

Effective Coverage vs. Lives Saved: A Comparison

In this document, we will find the optimal set of nutritional interventions over space and time, first using each micronutrient's effect on effective coverage separately, and then using all interventions together to see if they yield different optimal results, using a measure of the number of lives saved from each intervention.

Summary of Results

```
1 import import_ipynb
2 import os
3 import geopandas as gpd
4 os.chdir('/home/lordflaron/Documents/minimod')
5 import optimization_work.demewoz_lives_saved.lives_saved_analysis as lives_saved
6 import optimization_work.demewoz_lives_saved.folate_effective_coverage_analysis as folate
7 import optimization_work.demewoz_lives_saved.zinc_effective_coverage_analysis as zinc
8 import optimization_work.demewoz_lives_saved.vas_effective_coverage_analysis as vas
9 import pandas as pd
10
11 # First only get the optimal values
12 lives_saved_opt = lives_saved.models['lives_saved'][None].opt_df.loc[lambdas df: df['opt_vals']>0].groupby(['region', 'time']).sum()
13 lives_saved_high_opt = lives_saved.models['lives_saved_high'][None].opt_df.loc[lambdas df: df['opt_vals']>0].groupby(['region', 'time']).sum()
14 zinc_opt = zinc.models[None].opt_df.loc[lambdas df: df['opt_vals']>0].groupby(['region', 'time']).sum()
15 vas_opt = vas.models[None].opt_df.loc[lambdas df: df['opt_vals']>0].groupby(['region', 'time']).sum()
16 folate_opt = folate.models[None].opt_df.loc[lambdas df: df['opt_vals']>0].groupby(['region', 'time']).sum()
17
18 # Get names of optimal interventions
19 lives_saved_name = lives_saved.models['lives_saved'][None].optimal_interventions
20 lives_saved_high_name = lives_saved.models['lives_saved_high'][None].optimal_interventions
21 zinc_name = zinc.models[None].optimal_interventions
22 vas_name = vas.models[None].optimal_interventions
23 folate_name = folate.models[None].optimal_interventions
24
25 # Get names of bau
26 lives_saved_bau = lives_saved.models['lives_saved'][None].bau_df.index.get_level_values(level='intervention').unique().tolist()
27 lives_saved_high_bau = lives_saved.models['lives_saved_high'][None].bau_df.index.get_level_values(level='intervention').unique().tolist()
28 zinc_bau = zinc.models[None].bau_df.index.get_level_values(level='intervention').unique().tolist()
29 vas_bau = vas.models[None].bau_df.index.get_level_values(level='intervention').unique().tolist()
30 folate_bau = folate.models[None].bau_df.index.get_level_values(level='intervention').unique().tolist()
31
32 opt_dict = {'Zinc\n(Children Eff. Cov.)' : zinc_opt,
33 'VA\n(Children Eff. Cov.)' : vas_opt,
34 'Folic Acid\n(WRA Eff. Cov.)' : folate_opt,
35 'All (Lives Saved)' : lives_saved_opt,
36 'All (Lives Saved, Alt.)' : lives_saved_high_opt}
37
38 df_all = pd.concat(opt_dict.values(), axis=1)
```

39

40 df_all.columns = pd.MultiIndex.from_product([opt_dict.keys(), ['CB', 'CC', 'BAU* Cost per Benefit']])

```

importing Jupyter notebook from /home/lordflaron/Documents/minimod/optimization_work/demewoz_lives_saved
.pipe(observation_adjustment,
int1 = "cube",
int2 = 0,
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "cubezcube",
int2 = 0,
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcube",
int2 = "maxoil",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcube",
int2 = "oil",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubevas",
int2 = "oilvas",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubevas",
int2 = "maxoilvas",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "cubevas",
int2 = "vas",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "cubeclinic",
int2 = "clinic",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubeclinic",
int2 = "maxoilclinic",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubeclinic",
int2 = "oilclinic",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "cubezflour",
int2 = "zflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubezflour",
int2 = "maxoilzflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubezflour",

```

```

int2 = "oilzflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubevaszflour",
int2 = "oilvaszflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "cubevaszflour",
int2 = "vaszflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubevaszflour",
int2 = "maxoilvaszflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubecliniczflour",
int2 = "oilcliniczflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubecliniczflour",
int2 = "maxoilcliniczflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "cubezflourzcube",
int2 = "zflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubezflourzcube",
int2 = "maxoilzflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubezflourzcube",
int2 = "oilzflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubevaszflourzcube",
int2 = "oilvaszflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilcubecliniczflourzcube",
int2 = "oilcliniczflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubecliniczflourzcube",
int2 = "maxoilcliniczflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "zcube",
int2 = 0,
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "zflourzcube",
int2 = "zflour",
time_to_replace = [1,2,3])

```

```

.pipe(observation_adjustment,
int1 = "oilvaszflourzcube",
int2 = "oilvaszflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoilcubezflourzcube",
int2 = "maxoilzflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilzflourzcube",
int2 = "oilzflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "oilzflourfcubefflour",
int2 = "oilzflourfflour",
time_to_replace = [1,2,3])
.pipe(observation_adjustment,
int1 = "maxoil",
int2 = "oil",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcube",
int2 = "oilcube",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilvas",
int2 = "oilvas",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubevas",
int2 = "oilcubevas",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilclinic",
int2 = "oilclinic",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubecclinic",
int2 = "oilcubecclinic",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilzflour",
int2 = "oilzflour",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubezflour",
int2 = "oilcubezflour",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilvaszflour",
int2 = "oilvaszflour",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcliniczflour",

```

```

int2 = "oilcliniczflour",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubevaszflour",
int2 = "oilcubevaszflour",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubecliniczflour",
int2 = "oilcubecliniczflour",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilzflourzcube",
int2 = "oilzflourzcube",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubezflourzcube",
int2 = "oilcubezflourzcube",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilvaszflourzcube",
int2 = "oilvaszflourzcube",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcliniczflourzcube",
int2 = "oilcliniczflourzcube",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubevaszflourzcube",
int2 = "oilcubevaszflourzcube",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "maxoilcubecliniczflourzcube",
int2 = "oilcubecliniczflourzcube",
time_to_replace = [1,2])
.pipe(observation_adjustment,
int1 = "zflourfflour33",
int2 = 0,
time_to_replace = slice(None))
.pipe(observation_adjustment,
int1 = "fflour33",
int2 = 0,
time_to_replace = slice(None))
Changed cube to 0
Changed cubezcube to 0
Changed maxoilcube to maxoil
Changed oilcube to oil
Changed oilcubevas to oilvas
Changed maxoilcubevas to maxoilvas
Changed cubevas to vas
Changed cubeclinic to clinic
Changed maxoilcubeclinic to maxoilclinic
Changed oilcubeclinic to oilclinic
Changed cubezflour to zflour
Changed maxoilcubezflour to maxoilzflour

```

Changed oilcubezflour to oilzflour
 Changed oilcubevaszflour to oilvaszflour
 Changed cubevaszflour to vaszflour
 Changed maxoilcubevaszflour to maxoilvaszflour
 Changed oilcubecliniczflour to oilcliniczflour
 Changed maxoilcubecliniczflour to maxoilcliniczflour
 Changed cubezflourzcube to zflour
 Changed maxoilcubezflourzcube to maxoilzflour
 Changed oilcubezflourzcube to oilzflour
 Changed oilcubevaszflourzcube to oilvaszflour
 Changed oilcubecliniczflourzcube to oilcliniczflour
 Changed maxoilcubecliniczflourzcube to maxoilcliniczflour
 Changed zcube to 0
 Changed zflourzcube to zflour
 Changed oilvaszflourzcube to oilvaszflour
 Changed maxoilcubezflourzcube to maxoilzflour
 Changed oilzflourzcube to oilzflour
 Changed oilzflourfcubefflour to oilzflourfflour
 Changed maxoil to oil
 Changed maxoilcube to oilcube
 Changed maxoilvas to oilvas
 Changed maxoilcubevas to oilcubevas
 Changed maxoilclinic to oilclinic
 Changed maxoilcubeclinic to oilcubeclinic
 Changed maxoilzflour to oilzflour
 Changed maxoilcubezflour to oilcubezflour
 Changed maxoilvaszflour to oilvaszflour
 Changed maxoilcliniczflour to oilcliniczflour
 Changed maxoilcubevaszflour to oilcubevaszflour
 Changed maxoilcubecliniczflour to oilcubecliniczflour
 Changed maxoilzflourzcube to oilzflourzcube
 Changed maxoilcubezflourzcube to oilcubezflourzcube
 Changed maxoilvaszflourzcube to oilvaszflourzcube
 Changed maxoilcliniczflourzcube to oilcliniczflourzcube
 Changed maxoilcubevaszflourzcube to oilcubevaszflourzcube
 Changed maxoilcubecliniczflourzcube to oilcubecliniczflourzcube
 Changed zflourfflour33 to 0
 Changed fflour33 to 0
 Changed cube to 0
 Changed cubezcube to 0
 Changed maxoilcube to maxoil
 Changed oilcube to oil
 Changed oilcubevas to oilvas
 Changed maxoilcubevas to maxoilvas
 Changed cubevas to vas
 Changed cubeclinic to clinic
 Changed maxoilcubeclinic to maxoilclinic
 Changed oilcubeclinic to oilclinic
 Changed cubezflour to zflour
 Changed maxoilcubezflour to maxoilzflour
 Changed oilcubezflour to oilzflour
 Changed oilcubevaszflour to oilvaszflour
 Changed cubevaszflour to vaszflour
 Changed maxoilcubevaszflour to maxoilvaszflour

Changed oilcubecliniczflour to oilcliniczflour
 Changed maxoilcubecliniczflour to maxoilcliniczflour
 Changed cubezflourzcube to zflour
 Changed maxoilcubezflourzcube to maxoilzflour
 Changed oilcubezflourzcube to oilzflour
 Changed oilcubevaszflourzcube to oilvaszflour
 Changed oilcubecliniczflourzcube to oilcliniczflour
 Changed maxoilcubecliniczflourzcube to maxoilcliniczflour
 Changed zcube to 0
 Changed zflourzcube to zflour
 Changed oilvaszflourzcube to oilvaszflour
 Changed maxoilcubezflourzcube to maxoilzflour
 Changed oilzflourzcube to oilzflour
 Changed oilzflourfcubefflour to oilzflourfflour
 Changed maxoil to oil
 Changed maxoilcube to oilcube
 Changed maxoilvas to oilvas
 Changed maxoilcubevas to oilcubevas
 Changed maxoilclinic to oilclinic
 Changed maxoilcubeclinic to oilcubeclinic
 Changed maxoilzflour to oilzflour
 Changed maxoilcubezflour to oilcubezflour
 Changed maxoilvaszflour to oilvaszflour
 Changed maxoilcliniczflour to oilcliniczflour
 Changed maxoilcubevaszflour to oilcubevaszflour
 Changed maxoilcubecliniczflour to oilcubecliniczflour
 Changed maxoilzflourzcube to oilzflourzcube
 Changed maxoilcubezflourzcube to oilcubezflourzcube
 Changed maxoilvaszflourzcube to oilvaszflourzcube
 Changed maxoilcliniczflourzcube to oilcliniczflourzcube
 Changed maxoilcubevaszflourzcube to oilcubevaszflourzcube
 Changed maxoilcubecliniczflourzcube to oilcubecliniczflourzcube
 Changed zflourfflour33 to 0
 Changed fflour33 to 0
 Changed cubezflourzcubeffcubefflour to zflourfflour
 Changed cubezflourfcubefflour to zflourfflour
 Changed oilzflourzcubeffcubefflour to oilzflourfflour
 Changed fcube to 0
 Changed fcube to 0
 Changed oilcube to oil
 Changed oilcubecliniccube to oilclinic
 Changed maxoilzflourfcubefflour to maxoilzflourfflour
 Changed maxoilcliniczflourfcubefflour to maxoilcliniczflourfflour
 Changed maxoilzflourfcubefflour to oilzflourfcubefflour
 Changed maxoilcliniczflourfcubefflour to oilcliniczflourfcubefflour
 Running lives_saved with None
 [Note]: Processing Data...
 [Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool
 Optimization Method: MIN
 Version: 0.0.6dev
 Solver: CBC,
 Show Output: True

[Note]: Optimizing...
 [Note]: Optimal Solution Found
 Running lives_saved_high with None
 [Note]: Processing Data...
 [Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool
 Optimization Method: MIN
 Version: 0.0.6dev
 Solver: CBC,
 Show Output: True

[Note]: Optimizing...
 [Note]: Optimal Solution Found

MiniMod Solver Results	
Method:	MIN
Solver:	CBC
Optimization Status:	OptimizationStatus.OPTIMAL
Number of Solutions Found:	1

No. of Variables:	6750
No. of Integer Variables:	6750
No. of Constraints	1159
No. of Non-zeros in Constr.	231504

Interventions Chosen:

Minimum Benefit	22019.1
Objective Bounds	1.15113e+07
Total Cost	1.15113e+07
Total Lives Saved	23220.5

Cost per Benefit	495.74
------------------	--------

Total Cost and Benefits over Time	
time	opt_vals

time	opt_vals	opt_benefit	opt_costs
1	3	2290	939944
2	3	2321	796726
3	3	2351	1.17346e+06
4	3	2696	1.13025e+06
5	3	2732	1.23289e+06
6	3	2770	1.22924e+06
7	3	2807	1.18669e+06
8	3	2844	1.20595e+06
9	3	2884	1.28702e+06

| 10 | 3 | 2919 | 1.32918e+06 |

Optimal Interventions

	1	2	3	4	5	6	7	8	9	10
('oilzflourfcubefflour', 'Cities')	1	1	1	1	1	1	1	1	1	1
('oilzflourfcubefflour', 'North')	1	1	1	1	1	1	1	1	1	1
('oilzflourfcubefflour', 'South')	1	1	1	1	1	1	1	1	1	1

importing Jupyter notebook from /home/lordflaron/Documents/minimod/optimization_work/demewoz_lives_saved

Changed fcube to 0

Changed fflour to fflour33

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool

Optimization Method: MIN

Version: 0.0.6dev

Solver: CBC,

Show Output: True

[Note]: Optimizing...

