06, 0.06)
axes[3,0].set_ylim(-0.5, 1)
axes[3,1].set_ylim(-0.02, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("business_mc.pdf", format="PDF")
plt.show()

```



Monetary Shocks

```
In [45]: irf_lower = np.zeros((25 + 1, 10))
         irf_upper = np.zeros((25 + 1, 10))

         for k in range(25 + 1):
             for i in range(10):
                 cholesky_irf_lower = lower[k, i, :]
                 cholesky_irf_upper = upper[k, i, :]

                 irf_lower[k, i] = np.dot(cholesky_irf_lower, q_monetary_shocks)
                 irf_upper[k, i] = np.dot(cholesky_irf_upper, q_monetary_shocks)
         print(irf_lower.shape, irf_upper.shape)
```

(26, 10) (26, 10)

```
In [46]: fig, axes = plt.subplots(5, 2, figsize=(10, 10))
         for i in range(5):
```



```

for j in range(2):
    axes[i,j].plot(r_monetary[:, i*2+j], color = "black", lw=1.25)
    axes[i,j].plot(irf_lower[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].plot(irf_upper[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
    axes[i,j].set_xlim(0,25)
    axes[i,j].set_title(names[i*2+j], fontsize=12)
    axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[0,0].axvspan(0, 3, color='gray', alpha=0.3)
axes[0,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[2,1].axvspan(0, 3, color='gray', alpha=0.3)

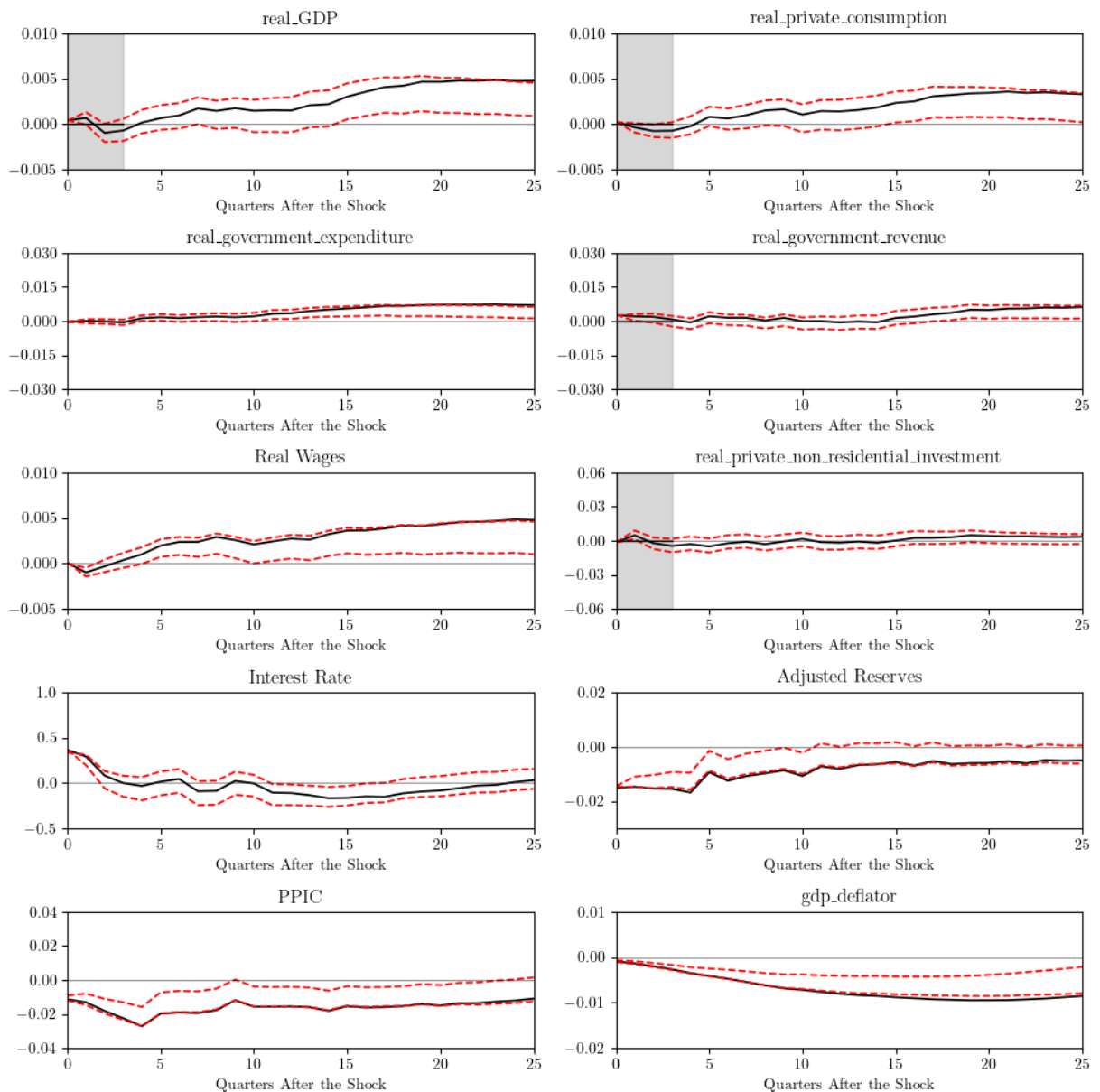
axes[0,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[0,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[2,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.005, 0.01)
axes[0,1].set_ylim(-0.005, 0.01)
axes[1,0].set_ylim(-0.02, 0.02)
axes[1,1].set_ylim(-0.02, 0.03)
axes[2,0].set_ylim(-0.005, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 1)
axes[3,1].set_ylim(-0.03, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.02, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("monetary_mc.pdf", format="PDF")
plt.show()

```



Revenue Shocks

```
In [47]: irf_lower = np.zeros((25 + 1, 10))
         irf_upper = np.zeros((25 + 1, 10))

         for k in range(25 + 1):
             for i in range(10):
                 cholesky_irf_lower = lower[k, i, :]
                 cholesky_irf_upper = upper[k, i, :]

                 irf_lower[k, i] = np.dot(cholesky_irf_lower, q_revenue_shocks)
                 irf_upper[k, i] = np.dot(cholesky_irf_upper, q_revenue_shocks)
         print(irf_lower.shape, irf_upper.shape)
```

(26, 10) (26, 10)

```
In [48]: fig, axes = plt.subplots(5, 2, figsize=(10, 10))
         for i in range(5):
```

```

for j in range(2):
    axes[i,j].plot(r_revenue[:, i*2+j], color = "black", lw=1.25)
    axes[i,j].plot(irf_lower[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].plot(irf_upper[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
    axes[i,j].set_xlim(0,25)
    axes[i,j].set_title(names[i*2+j], fontsize=12)
    axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[0,0].axvspan(0, 3, color='gray', alpha=0.3)
axes[0,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[2,1].axvspan(0, 3, color='gray', alpha=0.3)

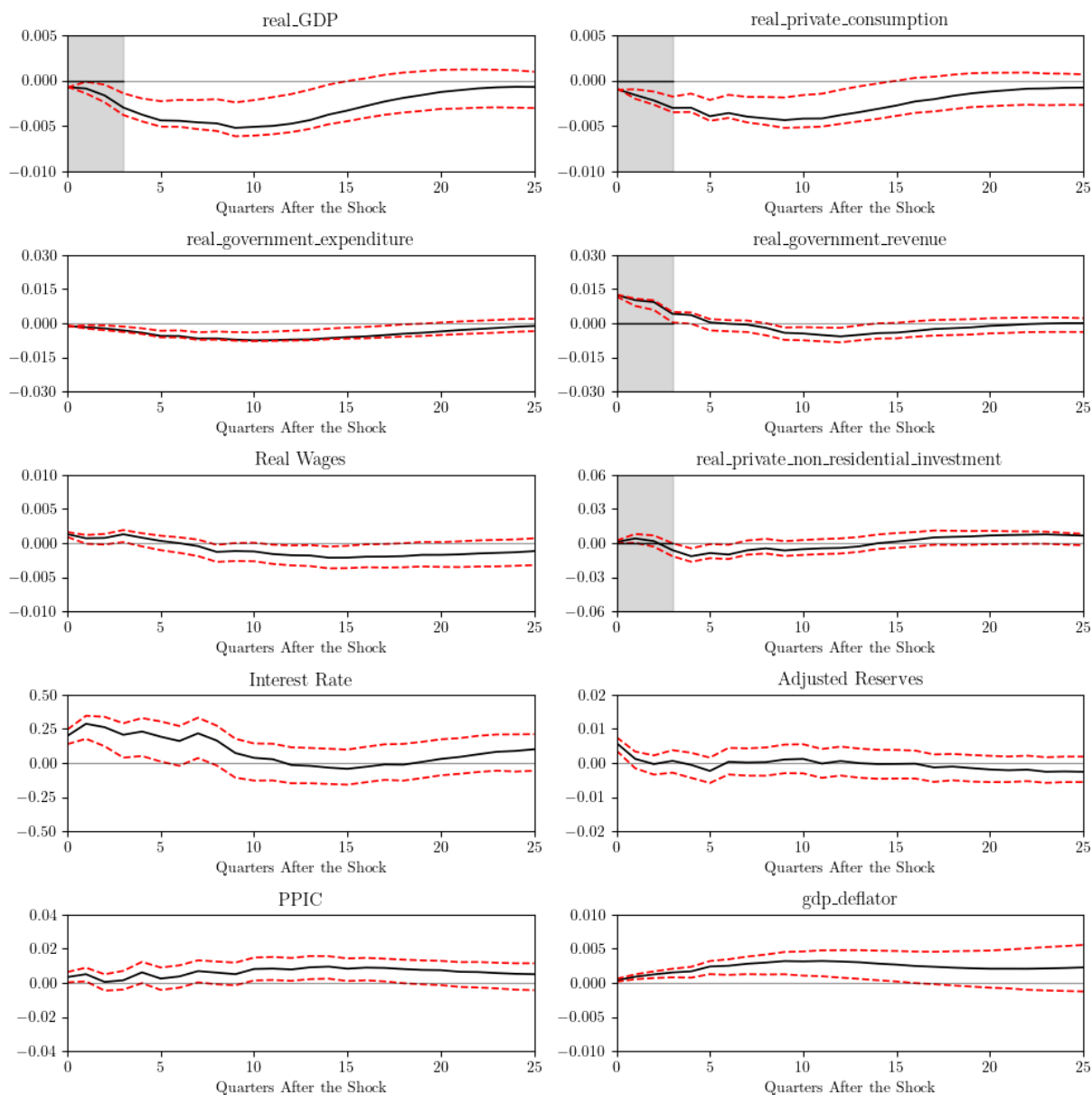
axes[0,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[0,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[2,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.01, 0.005)