[Note]: Optimal Solution Found

MiniMod Solver Results	
Method:	MIN
Solver:	CBC
Optimization Status:	OptimizationStatus.OPTIMAL
Number of Solutions Found:	1

No. of Variables:	120
No. of Integer Variables:	120
No. of Constraints	313
No. of Non-zeros in Constr.	1623

Interventions Chosen:

Minimum Benefit	1.45004e+07
Objective Bounds	1.55095e+06
Total Cost	1.55095e+06
Total WRA Effectively Covered (Folate)	1.45004e+07

Cost per Benefit	0.106959
------------------	----------

Total Cost and Benefits over Time	
-----------------------------------	--

time	opt_vals	opt_benefit	opt_costs
:	:	:	:

	1		3		0		286412	
	2		3		0		134017	
	3		3		1.92897e+06		135594	
	4		3		1.98616e+06		137190	
	5		3		2.04362e+06		138805	
	6		3		2.10248e+06		140439	
	7		3		2.16447e+06		142093	
	8		3		2.22715e+06		143766	
	9		3		2.29016e+06		145460	
	10		3		2.35346e+06		147172	

Optimal Interventions

	1	2	3	4	5	6	7	8	9	10
('fflour33', 'Cities')	1	1	1	1	1	1	1	1	1	1
('fflour33', 'North')	1	1	1	1	1	1	1	1	1	1
('fflour33', 'South')	1	1	1	1	1	1	1	1	1	1

Running with None

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

```

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

```

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with time

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

```

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

```

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with space

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

```

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,

```

Show Output: True

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with both

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool

Optimization Method: MIN

Version: 0.0.6dev

Solver: CBC,

Show Output: True

[Note]: Optimizing...

[Note]: Optimal Solution Found

+-----+-----+	
MiniMod Solver Results	
Method:	MIN
Solver:	CBC
Optimization Status:	OptimizationStatus.OPTIMAL
Number of Solutions Found:	1
+-----+-----+	

+-----+-----+	
No. of Variables:	120
No. of Integer Variables:	120
No. of Constraints	313
No. of Non-zeros in Constr.	1623
+-----+-----+	

Interventions Chosen:

+-----+-----+	
Minimum Benefit	1.45004e+07
Objective Bounds	1.55095e+06
Total Cost	1.55095e+06
Total WRA Effectively Covered (Folate)	1.45004e+07
+-----+-----+	

+-----+-----+	
Cost per Benefit	0.106959
+-----+-----+	

+-----+-----+	
Total Cost and Benefits over Time	
+-----+-----+	

time	opt_vals	opt_benefit	opt_costs
-----:	-----:	-----:	-----:
1	3	0	286412
2	3	0	134017
3	3	1.92897e+06	135594
4	3	1.98616e+06	137190
5	3	2.04362e+06	138805
6	3	2.10248e+06	140439
7	3	2.16447e+06	142093
8	3	2.22715e+06	143766

time	opt_vals	opt_benefit	opt_costs
1	3	833571	494593
2	3	846033	343359
3	3	858734	641157
4	3	872058	588019
5	3	885833	680607
6	3	899011	666795
7	3	912323	613962
8	3	925599	622817
9	3	938577	693359
10	3	951450	724881

Optimal Interventions

	1	2	3	4	5	6	7	8	9	10
('zflour', 'Cities')	1	1	1	1	1	1	1	1	1	1
('zflour', 'North')	1	1	1	1	1	1	1	1	1	1
('zflour', 'South')	1	1	1	1	1	1	1	1	1	1

Running with None

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

```

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

```

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with time

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

```

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

```

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with space

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

```

MiniMod Nutrition Intervention Tool

```

Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

[Note]: Optimizing...
[Note]: Optimal Solution Found
Running with both
[Note]: Processing Data...
[Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

[Note]: Optimizing...
[Note]: Optimal Solution Found

MiniMod Solver Results	
Method:	MIN
Solver:	CBC
Optimization Status:	OptimizationStatus.OPTIMAL
Number of Solutions Found:	1

No. of Variables:	90
No. of Integer Variables:	90
No. of Constraints	313
No. of Non-zeros in Constr.	1278

Interventions Chosen:

Minimum Benefit	7.81179e+06
Objective Bounds	6.06955e+06
Total Cost	6.06955e+06
Total Children Effectively Covered (Zinc)	7.81179e+06

Cost per Benefit	0.776973
------------------	----------

Total Cost and Benefits over Time	
-----------------------------------	--

time	opt_vals	opt_benefit	opt_costs
1	3	833571	494593
2	3	846033	343359
3	3	858734	641157
4	3	872058	588019
5	3	885833	680607


```

| No. of Non-zeros in Constr. | 6933 |
+-----+-----+
Interventions Chosen:
+-----+-----+
| Minimum Benefit | 1.24652e+07 |
| Objective Bounds | 2.83901e+07 |
| Total Cost | 2.83901e+07 |
| Total Children Effectively Covered (VA) | 1.2496e+07 |
+-----+-----+
+-----+-----+
| Cost per Benefit | 2.27193 |
+-----+-----+
+-----+-----+
| Total Cost and Benefits over Time | |
+-----+-----+
| time | opt_vals | opt_benefit | opt_costs |
|-----:|-----:|-----:|-----:|
| 1 | 3 | 1.345e+06 | 3.56153e+06 |
| 2 | 3 | 1.35927e+06 | 3.52776e+06 |
| 3 | 3 | 1.05806e+06 | 1.72146e+06 |
| 4 | 3 | 1.4436e+06 | 2.66859e+06 |
| 5 | 3 | 1.46313e+06 | 2.70103e+06 |
| 6 | 3 | 1.48317e+06 | 2.79537e+06 |
| 7 | 3 | 1.50492e+06 | 2.76685e+06 |
| 8 | 3 | 1.52759e+06 | 2.88426e+06 |
| 9 | 3 | 1.55068e+06 | 2.89545e+06 |
| 10 | 3 | 1.57489e+06 | 2.8678e+06 |

```

Optimal Interventions

```

+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+
| ('oilcube', 'Cities') | 0 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| ('oilcube', 'South') | 0 | 0 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| ('oilcubevas', 'Cities') | 1 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| ('oilcubevas', 'North') | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| ('oilcubevas', 'South') | 1 | 1 | 0 | 0 | 0 | 0 | 0 | 0 | 0 | 0 |
+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+-----+

```

Running with None

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool

Optimization Method: MIN

Version: 0.0.6dev

Solver: CBC,

Show Output: True

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with time

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with space

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

[Note]: Optimizing...

[Note]: Optimal Solution Found

Running with both

[Note]: Processing Data...

[Note]: Creating Base Model with constraints

MiniMod Nutrition Intervention Tool
Optimization Method: MIN
Version: 0.0.6dev
Solver: CBC,
Show Output: True

[Note]: Optimizing...

[Note]: Optimal Solution Found

+-----+-----+	
MiniMod Solver Results	
Method:	MIN
Solver:	CBC
Optimization Status:	OptimizationStatus.OPTIMAL
Number of Solutions Found:	1
+-----+-----+	

+-----+-----+	
No. of Variables:	510
No. of Integer Variables:	510
No. of Constraints	454
No. of Non-zeros in Constr.	6933
+-----+-----+	

Interventions Chosen:

+-----+-----+	
Minimum Benefit	1.24652e+07
Objective Bounds	2.83901e+07

Total Cost	2.83901e+07
Total Children Effectively Covered (VA)	1.2496e+07
+-----+	
Cost per Benefit	2.27193
+-----+	
+-----+	
Total Cost and Benefits over Time	
+-----+	
time	opt_vals opt_benefit opt_costs
-----:	-----: -----: -----:
1	3 1.345e+06 3.56153e+06
2	3 1.35927e+06 3.52776e+06
3	3 1.05806e+06 1.72146e+06
4	3 1.4436e+06 2.66859e+06
5	3 1.46313e+06 2.70103e+06
6	3 1.48317e+06 2.79537e+06
7	3 1.50492e+06 2.76685e+06
8	3 1.52759e+06 2.88426e+06
9	3 1.55068e+06 2.89545e+06
10	3 1.57489e+06 2.8678e+06

Optimal Interventions

	1	2	3	4	5	6	7	8	9	10
-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+
('oilcube', 'Cities')	0	0	1	1	1	1	1	1	1	1
('oilcube', 'South')	0	0	1	1	1	1	1	1	1	1
('oilcubevas', 'Cities')	1	1	0	0	0	0	0	0	0	0
('oilcubevas', 'North')	1	1	1	1	1	1	1	1	1	1
('oilcubevas', 'South')	1	1	0	0	0	0	0	0	0	0
-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+	-----+

```

/usr/lib/python3.8/site-packages/geopandas/_compat.py:84: UserWarning: The Shapely GEOS version (3.8.0-
  warnings.warn(
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
  df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
  df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/usr/lib/python3.8/site-packages/IPython/core/interactiveshell.py:3417: PerformanceWarning: indexing pa
  exec(code_obj, self.user_global_ns, self.user_ns)
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
  df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
  df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
  df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
  df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/usr/lib/python3.8/site-packages/IPython/core/interactiveshell.py:3417: PerformanceWarning: indexing pa
  exec(code_obj, self.user_global_ns, self.user_ns)
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
  df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,

```

```

/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/usr/lib/python3.8/site-packages/IPython/core/interactiveshell.py:3417: PerformanceWarning: indexing pa
exec(code_obj, self.user_global_ns, self.user_ns)
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,

```

```

-----
TypeError                                Traceback (most recent call last)
/usr/lib/python3.8/site-packages/ipywidgets/widgets/interaction.py in update(self, *args)
    254         value = widget.get_interact_value()
    255         self.kwargs[widget._kwarg] = value
--> 256         self.result = self.f(**self.kwargs)
    257         show_inline_matplotlib_plots()
    258         if self.auto_display and self.result is not None:

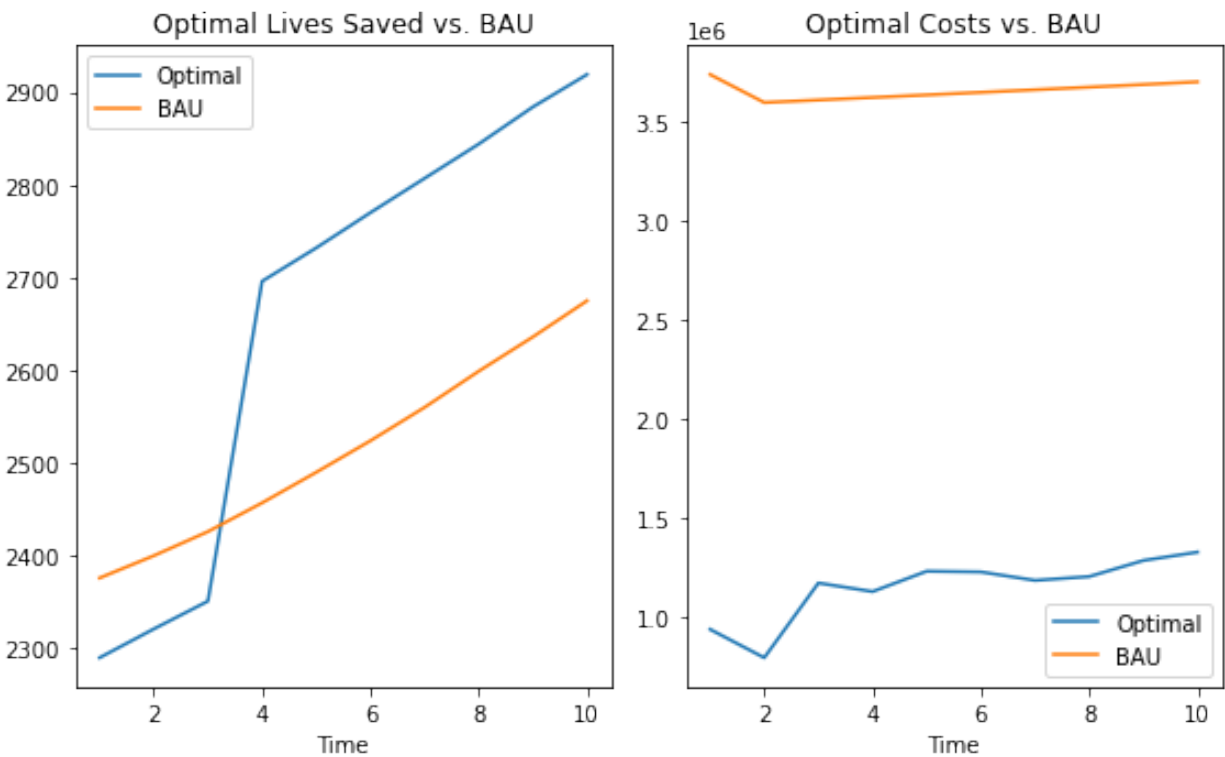
```

~/Documents/minimod/optimization_work/demewoz_lives_saved/vas_effective_coverage_analysis.ipynb in plot.