axes[0,1].set_ylim(-0.01, 0.005)
axes[1,0].set_ylim(-0.03, 0.02)
axes[1,1].set_ylim(-0.03, 0.02)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 0.5)
axes[3,1].set_ylim(-0.02, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("revenue_mc.pdf", format="PDF")
plt.show()

```



Spending Shocks

```
In [49]: irf_lower = np.zeros((25 + 1, 10))
irf_upper = np.zeros((25 + 1, 10))

for k in range(25 + 1):
    for i in range(10):
        cholesky_irf_lower = lower[k, i, :]
        cholesky_irf_upper = upper[k, i, :]

        irf_lower[k, i] = np.dot(cholesky_irf_lower, q_spending_shocks)
        irf_upper[k, i] = np.dot(cholesky_irf_upper, q_spending_shocks)
print(irf_lower.shape, irf_upper.shape)
```

(26, 10) (26, 10)

```
In [50]: fig, axes = plt.subplots(5, 2, figsize=(10, 10))
for i in range(5):
```

```

for j in range(2):
    axes[i,j].plot(r_spending[:, i*2+j], color = "black", lw=1.25)
    axes[i,j].plot(irf_lower[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].plot(irf_upper[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
    axes[i,j].set_xlim(0,25)
    axes[i,j].set_title(names[i*2+j], fontsize=12)
    axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[0,0].axvspan(0, 3, color='gray', alpha=0.3)
axes[0,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)
axes[2,1].axvspan(0, 3, color='gray', alpha=0.3)

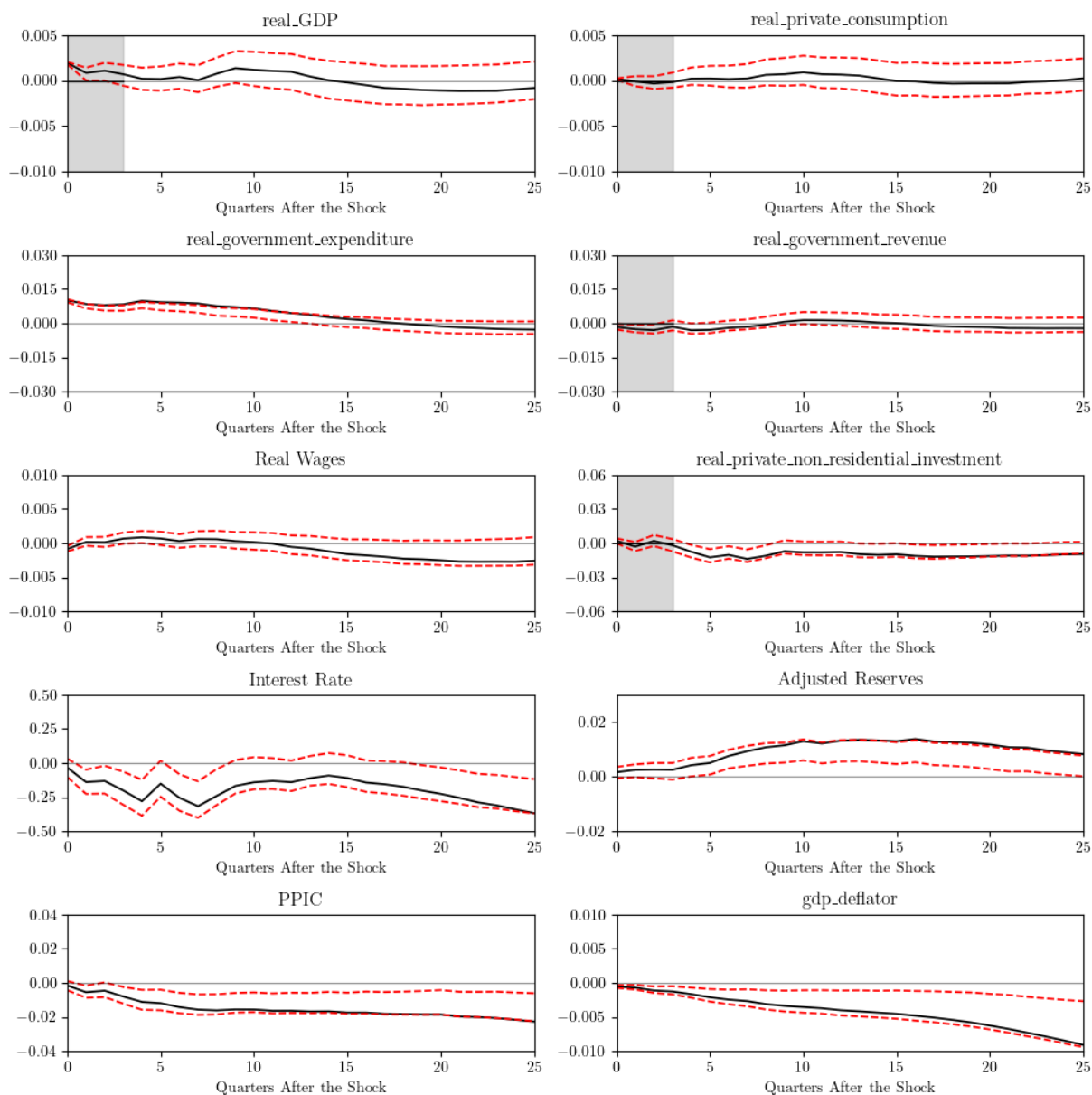
axes[0,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[0,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[2,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.01, 0.005)
axes[0,1].set_ylim(-0.01, 0.005)
axes[1,0].set_ylim(-0.03, 0.02)
axes[1,1].set_ylim(-0.03, 0.02)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 0.5)
axes[3,1].set_ylim(-0.02, 0.03)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("spending_mc.pdf", format="PDF")
plt.show()

```



Zero-Restriction Revenue Shocks

```
In [51]: irf_lower = np.zeros((25 + 1, 10))
irf_upper = np.zeros((25 + 1, 10))

for k in range(25 + 1):
    for i in range(10):
        cholesky_irf_lower = lower[k, i, :]
        cholesky_irf_upper = upper[k, i, :]

        irf_lower[k, i] = np.dot(cholesky_irf_lower, q_anticipated_revenue_shocks)
        irf_upper[k, i] = np.dot(cholesky_irf_upper, q_anticipated_revenue_shocks)
print(irf_lower.shape, irf_upper.shape)
```

(26, 10) (26, 10)

```
In [52]: fig, axes = plt.subplots(5, 2, figsize=(10, 10))
for i in range(5):
```