TypeError: plot_map_benchmark() got an unexpected keyword argument 'bau_intervention_bubbles'

interactive(children=(Dropdown(description='col', options=('lives_saved', 'lives_saved_high'), value='lives_saved'),

interactive(children=(Dropdown(description='col', options=('lives_saved', 'lives_saved_high'), value='lives_saved'),



interactive(children=(Dropdown(description='col', options=('lives_saved', 'lives_saved_high'), value='lives_saved'),

Lives Saved
Optimal Interventions:
oilzflourfcubefflour

Optimal Scenario

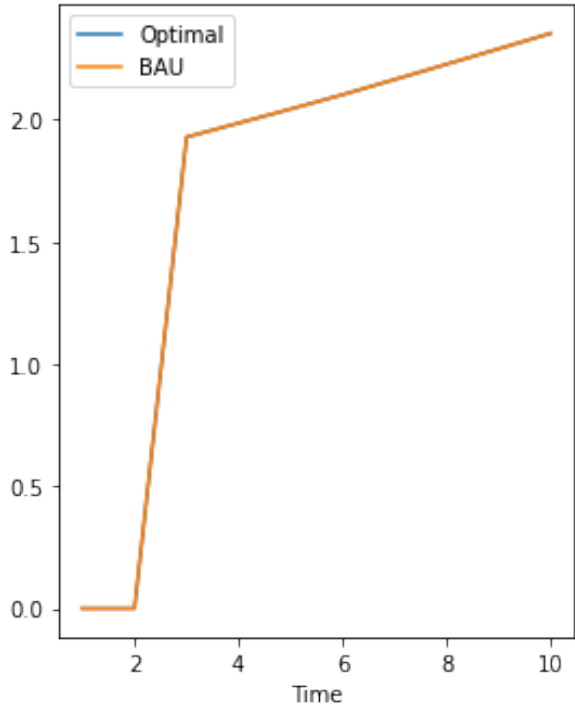


BAU* Scenario

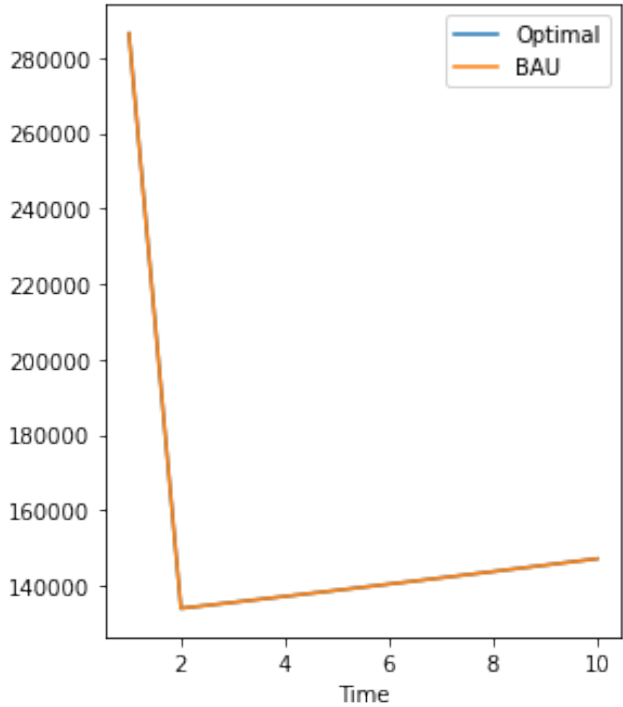


Note: Colors describe Lives Saved
Optimal Interventions:
oilzflourfcubefflour (in millions)

Optimal WRA Effectively Covered (Folate) vs. BAU

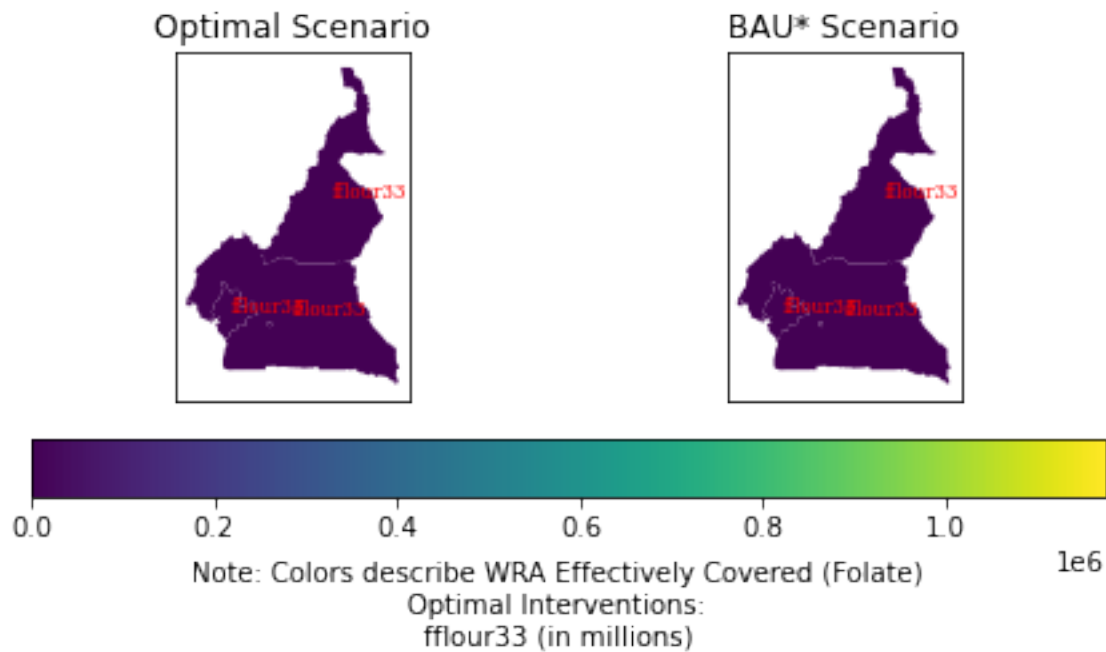


Optimal Costs vs. BAU



interactive(children=(IntSlider(value=1, description='time', max=10, min=1), Dropdown(description='opti

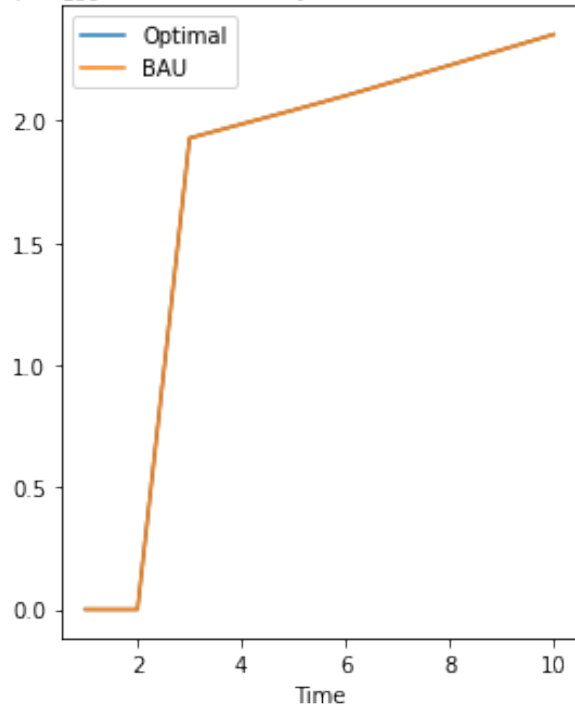
WRA Effectively Covered (Folate)
Optimal Interventions:
fflour33



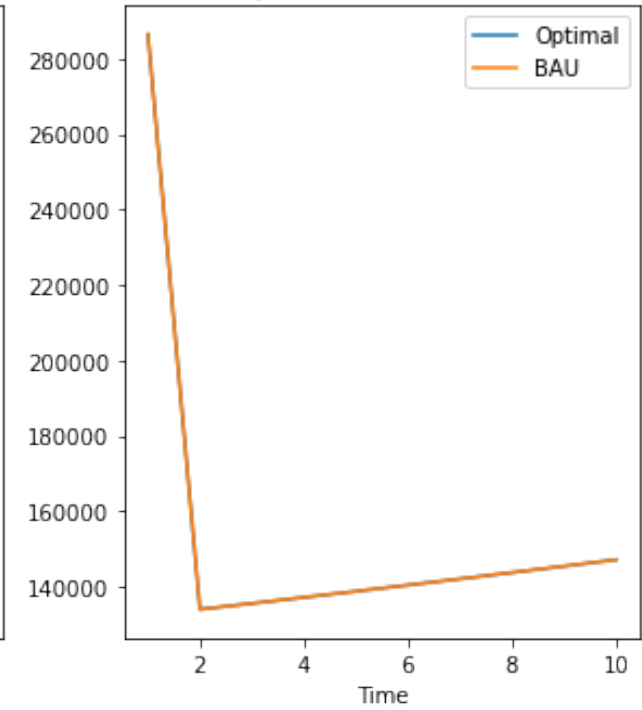
interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver of

interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver of

Optimal WRA Effectively Covered (Folate) vs. BAU



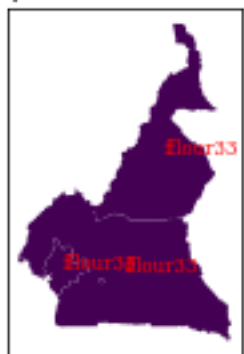
Optimal Costs vs. BAU



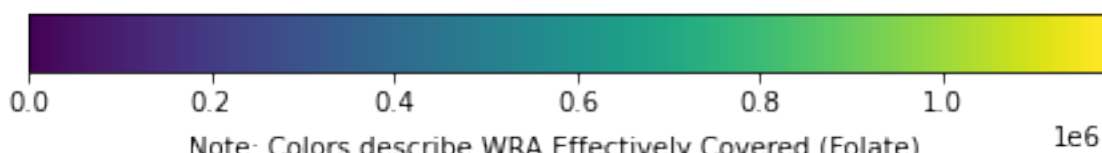
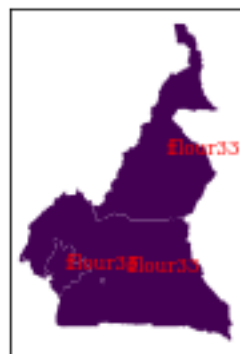
interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver of

WRA Effectively Covered (Folate)
Optimal Interventions:
fflour33

Optimal Scenario



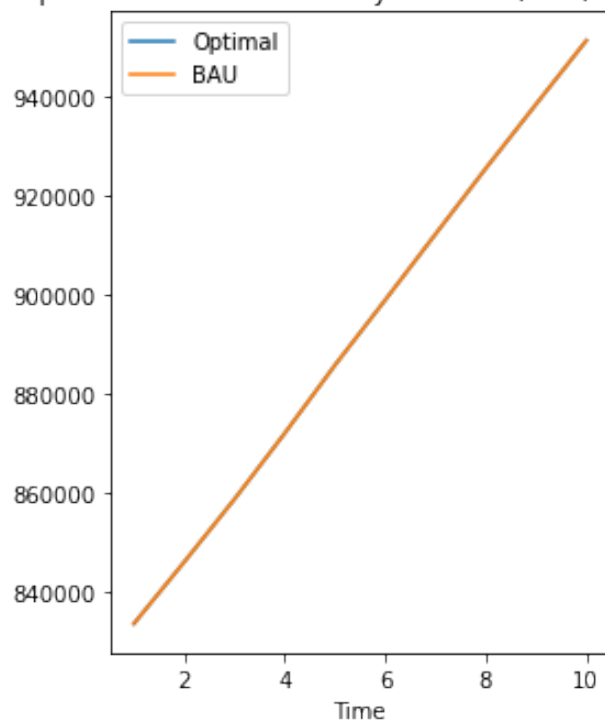
BAU* Scenario



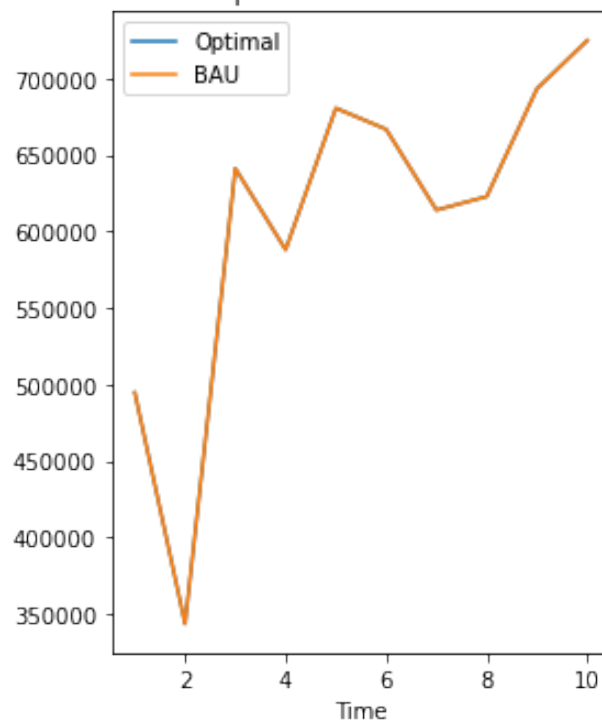
Note: Colors describe WRA Effectively Covered (Folate)
Optimal Interventions:
fflour33 (in millions)

1e6

Optimal Children Effectively Covered (Zinc) vs. BAU

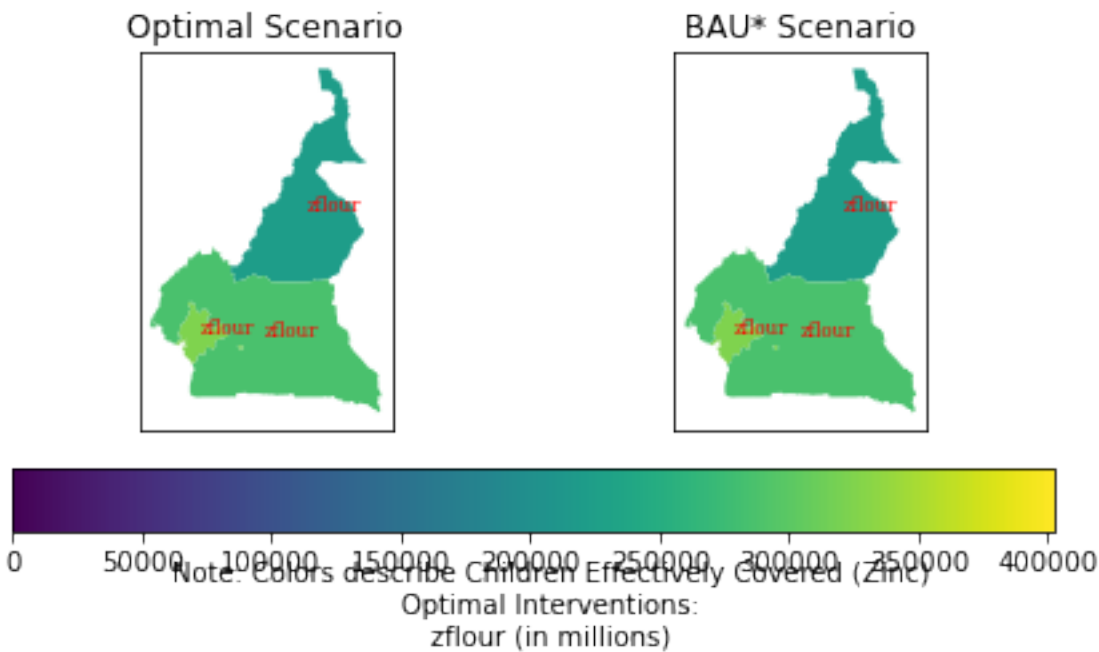


Optimal Costs vs. BAU



```
interactive(children=(IntSlider(value=1, description='time', max=10, min=1), Dropdown(description='opti
```