```

for j in range(2):
    axes[i,j].plot(r_anticipated_revenue[:, i*2+j], color = "black", lw=1.25)
    axes[i,j].plot(irf_lower[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].plot(irf_upper[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
    axes[i,j].set_xlim(0,25)
    axes[i,j].set_title(names[i*2+j], fontsize=12)
    axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)

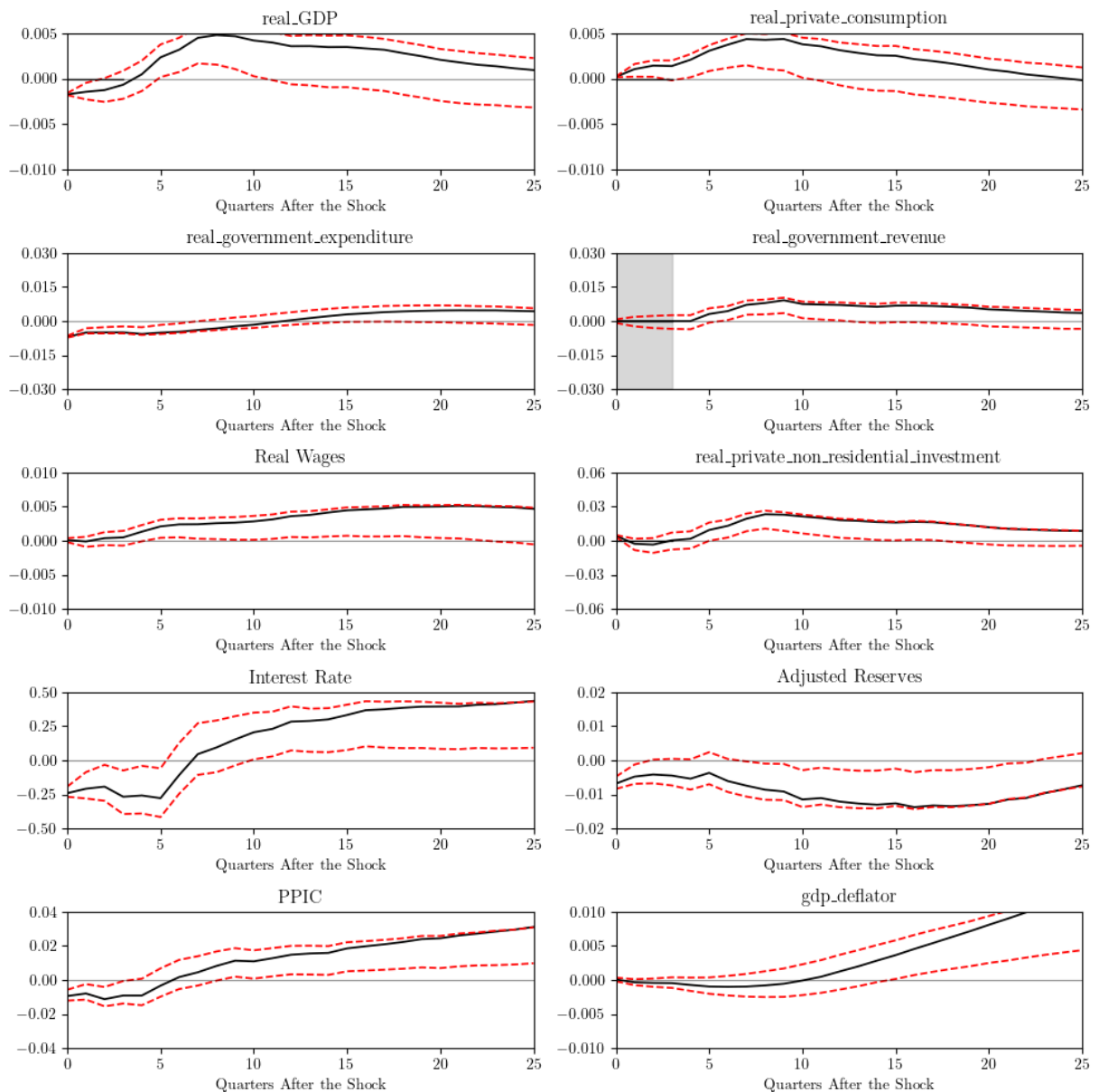
axes[0,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[0,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[2,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.01, 0.005)
axes[0,1].set_ylim(-0.01, 0.005)
axes[1,0].set_ylim(-0.03, 0.02)
axes[1,1].set_ylim(-0.03, 0.02)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 0.5)
axes[3,1].set_ylim(-0.02, 0.02)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("anticipated_spending_mc.pdf", format="PDF")
plt.show()

```



Zero-Restriction Spending Shocks