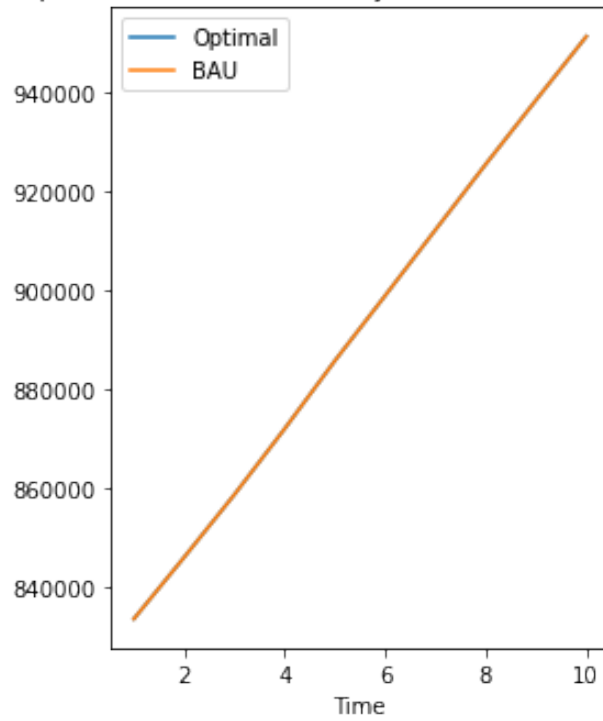
Children Effectively Covered (Zinc)
Optimal Interventions:
zflour



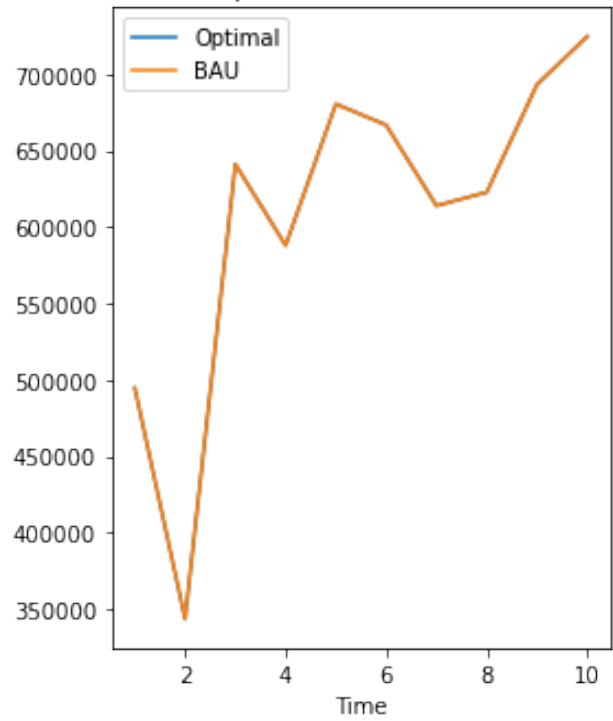
```
interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver of
```

```
interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver of
```


Optimal Children Effectively Covered (Zinc) vs. BAU



Optimal Costs vs. BAU



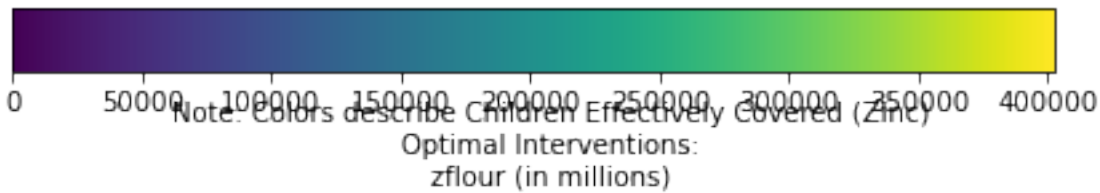
interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver object at 0x104a00000>}, value=None),

Children Effectively Covered (Zinc) Optimal Interventions: zflour

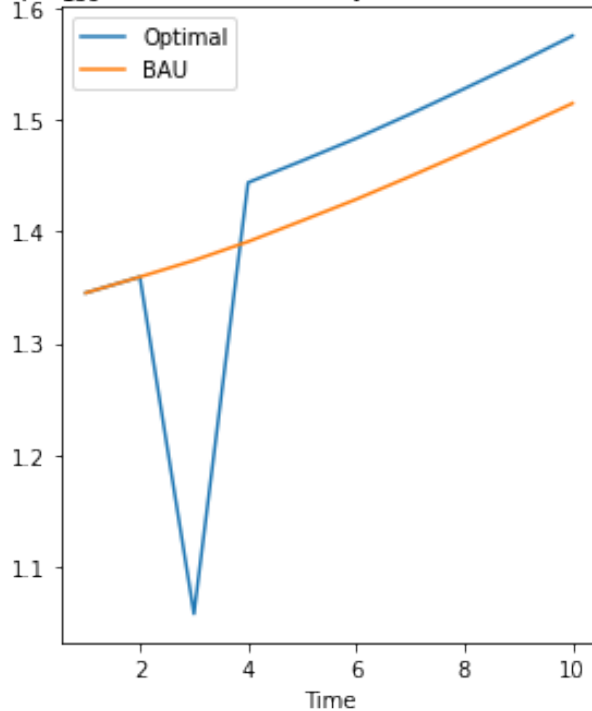
Optimal Scenario



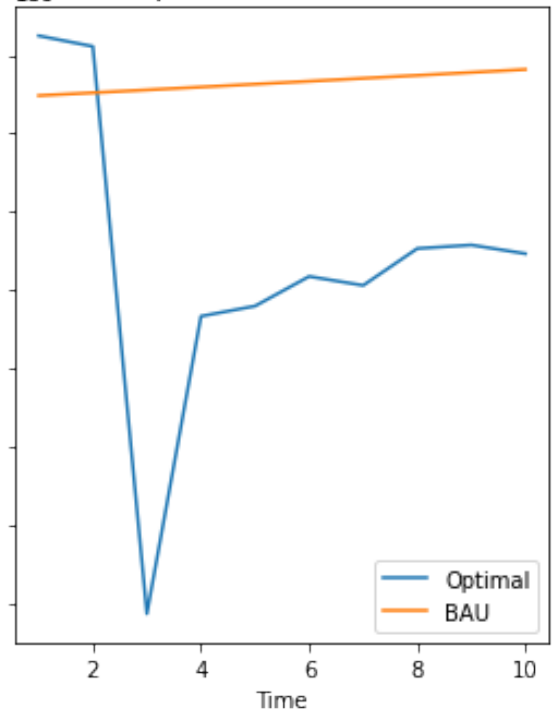
BAU* Scenario



Optimal Children Effectively Covered (VA) vs. BAU



Optimal Costs vs. BAU



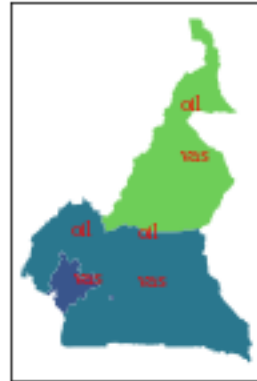
```
interactive(children=(IntSlider(value=1, description='time', max=10, min=1), Dropdown(description='opti
```

Children Effectively Covered (VA)
Optimal Interventions:
oilcube, oilcubevas

Optimal Scenario



BAU* Scenario

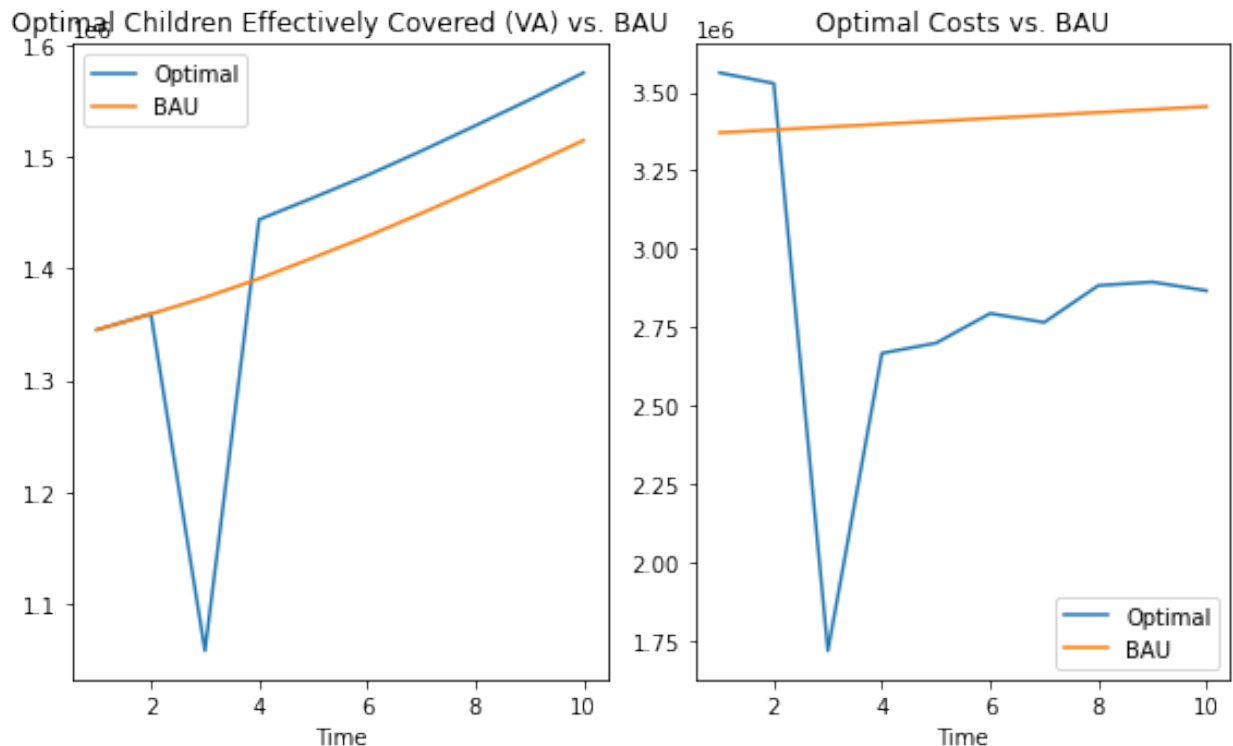


Note: Colors describe Children Effectively Covered (VA)

Optimal Interventions:
oilcube, oilcubevas (in millions)

```
interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver of
```

```
interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver of
```



interactive(children=(Dropdown(description='m', options={None: <minimod.solvers.costsolver.CostSolver object at 0x7f8b1c1c1c1c>}, value=None),

Table 1: Optimally Chosen Interventions vs. BAU*. {#tbl:df_int_opt}

	BAU*	Optimal Intervention Chosen	Interventions Not Chosen
Zinc (Children Eff. Cov.)	Zinc Flour (95 mg/kg)	Zinc Flour (95 mg/kg)	Zinc Cube 600 mg/kg
VA (Children Eff. Cov.)	VA Oil 9 mg/kg + VAS-CHD	VA Oil 12 mg/kg + VAS-CHD (partial) + VA Cube (80 mg/kg)	VA Oil 12 mg/kg, VAS Routine
Folic Acid (WRA Eff. Cov.)	Folic Acid Flour 1.65 mg/kg	Folic Acid Flour 1.65 mg/kg	Folic Acid Cube 100 mg/kg
All (Lives Saved)	VA Oil 9 mg/kg + VAS-CHD + Zinc Flour 95 mg/kg + Folic Acid Flour 1.65 mg/kg	VA Oil 9 mg/kg + Zinc Flour 95 mg/kg + Folic Acid Flour 5 mg/kg + Folic Acid Cube 100 mg/kg	Zinc Cube 600 mg/kg, VA Oil 12 mg/kg, VAS Routine, Folic Acid Cube 100 mg/kg
All (Lives Saved, Alt.)	VA Oil 9 mg/kg + VAS-CHD + Zinc Flour 95 mg/kg + Folic Acid Flour 1.65 mg/kg	VA Oil 9 mg/kg + VAS-CHD (partial) + Zinc Flour (95 mg/kg) + Folic Acid Cube (100 mg/kg) + Folic Acid Flour (5 mg/kg)	Zinc Cube 600 mg/kg, VA Oil 12 mg/kg, VAS Routine, Folic Acid Cube 100 mg/kg

@tbl:df_int_opt shows the set of optimal interventions chosen for each micronutrient simulation as well as their Business as Usual Scenario (BAU).

The BAU* was chosen for each simulation so that it would be consistent across effective coverage and lives

Table 2: Total Benefits and Costs Across Space

		% North	% South	% Cities	National
Zinc (Children Eff. Cov.)	CB	0.26	0.33	0.41	8,923,189.41
	CC	0.24	0.37	0.39	6,069,548.82
VA (Children Eff. Cov.)	CB	0.64	0.20	0.16	13,069,585.95
	CC	0.64	0.20	0.16	23,736,675.25
Folic Acid (WRA Eff. Cov.)	CB	0.18	0.33	0.49	17,096,476.30
	CC	0.24	0.37	0.39	1,550,946.68
All (Lives Saved)	CB	0.37	0.35	0.28	26,614.00
	CC	0.27	0.33	0.40	11,511,343.00
All (Lives Saved, Alt.)	CB	0.45	0.30	0.25	32,849.00
	CC	0.42	0.26	0.31	14,577,066.00

Note: TB= Total Benefits and TC= Total Costs

saved simulations. In the case of the micronutrient effective coverage simulations, the optimal solution is the BAU* scenario for zinc and folic acid. For vitamin A, in contrast to the BAU* scenario, fortified boullion cube is also chosen.

When we consider all interventions together, however, the optimal choice becomes different than the BAU* (which is just a composite of the micronutrient scenarios' BAU).¹ *In contrast to the BAU*, the optimally chosen set of interventions may not include VAS campaign, and may include an addition of folic acid fortified boullion cube. Folic acid fortified flour is included at 1.65 mg/kg, not 5.0 mg/kg.

Note that, although some of the interventions chosen optimally are the same as in the BAU*, the timing and spatial distribution of each intervention may not be the same. Since MINIMOD chooses the optimal set of interventions across space and time, it may be that the appearance of an intervention may happen earlier or later in time than others and may only occur in certain parts of the country.

The BAU* scenarios all assume a constant set of interventions across space and time.

We can also see the differences in how each region is affected after 10 years in terms of accumulated benefits and costs.

*@tbl:opt_space shows the accumulated benefits and costs for each simulation across space. Note that the units of the benefits are dependent on the simulation.

We can also see how benefits accumulate over time, in +@tbl:opt_time.

To illustrate how different these outcomes are to the BAU* scenarios, +@fig:bau_comp shows the difference in cost-per child on a bar graph for each simulation, side by side.

```

113 import matplotlib.pyplot as plt
114
115 cpb = (
116     df_all
117     .loc[(slice(None), 10), :]
118     .sum()
119     .unstack()
120     .assign(cost_per_benefit = lambda df: df['CC']/df['CB'])
121     [['BAU* Cost per Benefit', 'cost_per_benefit']]
122     .rename({'cost_per_benefit' : 'Optimal Cost per Benefit'}, axis=1)
123 )
124
```

¹Notice that there are two alternative definitions for the lives saved estimates of each nutritional intervention. The resulting optimally chosen interventions are similar, but the alternative specification includes VAS campaign as well.

Table 3: Optimal Cumulative Benefits over Time

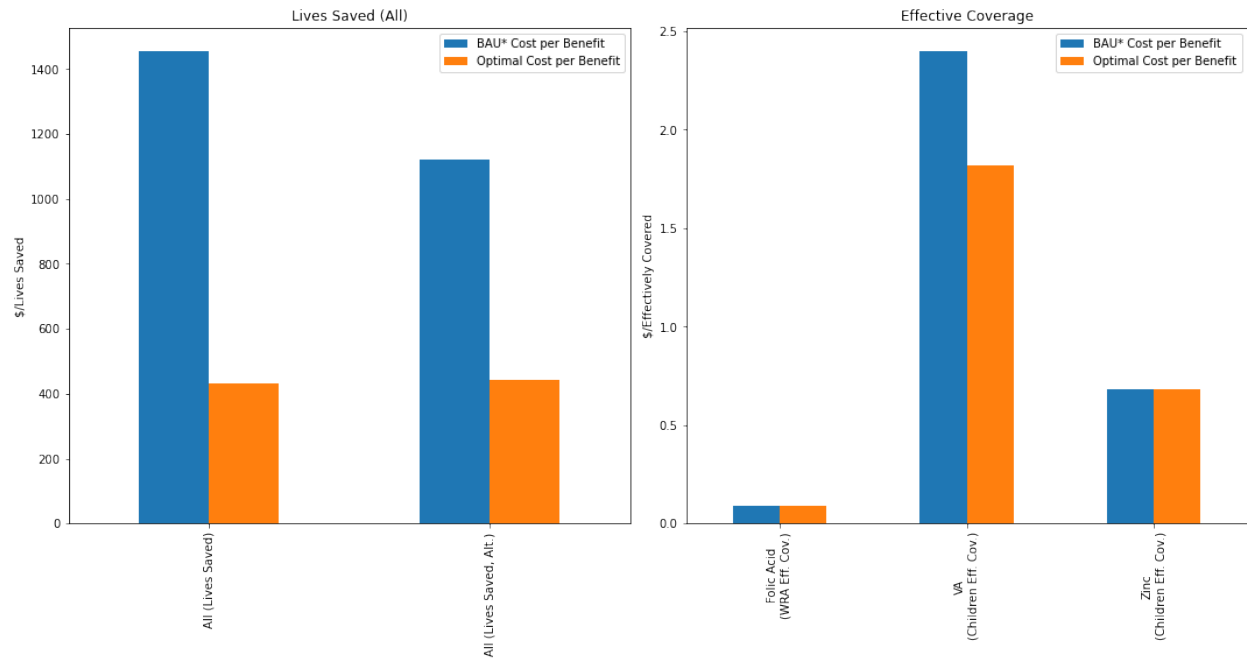
Time	Zinc (Children EC) CB	VA (Children EC) CB	Folic Acid (WRA EC) CB	All (LS) CB	All (LS, Alt.) CB
1	833,571	6,011,853	0	2,290	3,741
2	1,679,604	7,371,118	0	4,611	7,524
3	2,538,338	2,521,598	1,928,972	6,962	11,350
4	3,410,396	3,965,198	3,915,128	9,658	14,299
5	4,296,229	5,428,330	5,958,749	12,390	17,287
6	5,195,240	6,911,502	8,061,227	15,160	20,317
7	6,107,563	8,416,423	10,225,700	17,967	23,387
8	7,033,162	9,944,016	12,452,854	20,811	26,498
9	7,971,739	11,494,698	14,743,015	23,695	29,653
10	8,923,189	13,069,586	17,096,476	26,614	32,849

Note: CB= Cumulative Benefits over a 10 year period, EC= Effective Coverage and LS=Lives Saved.

```

125 fig, (ax1,ax2) = plt.subplots(1,2, figsize= (15,8))
126
127 all_ls = (
128     cpb
129     .loc[cpb.index.str.contains('All')]
130     .assign(perc_change= lambda df: (df['BAU* Cost per Benefit'] - df['Optimal Cost per Benefit'])/df['
131 )
132 all_ls[['BAU* Cost per Benefit', 'Optimal Cost per Benefit']].plot.bar(ax=ax1)
133 eff_cov = cpb.loc[~cpb.index.str.contains('All')].assign(perc_change= lambda df: (df['BAU* Cost per Ben
134 eff_cov[['BAU* Cost per Benefit', 'Optimal Cost per Benefit']].plot.bar(ax=ax2)
135
136 ax1.set_title("Lives Saved (All)")
137 ax2.set_title("Effective Coverage")
138
139 ax1.set_ylabel("$ /Lives Saved")
140 ax2.set_ylabel("$ /Effectively Covered")
141
142 ax1.set_xticklabels(ax1.get_xticklabels(), rotation = 90)
143 ax2.set_xticklabels(ax2.get_xticklabels(), rotation = 90)
144
145 fig.text(.5,-.1, "Note: EC= Effective Coverage", ha='center')
146
147
148 plt.tight_layout()
149
150 plt.savefig("optimization_work/demewoz_lives_saved/reports/multi_mn_plus_lives_saved/bauvsopt.png", dpi=

```



Note: EC= Effective Coverage

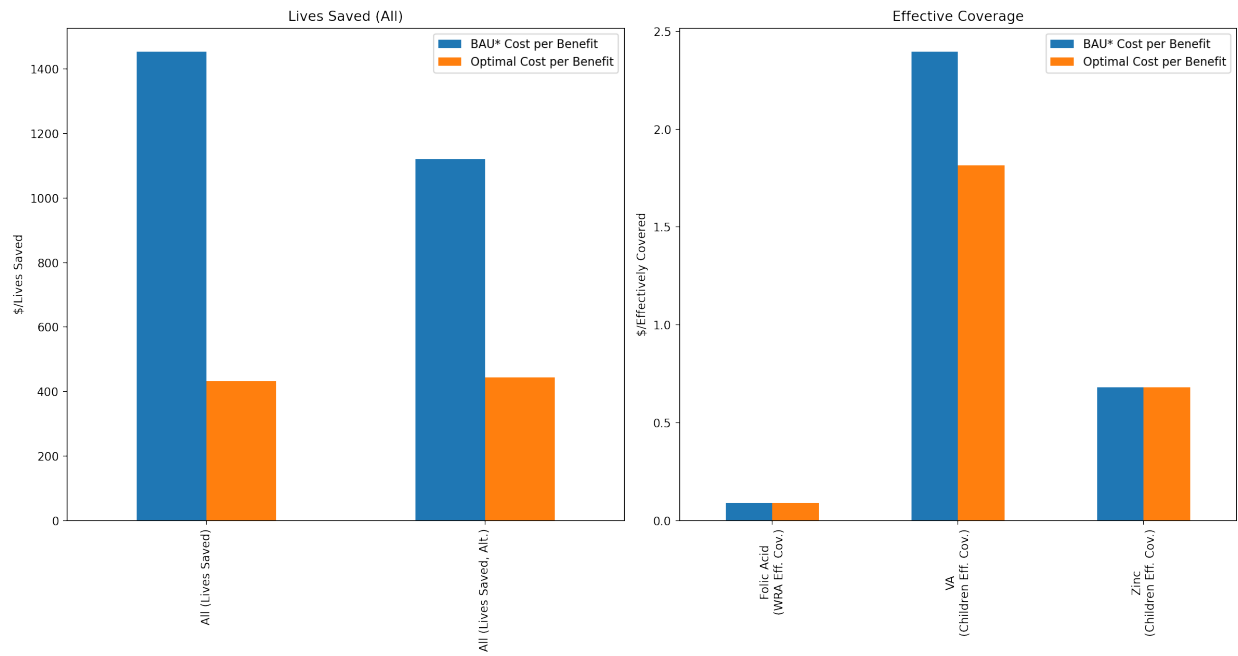


Figure 1: Cost per Benefit Comparison of BAU* vs. Optimal

In the following sections, we will look at relevant maps that will show us how accumulated benefits evolve through time. For these sections we will use shorthand for interventions, outlined in the following table:

Intervention	Abbreviation
Zinc Flour (95 mg/kg)	zflour
VA Oil 9mg/kg	oil
VAS-CHD	vas
VA Cube (80 mg/kg)	cube
Folic Acid Cube (100 mg/kg)	fcube
Folic Acid Flour (5 mg/kg)	fflour
Folic Acid Flour (1.65 mg/kg)	fflour33

Effective Coverage Simulations

Vitamin A

```

152 # Load data
153 geo_df = gpd.read_file("examples/data/maps/cameroon/CAM.shp")
154
155 # Now we create the boundaries for North, South and Cities
156 # Based on "Measuring Costs of Vitamin A...", Table 2"
157 north = r"Adamaoua|Nord|Extreme-Nord"
158 south = r"Centre|Est|Nord-Ouest|Ouest|Sud|Sud-Ouest"
159 cities= r"Littoral" # Duala
160 # Yaounde is in Mfoundi
161 geo_df.loc[lambda df: df['ADM1'].str.contains(north), 'space'] = 'North'
162 geo_df.loc[lambda df: df['ADM1'].str.contains(south), 'space'] = 'South'
163 geo_df.loc[lambda df: df['ADM1'].str.contains(cities), 'space'] = 'Cities'
164 geo_df.loc[lambda df: df['ADM2'].str.contains(r"Mfoundi"), 'space'] = 'Cities'
165
166 # Now we aggregate the data to the `space` variable
167 agg_geo_df = geo_df.dissolve(by = 'space')
168
169
170 vas.models[None].plot_map_benchmark(intervention = None,
171 time = 1,
172 optimum_interest = 'cb',
173 bench_intervention = 'oilvas',
174 map_df = agg_geo_df,
175 merge_key = 'space',
176 intervention_in_title = False,
177 intervention_bubbles= True,
178 intervention_bubble_names = ['oil', 'vas', 'cube'],
179 bau_intervention_bubble_names = ['oil', 'vas', 'cube'],
180 save = 'optimization_work/demewoz_lives_saved/reports/multi_mn_plus_lives_saved/vas1.png')
181
182 plt.gcf().text(0.5,-.05, 'Note: Effective coverage (in millions)', ha='center')
183
184 vas.models[None].plot_map_benchmark(intervention = None,
185 time = 3,
186 optimum_interest = 'cb',
187 bench_intervention = 'oilvas',
188 map_df = agg_geo_df,
189 merge_key = 'space',
190 intervention_in_title = False,
191 intervention_bubbles= True,