```
In [53]: irf_lower = np.zeros((25 + 1, 10))
         irf_upper = np.zeros((25 + 1, 10))

         for k in range(25 + 1):
             for i in range(10):
                 cholesky_irf_lower = lower[k, i, :]
                 cholesky_irf_upper = upper[k, i, :]

                 irf_lower[k, i] = np.dot(cholesky_irf_lower, q_anticipated_spending_shocks)
                 irf_upper[k, i] = np.dot(cholesky_irf_upper, q_anticipated_spending_shocks)
         print(irf_lower.shape, irf_upper.shape)

(26, 10) (26, 10)
```

```
In [54]: fig, axes = plt.subplots(5, 2, figsize=(10, 10))
         for i in range(5):
```



```

for j in range(2):
    axes[i,j].plot(r_anticipated_spending[:, i*2+j], color = "black", lw=1.25)
    axes[i,j].plot(irf_lower[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].plot(irf_upper[:, i*2+j], color = "red", ls="dashed", lw=1.25)
    axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
    axes[i,j].set_xlim(0,25)
    axes[i,j].set_title(names[i*2+j], fontsize=12)
    axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[1,0].axvspan(0, 3, color='gray', alpha=0.3)

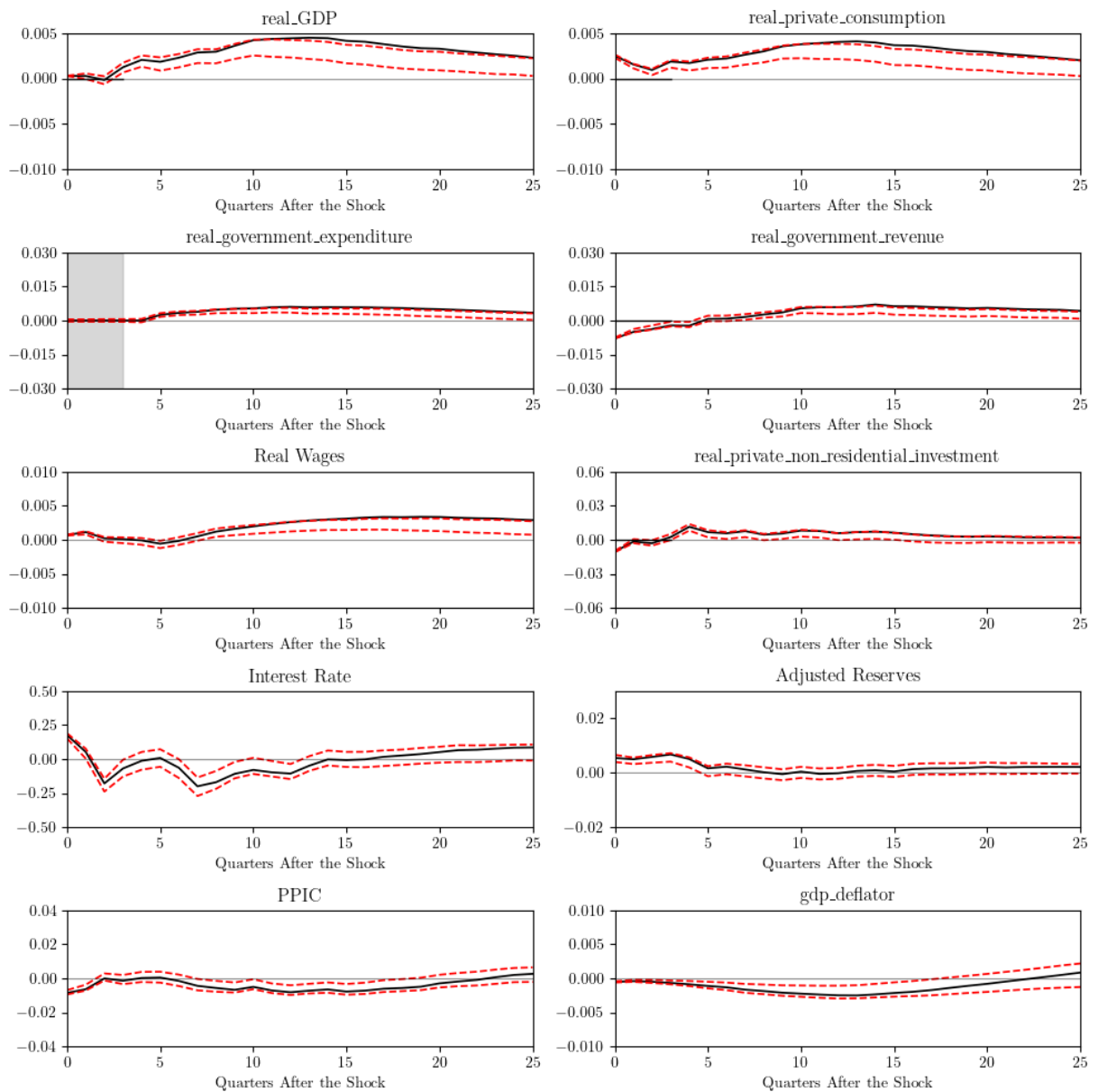
axes[0,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[0,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[2,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.01, 0.005)
axes[0,1].set_ylim(-0.01, 0.005)
axes[1,0].set_ylim(-0.03, 0.02)
axes[1,1].set_ylim(-0.03, 0.02)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 0.5)
axes[3,1].set_ylim(-0.02, 0.03)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("anticipated_spending_mc.pdf", format="PDF")
plt.show()

```



Extention 4 - Policy Analysis