```

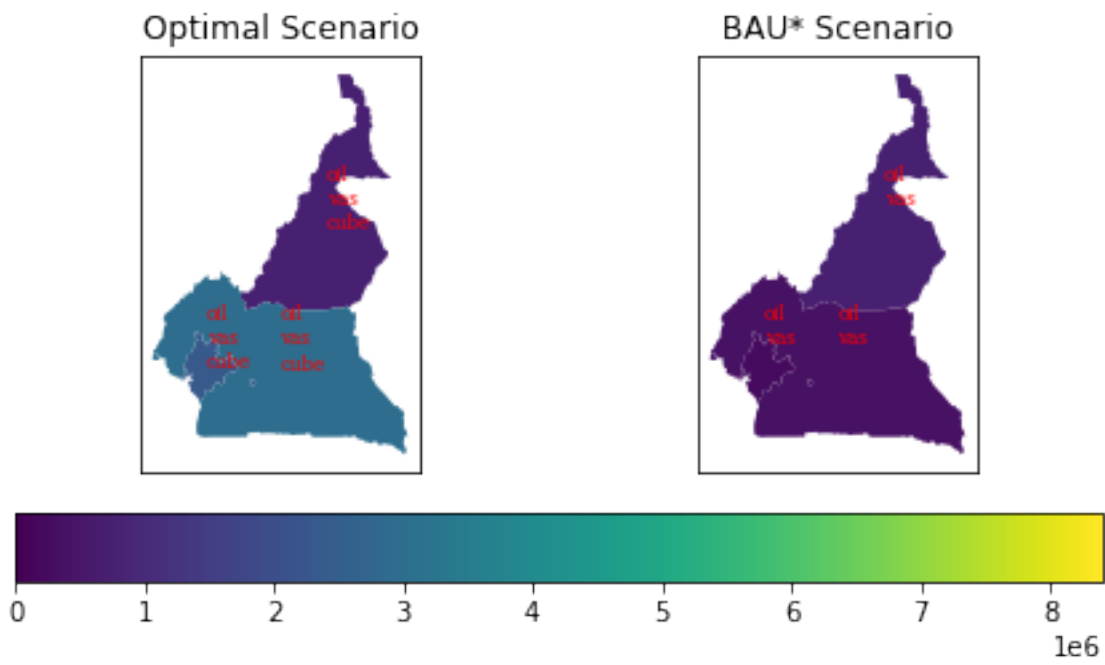


```

192 intervention_bubble_names = ['oil', 'vas', 'cube'],
193 bau_intervention_bubble_names = ['oil', 'vas', 'cube'],
194 save = 'optimization_work/demewoz_lives_saved/reports/multi_mn_plus_lives_saved/vas3.png')
195
196 plt.gcf().text(0.5,-.05, 'Note: Effective coverage (in millions)', ha='center')
197
198 vas.models[None].plot_map_benchmark(intervention = None,
199 time = 10,
200 optimum_interest = 'cb',
201 bench_intervention = 'oilvas',
202 map_df = agg_geo_df,
203 merge_key = 'space',
204 intervention_in_title = False,
205 intervention_bubbles= True,
206 intervention_bubble_names = ['oil', 'vas', 'cube'],
207 bau_intervention_bubble_names = ['oil', 'vas', 'cube'],
208 save = 'optimization_work/demewoz_lives_saved/reports/multi_mn_plus_lives_saved/vas10.png')
209
210 plt.gcf().text(0.5,-.05, 'Note: Effective coverage (in millions)', ha='center')
211
212 fig, (ax1, ax2) = plt.subplots(1,2, figsize=(10,6))
213
214 vas.models[None].plot_bau_time('b', ax=ax1)
215 vas.models[None].plot_bau_time('c', ax=ax2)
216 ax1.set_ylabel("Millions Effectively Covered")
217 ax2.set_ylabel("USD (in millions)")
218
219 plt.tight_layout()
220 plt.savefig('optimization_work/demewoz_lives_saved/reports/multi_mn_plus_lives_saved/vas_b_c.png', dpi=
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,
/home/lordflaron/Documents/minimod/minimod/utils/plotting.py:250: MatplotlibDeprecationWarning: The 's'
df_bubble_final.apply(lambda x: ax.annotate(s = x.bubble_name,

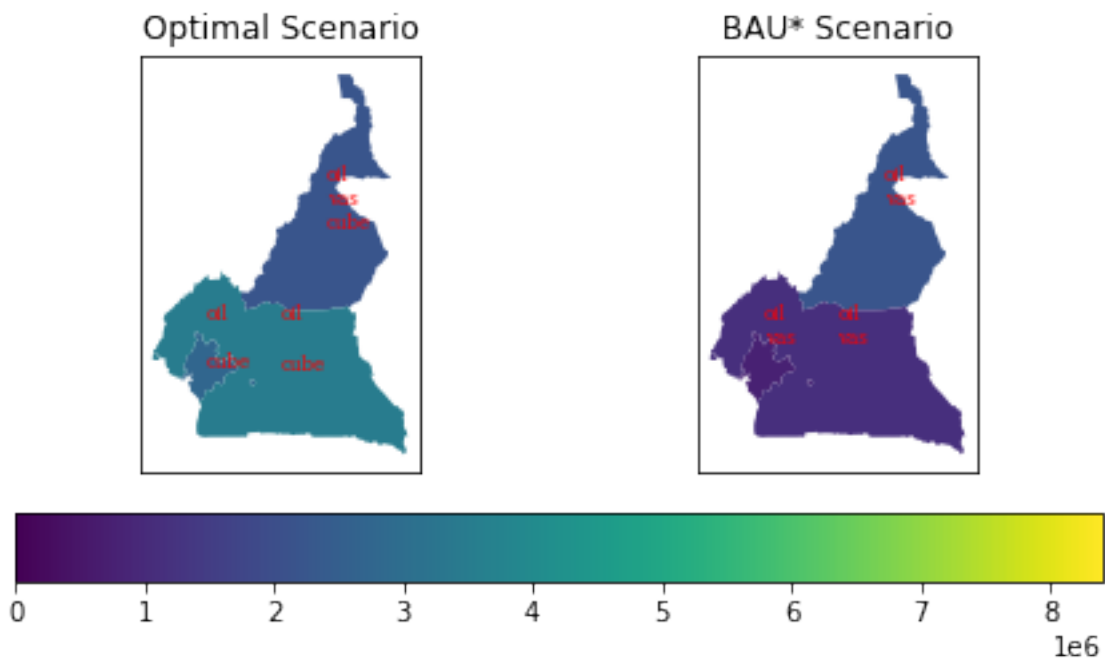
```

Cumulative Children Effectively Covered (VA)



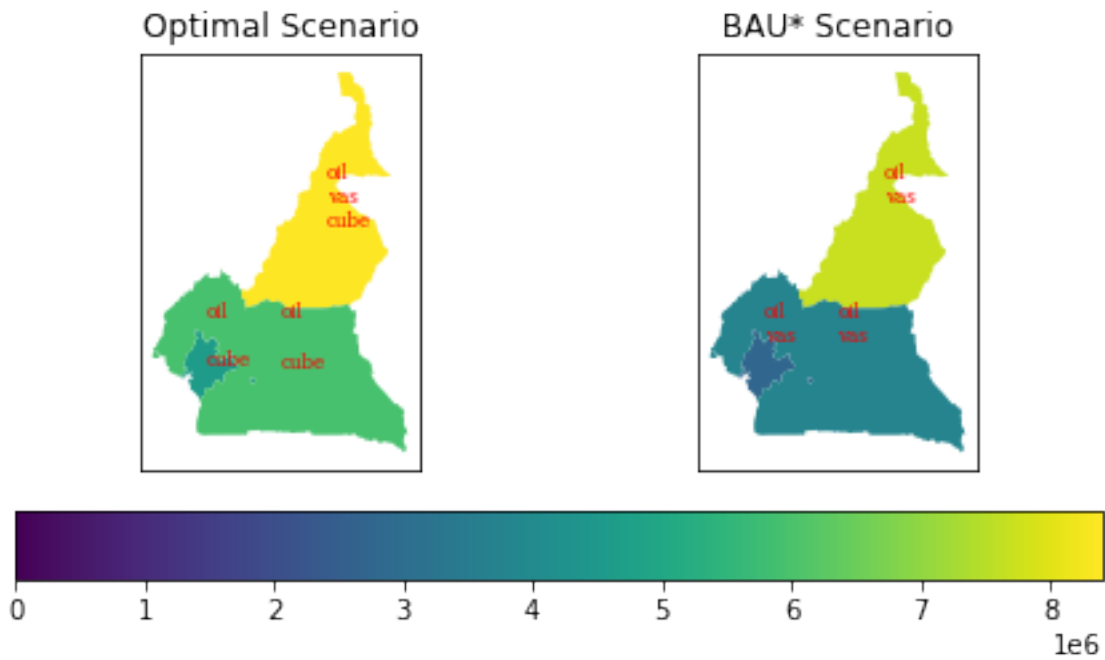
Note: Colors describe Cumulative Children Effectively Covered (VA) (in millions)

Cumulative Children Effectively Covered (VA)

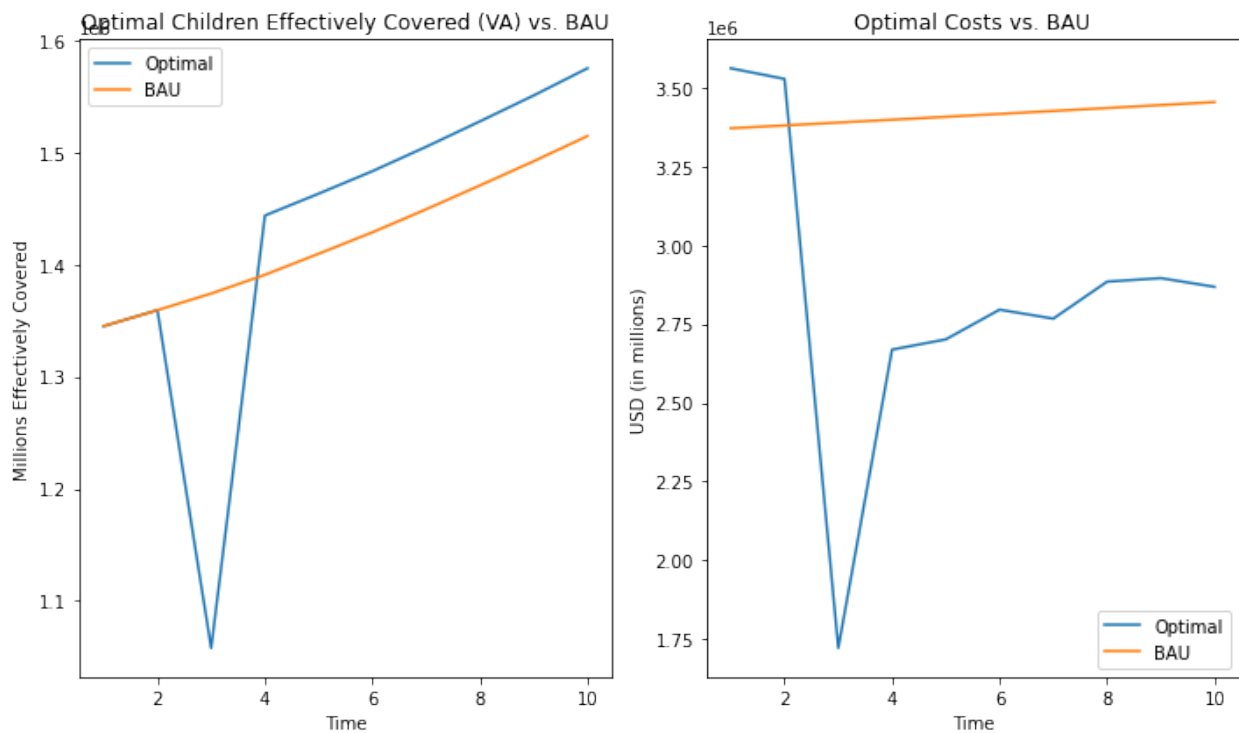


Note: Colors describe Cumulative Children Effectively Covered (VA) (in millions)

Cumulative Children Effectively Covered (VA)



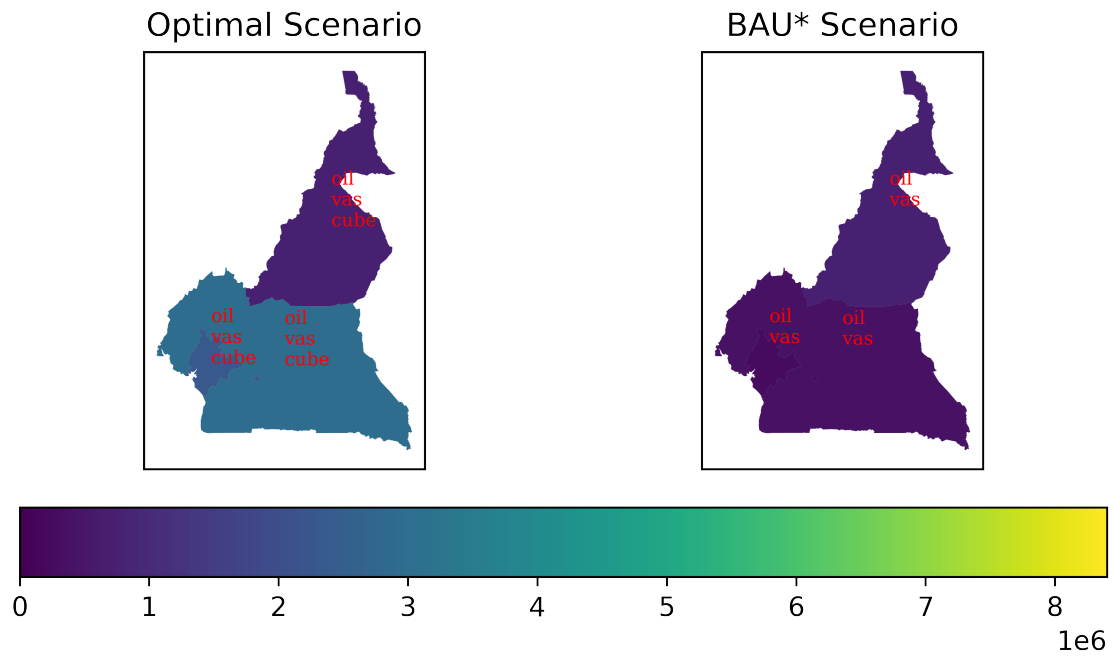
Note: Colors describe Cumulative Children Effectively Covered (VA) (in millions)



For the Vitamin A simulations, we start with fortified oil (at 75% coverage), fortified cube and a VAS routine

as can be seen in +@fig:vas1. By Period 4 (in +@fig:vas3), the south and cities stop VAS campaigns, but proceed with cube and oil interventions. This continues until the last period. If we compare this to the BAU* scenario, accumulated benefits have reached around the same levels in the South and cities, but with higher benefits in the North.