```
In [55]: cashe = get_r_ja(var_res,q_revenue_shocks,4)
gov_spending_by_revenue = cashe[:4, 2]
gov_revenue_by_revenue = cashe[:4, 3]
print(gov_spending_by_revenue)
```

```
[-0.00121186 -0.00178498 -0.00228191 -0.00314356]
```

```
In [56]: cashe = get_r_ja(var_res,q_spending_shocks,4)
gov_spending_by_spending = cashe[:4, 2]
gov_revenue_by_spending = cashe[:4, 3]
print(gov_spending_by_spending)
```

```
[0.01004194 0.00843614 0.0079416 0.00828326]
```

```
In [57]: A_spending_gs = np.array([
    [gov_spending_by_spending[0], 0, 0, 0],
    [gov_spending_by_spending[1], gov_spending_by_spending[0], 0, 0],
```

```

    [gov_spending_by_spending[2], gov_spending_by_spending[1], gov_spending_by_spen
    [gov_spending_by_spending[3], gov_spending_by_spending[2], gov_spending_by_spen
])

A_spending_gr = np.array([
    [gov_spending_by_revenue[0], 0, 0, 0],
    [gov_spending_by_revenue[1], gov_spending_by_revenue[0], 0, 0],
    [gov_spending_by_revenue[2], gov_spending_by_revenue[1], gov_spending_by_revenu
    [gov_spending_by_revenue[3], gov_spending_by_revenue[2], gov_spending_by_revenu
])

A_revenue_gs = np.array([
    [gov_revenue_by_spending[0], 0, 0, 0],
    [gov_revenue_by_spending[1], gov_revenue_by_spending[0], 0, 0],
    [gov_revenue_by_spending[2], gov_revenue_by_spending[1], gov_revenue_by_spendin
    [gov_revenue_by_spending[3], gov_revenue_by_spending[2], gov_revenue_by_spendin
])

A_revenue_gr = np.array([
    [gov_revenue_by_revenue[0], 0, 0, 0],
    [gov_revenue_by_revenue[1], gov_revenue_by_revenue[0], 0, 0],
    [gov_revenue_by_revenue[2], gov_revenue_by_revenue[1], gov_revenue_by_revenue[0
    [gov_revenue_by_revenue[3], gov_revenue_by_revenue[2], gov_revenue_by_revenue[1
])

A_full_8d = np.block([
    [A_spending_gs, A_spending_gr],
    [A_revenue_gs, A_revenue_gr]
])

b_8d = np.concatenate([
    np.full(4, 0.01),
    np.zeros(4)
])

solution_8d = np.linalg.solve(A_full_8d, b_8d)

shock_labels_8d = [f"BGS_{i}" for i in range(4)] + [f"BGR_{i}" for i in range(4)]
solution_8d_series = pd.Series(solution_8d, index=shock_labels_8d)
solution_8d_series

```