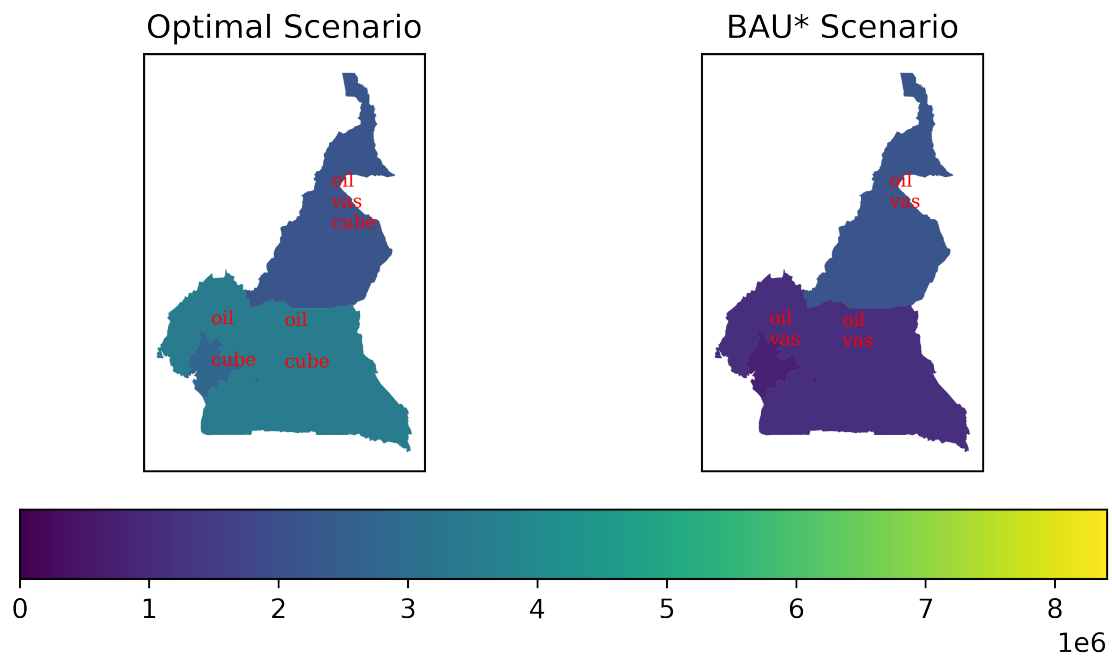
Cumulative Children Effectively Covered (VA)



Note: Colors describe Cumulative Children Effectively Covered (VA) (in millions)

Figure 2: Vitamin A Accumulated Benefits, T=1

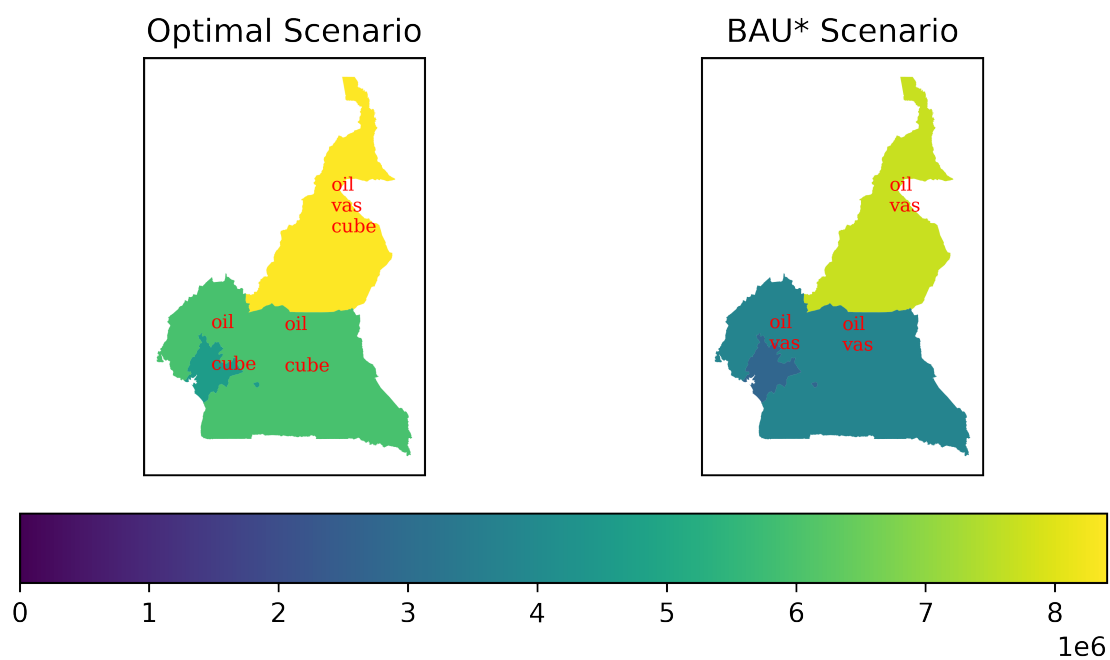
Cumulative Children Effectively Covered (VA)



Note: Colors describe Cumulative Children Effectively Covered (VA) (in millions)

Figure 3: Vitamin A Accumulated Benefits, T=3

Cumulative Children Effectively Covered (VA)



Note: Colors describe Cumulative Children Effectively Covered (VA) (in millions)

Figure 4: Vitamin A Accumulated Benefits, T=10

To illustrate the differences between benefits and costs between the optimal and BAU* scenarios, we compare the two in `fig:vas_b_c`.

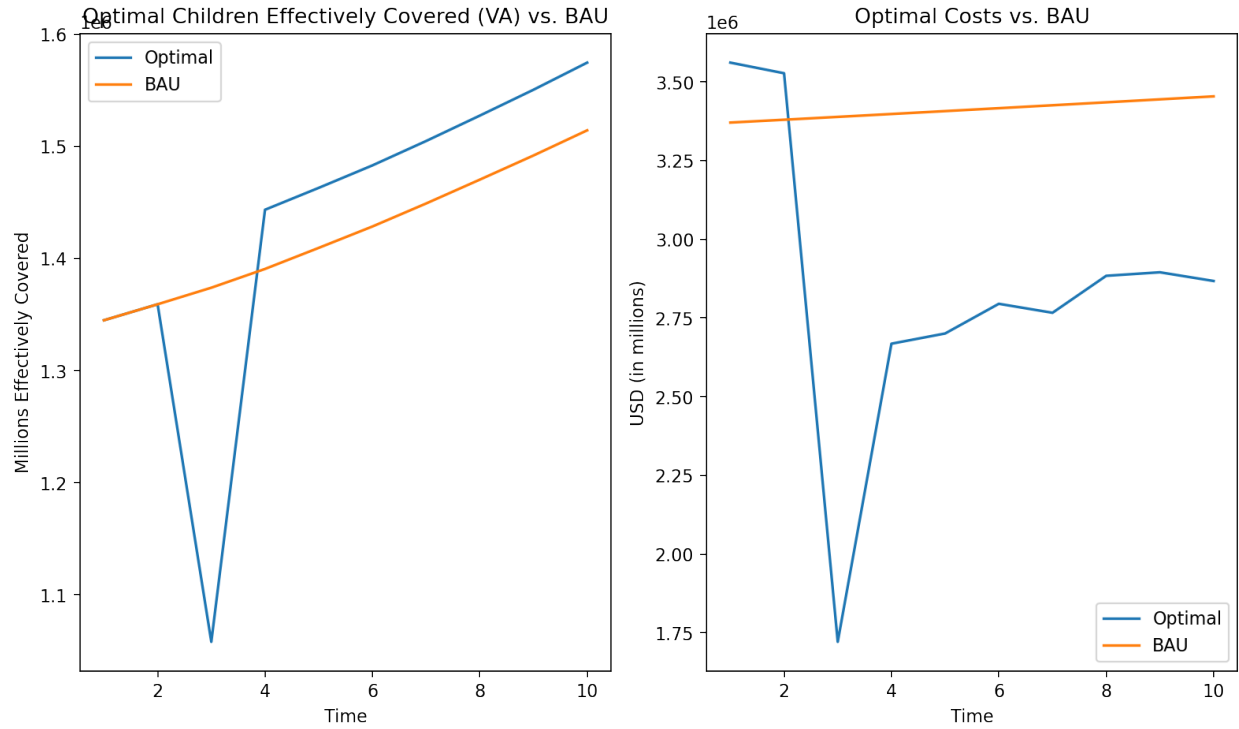
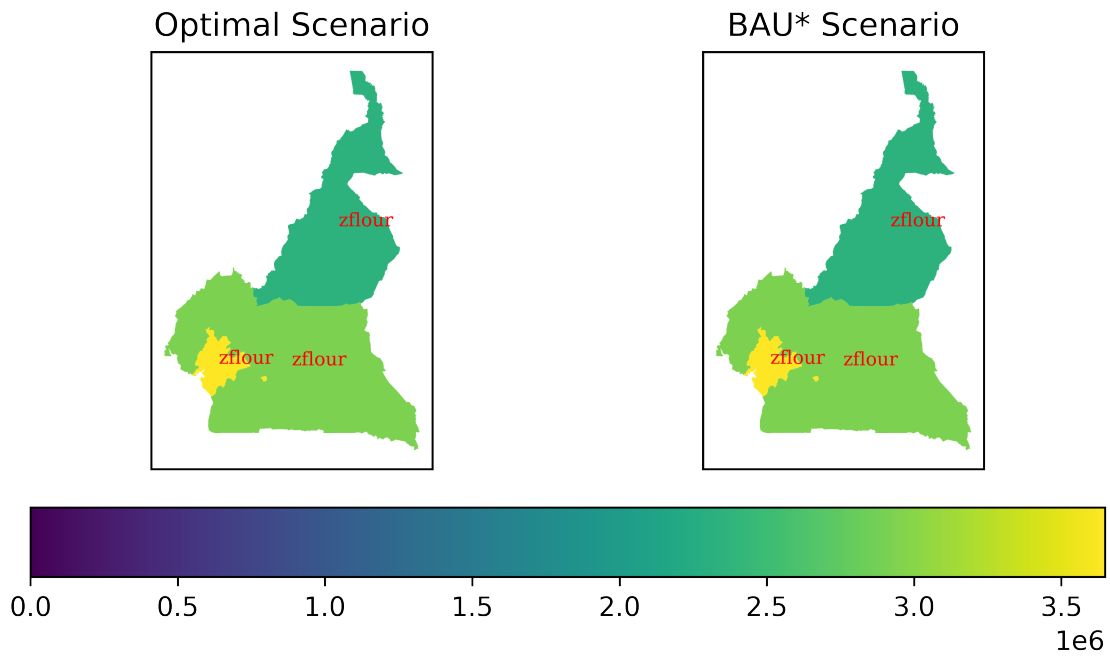


Figure 5: Vitamin A Per-year Benefits and Costs across Time

Zinc

For zinc interventions, we see that zinc fortified flour is chosen, which is the same as the BAU* scenario. This leads to the highest benefits being in the cities, followed by the south and then the North.

Cumulative Children Effectively Covered (Zinc)



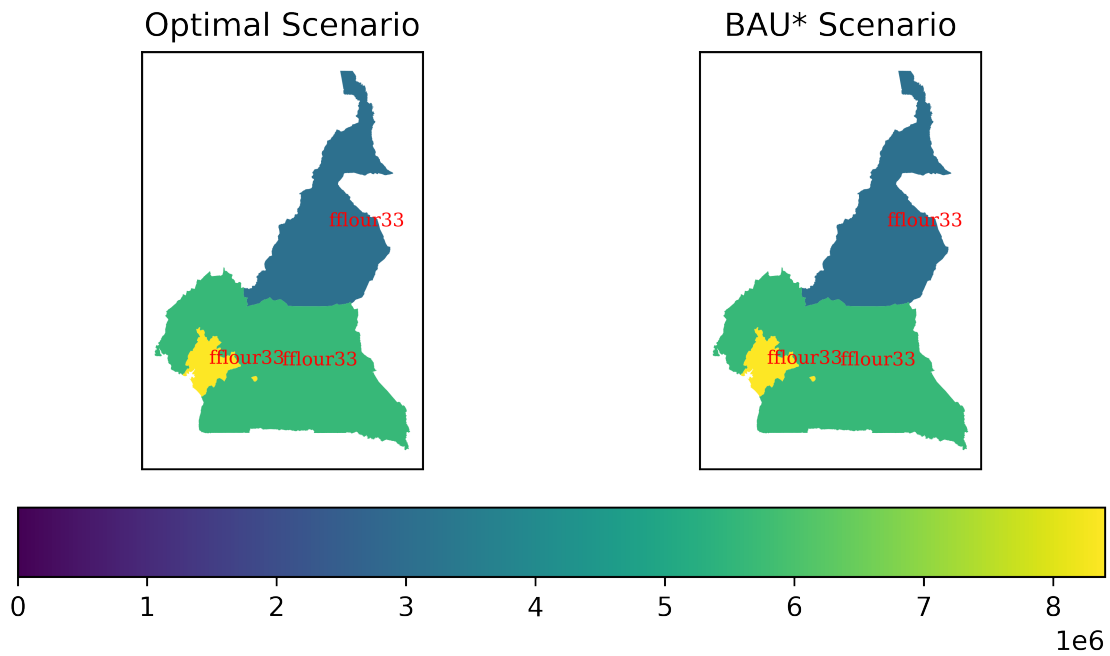
Note: Colors describe Cumulative Children Effectively Covered (Zinc) (in millions)

Figure 6: Zinc Accumulated Benefits, T=10

Folic Acid

Folic Acid is the same as the zinc interventions, in that the BAU* intervention is chosen as the optimal intervention. Cities are disproportionately affected, compared to the South and the North.

Cumulative WRA Effectively Covered (Folate)



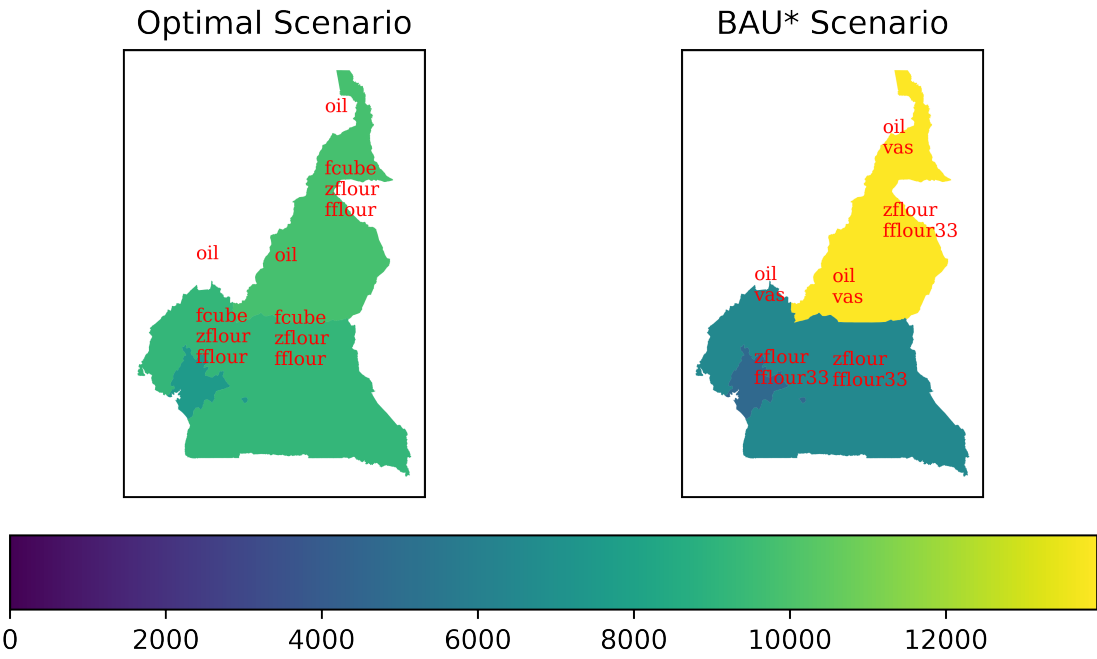
Note: Colors describe Cumulative WRA Effectively Covered (Folate) (in millions)

Figure 7: Folic Acid Accumulated Benefits, T=10

Lives Saved

For Lives Saved, we find that the same intervention is chosen for all periods everywhere leading to accumulated benefits that are more proportionately distributed across the country, while the BAU* scenario has higher benefits in the North.