```

Out[57]: BGS_0    1.011525
        BGS_1    0.184709
        BGS_2    0.104465
        BGS_3    0.011562
        BGR_0    0.130111
        BGR_1    0.128688
        BGR_2    0.093888
        BGR_3   -0.022494
        dtype: float64

```

```

In [58]: irfs_revenue = get_r_ja(var_res, q_revenue_shocks, 30)
        irfs_spending = get_r_ja(var_res, q_spending_shocks, 30)

        BGS_j = solution_8d_series.iloc[:4].values

```

```

BGR_j = solution_8d_series.iloc[4:].values

BGS_extended = np.zeros(30)
BGS_extended[:4] = BGS_j

BGR_extended = np.zeros(30)
BGR_extended[:4] = BGR_j

response_spending = np.zeros((30, 10))
response_revenue = np.zeros((30, 10))

for k in range(30):
    for j in range(k + 1):
        response_spending[k] += BGS_extended[j] * irfs_spending[k - j]
        response_revenue[k] += BGR_extended[j] * irfs_revenue[k - j]

response_total = response_spending + response_revenue

```

```

In [59]: irf_revenue_lower = np.zeros((30, 10))
irf_revenue_upper = np.zeros((30, 10))
irf_spending_lower = np.zeros((30, 10))
irf_spending_upper = np.zeros((30, 10))

for k in range(30):
    for i in range(10):
        cholesky_irf_lower = lower[k, i, :]
        cholesky_irf_upper = upper[k, i, :]

        irf_revenue_lower[k, i] = np.dot(cholesky_irf_lower, q_revenue_shocks)
        irf_revenue_upper[k, i] = np.dot(cholesky_irf_upper, q_revenue_shocks)

        irf_spending_lower[k, i] = np.dot(cholesky_irf_lower, q_spending_shocks)
        irf_spending_upper[k, i] = np.dot(cholesky_irf_upper, q_spending_shocks)
    print(irf_revenue_lower.shape)

(30, 10)

```

```

In [60]: response_lower = np.zeros((30, 10))
response_upper = np.zeros((30, 10))

for k in range(30):
    for j in range(k + 1):
        response_lower[k] += (
            BGS_extended[j] * irf_spending_lower[k - j]
            + BGR_extended[j] * irf_revenue_lower[k - j]
        )
        response_upper[k] += (
            BGS_extended[j] * irf_spending_upper[k - j]
            + BGR_extended[j] * irf_revenue_upper[k - j]
        )

```

```

In [61]: fig, axes = plt.subplots(5, 2, figsize=(10, 10))
for i in range(5):
    for j in range(2):

```

```

axes[i,j].plot(response_total[:, i*2+j], color = "black", lw=1.25)
axes[i,j].plot(response_lower[:, i*2+j], color = "red", ls="dashed", lw=1.2)
axes[i,j].plot(response_upper[:, i*2+j], color = "red", ls="dashed", lw=1.2)
axes[i,j].axhline(y=0, color = "black", alpha=0.4, lw=0.8)
axes[i,j].set_xlim(0,25)
axes[i,j].set_title(names[i*2+j], fontsize=12)
axes[i,j].set_xlabel("Quarters After the Shock", fontsize=10)

axes[1,0].axvspan(0, 3, color='gray', alpha=0.3)
axes[1,1].axvspan(0, 3, color='gray', alpha=0.3)

axes[1,1].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)
axes[1,0].axhline(y=0, xmin=0, xmax=0.12, color = "black", lw=1)

axes[0,0].set_ylim(-0.01, 0.005)
axes[0,1].set_ylim(-0.01, 0.005)
axes[1,0].set_ylim(-0.03, 0.02)
axes[1,1].set_ylim(-0.03, 0.02)
axes[2,0].set_ylim(-0.01, 0.01)
axes[2,1].set_ylim(-0.04, 0.04)
axes[3,0].set_ylim(-0.5, 0.5)
axes[3,1].set_ylim(-0.02, 0.03)
axes[4,0].set_ylim(-0.04, 0.04)
axes[4,1].set_ylim(-0.01, 0.01)

axes[1,0].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[1,1].set_yticks([-0.03,-0.015,0,0.015,0.03])
axes[2,1].set_yticks([-0.06,-0.03,0,0.03,0.06])

plt.tight_layout()
plt.savefig("deficit_spending_mc.pdf", format="PDF")
plt.show()

```