Cumulative Lives Saved



Note: Colors describe Cumulative Lives Saved (in millions)

Figure 8: Lives Saved Accumulated Benefits, T=10

We also show the difference in per-year benefits and costs in +@fig:ls_b_c.

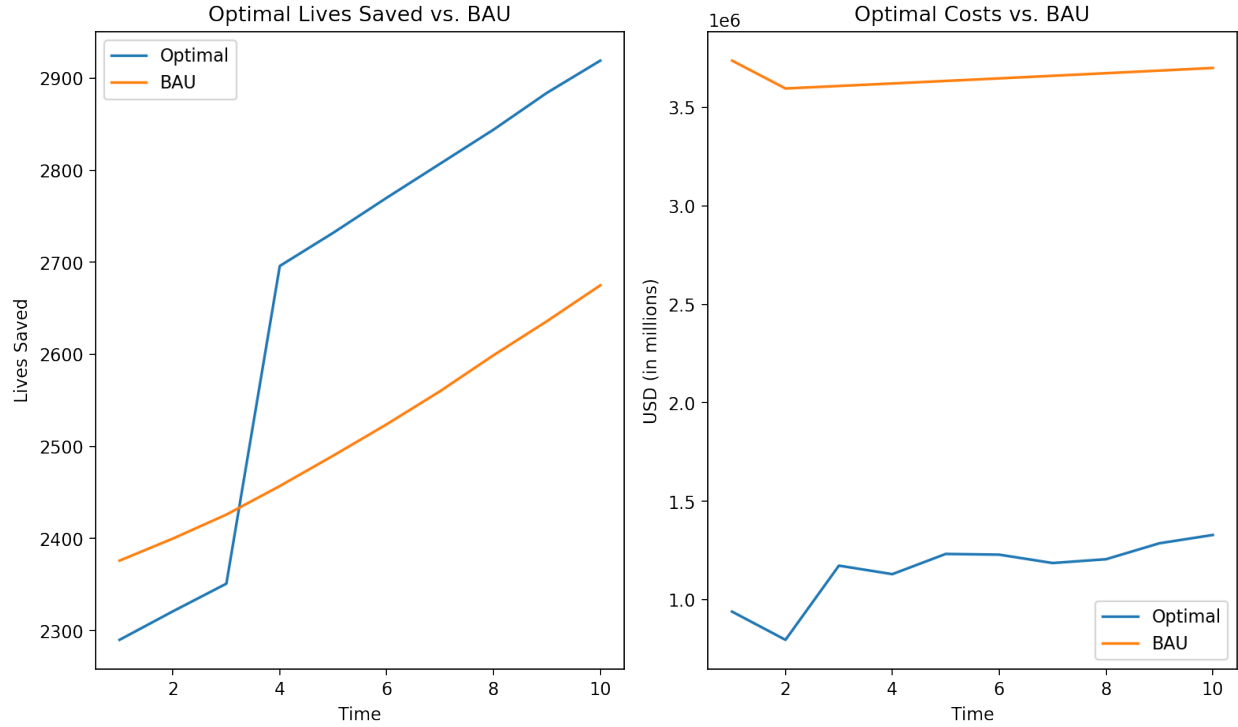
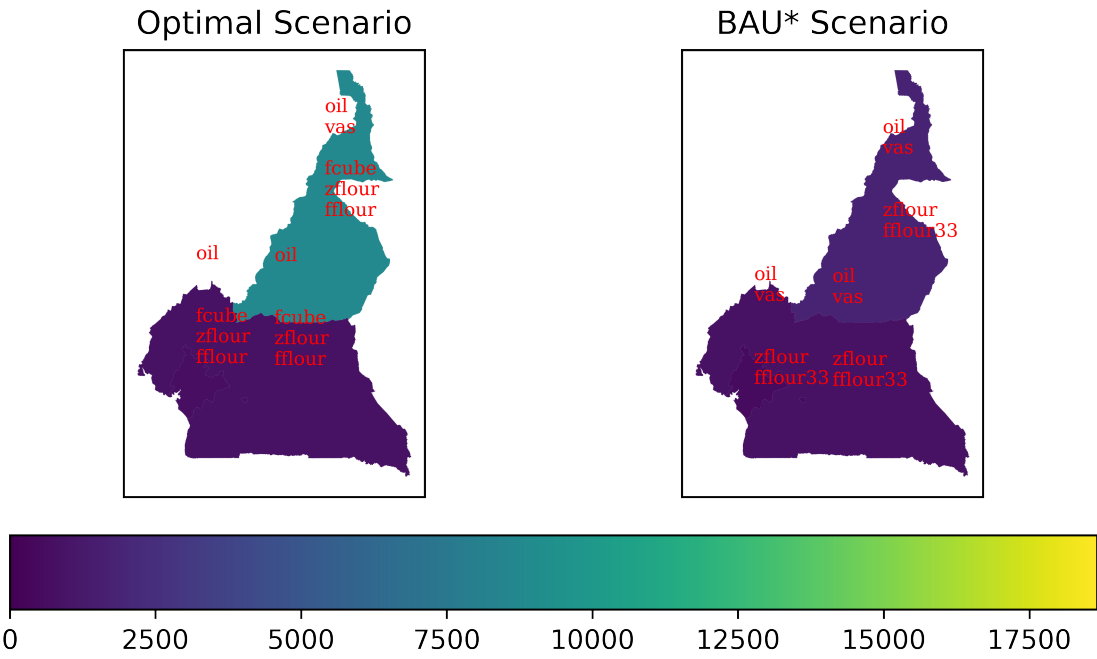


Figure 9: Lives Saved Per-Year Benefits and Costs

Lives Saved Alternative Definition

For the alternative definition of lives saved, we find that the same interventions are chosen across the country, apart from VAS campaigns in the north in periods 1-3 (+@fig:lshigh1). By period 4, we find that VAS campaigns stop and for the rest of time, the same interventions are used for all of the country. This leads to a similar geographic distribution by the end (+@fig:lshigh10) as the other lives saved definition, but with relatively higher benefits for the North.

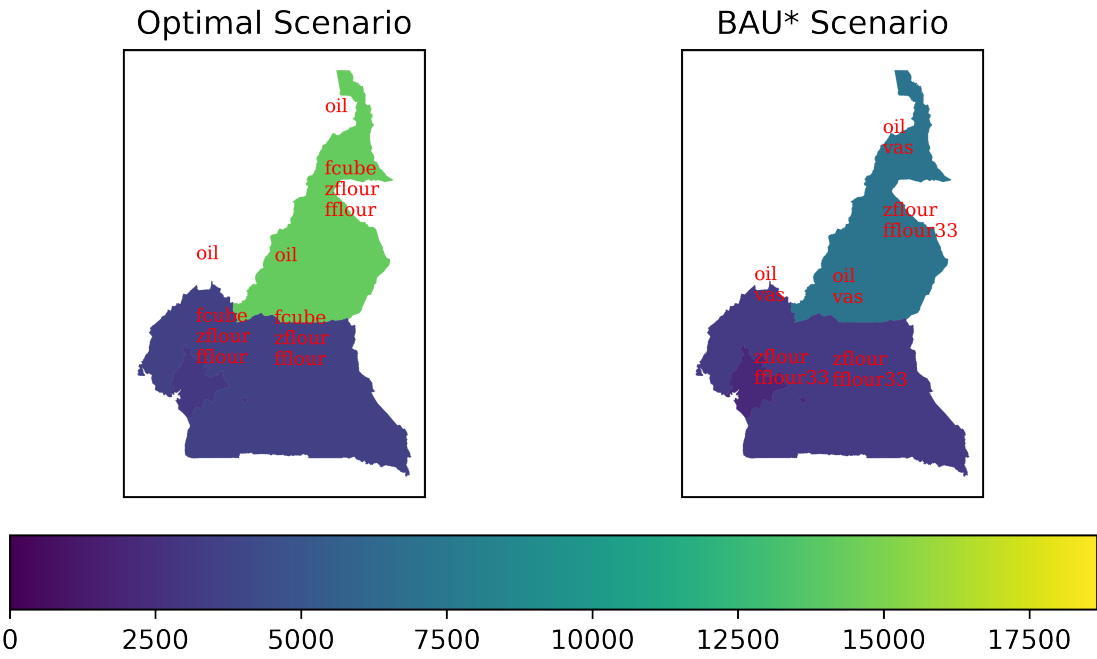
Cumulative Lives Saved



Note: Colors describe Cumulative Lives Saved (in millions)

Figure 10: Lives Saved Alternative Specification Accumulated Benefits, T=1

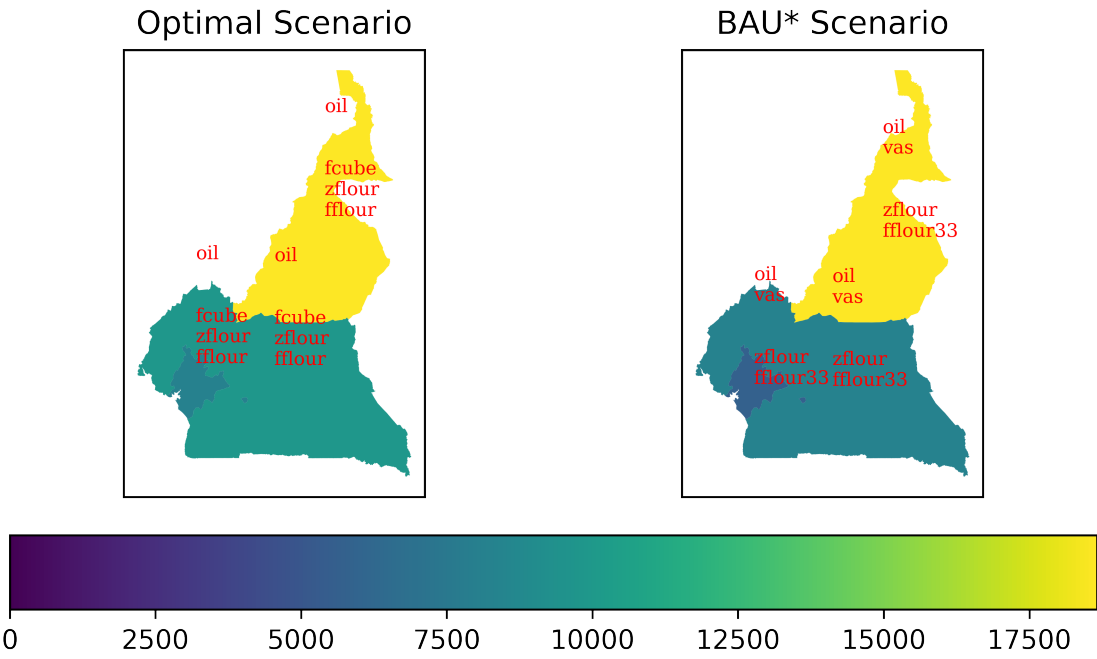
Cumulative Lives Saved



Note: Colors describe Cumulative Lives Saved (in millions)

Figure 11: Lives Saved Alternative Specification Accumulated Benefits, T=4

Cumulative Lives Saved



Note: Colors describe Cumulative Lives Saved (in millions)

Figure 12: Lives Saved Alternative Specification Accumulated Benefits, T=10

We also show the difference in per-year benefits and costs in +@fig:lshigh_b_c.

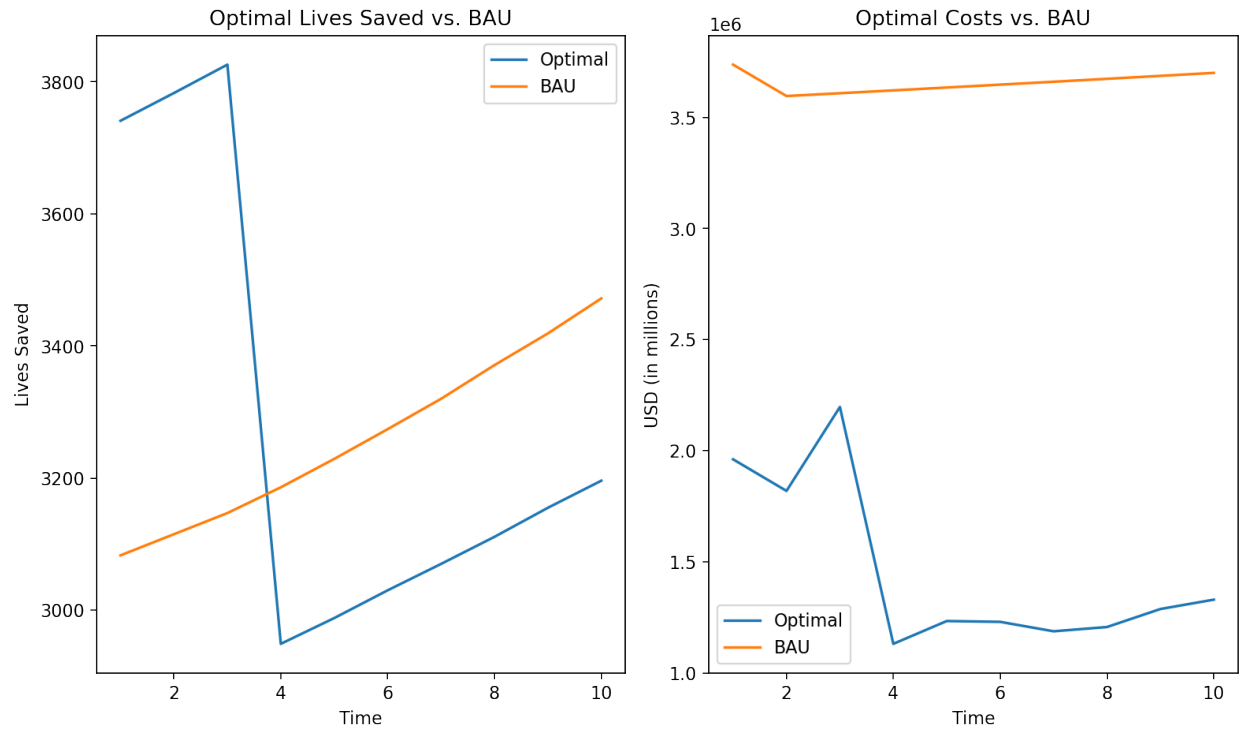


Figure 13: Lives Saved (Alt. Definition) Per-Year Benefits and Costs