

# Report

## Project 2: Smart Energy with Optimization Algorithms

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### Task 1: Solve the offline optimization problem, e.g., using tools CVX in Matlab or Python.

- Used CVXOPT library of python to calculate the offline optimization problem.
- True values are used to calculate the cost function and then took minimum cost function out of it using CVXOPT library.
- This is going to be the **datum** on which other algorithms are going to be judged.

#### **Part1: Static solution**

Fixed provision value i.e. x was used.

#### **Part2: Dynamic solution**

x varies as x (1), x (2), ..., x (672).

```
For home B:  
Offline Optimization Problem solution for static is: 1018.4013103622867  
Offline Optimization Problem solution for dynamic is: 836.0104660550957  
For home C:  
Offline Optimization Problem solution for static is: 1365.6704999799842  
Offline Optimization Problem solution for dynamic is: 1137.486313296603  
For home F:  
Offline Optimization Problem solution for static is: 6286.343226100506  
Offline Optimization Problem solution for dynamic is: 5232.1411726006945
```

### Task 2(i): Online gradient descent (with different step size)

- Using different kind of step sizes and by calculating partial derivative of cost function w.r.t x (t) to optimize the solution, we have calculated an optimized solution.

#### **Type of Step Sizes:**

```
step_size_type = 'A': 1/sqrt(t)
```

```
[53] print("For Home B,")  
      OGD_results_B = OGD_online_op(homeB, p_t = 0.40, a = 4, b= 4,step_size_type = 'A')  
      print("For Home C,")  
      OGD_results_C = OGD_online_op(homeC, p_t = 0.40, a = 4, b= 4,step_size_type = 'A')  
      print("For Home F,")  
      OGD_results_F = OGD_online_op(homeF, p_t = 0.40, a = 4, b= 4,step_size_type = 'A')
```

```
➤ For Home B,  
  OGD Optimization Problem solution is: 1892.720581127459  
  For Home C,  
  OGD Optimization Problem solution is: 2143.127367233312  
  For Home F,  
  OGD Optimization Problem solution is: 7521.536038487382
```

step\_size\_type = 'B': 1/(t)

```
print("For Home B,")
OGD_results_B = OGD_online_op(homeB, p_t = 0.40, a = 4, b= 4,step_size_type = 'B')
print("For Home C,")
OGD_results_C = OGD_online_op(homeC, p_t = 0.40, a = 4, b= 4,step_size_type = 'B')
print("For Home F,")
OGD_results_F = OGD_online_op(homeF, p_t = 0.40, a = 4, b= 4,step_size_type = 'B')
```

```
For Home B,
OGD Optimization Problem solution is: 1583.378957866783
For Home C,
OGD Optimization Problem solution is: 1614.6201205975656
For Home F,
OGD Optimization Problem solution is: 10516.319202234305
```

step\_size\_type = 'C': 0.01

```
[165] print("For Home B,")
OGD_results_B = OGD_online_op(homeB, p_t = 0.40, a = 4, b= 4,step_size_type = 'C')
print("For Home C,")
OGD_results_C = OGD_online_op(homeC, p_t = 0.40, a = 4, b= 4,step_size_type = 'C')
print("For Home F,")
OGD_results_F = OGD_online_op(homeF, p_t = 0.40, a = 4, b= 4,step_size_type = 'C')
```

```
For Home B,
OGD Optimization Problem solution is: 1244.8327133920006
For Home C,
OGD Optimization Problem solution is: 1771.6468845240013
For Home F,
OGD Optimization Problem solution is: 12635.242333375993
```

step\_size\_type = 'D': 1/672

```
print("For Home B,")
OGD_results_B = OGD_online_op(homeB, p_t = 0.40, a = 4, b= 4,step_size_type = 'D')
print("For Home C,")
OGD_results_C = OGD_online_op(homeC, p_t = 0.40, a = 4, b= 4,step_size_type = 'D')
print("For Home F,")
OGD_results_F = OGD_online_op(homeF, p_t = 0.40, a = 4, b= 4,step_size_type = 'D')
```

```
For Home B,
OGD Optimization Problem solution is: 1744.7828514939033
For Home C,
OGD Optimization Problem solution is: 2804.8918305921875
For Home F,
OGD Optimization Problem solution is: 17120.758057329414
```

- OGD is not a good optimizer use for time series data as this data is more dynamic like stock market data. OGD is better for static objective functions.

## Comment about different Step size of OGD for different home:

For Home B, best step size is `step_size_type = 'C': 0.01` with minimum cost = 1244.8327133920006

For Home C, best step size is `step_size_type = 'B': 1/(t)` with minimum cost = 1614.6201205975656

For Home F, best step size is `step_size_type = 'A': 1/sqrt(t)` with minimum cost = 7521.536038487382

### Task 2(ii): Receding horizon control (with different prediction window size):

- With  $v = 1$  and `predictHorizon = window_size`, fixed  $a = 4$  and  $b = 4$ , we have developed RHC function to optimize the cost function.
- Unlike OGD, this uses prediction values for computing. And it handles dynamic data better than OGD.
- **Prediction Algorithm 1: ARIMA**

For ARIMA Model Predictions:

```

window_size = [3,5,10]
for window in window_size:
    print("For window_size: ", window)
    [optimal_values_RHC_B, result_RHC_B] = control_algo_online_op(homeB, p_t = 0.40, a = 4, b = 4, predictionHorizon = window)
    print("For Home B, RHC Optimization Problem solution is: ", result_RHC_B)
    [optimal_values_RHC_C, result_RHC_C] = control_algo_online_op(homeC, p_t = 0.40, a = 4, b = 4, predictionHorizon = window)
    print("For Home C, RHC Optimization Problem solution is: ", result_RHC_C)
    [optimal_values_RHC_F, result_RHC_F] = control_algo_online_op(homeF, p_t = 0.40, a = 4, b = 4, predictionHorizon = window)
    print("For Home F, RHC Optimization Problem solution is: ", result_RHC_F)

For window_size: 3
For Home B, RHC Optimization Problem solution is: 750.3907458132975
For Home C, RHC Optimization Problem solution is: 989.8302788430145
For Home F, RHC Optimization Problem solution is: 4551.13984721234
For window_size: 5
For Home B, RHC Optimization Problem solution is: 750.3907458132975
For Home C, RHC Optimization Problem solution is: 989.8302788430145
For Home F, RHC Optimization Problem solution is: 4551.13984721234
For window_size: 10
For Home B, RHC Optimization Problem solution is: 750.3907458132975
For Home C, RHC Optimization Problem solution is: 989.8302788430145
For Home F, RHC Optimization Problem solution is: 4551.13984721234

```

- **Prediction Algorithm 2: Linear Regression**

For Linear Regression Model Predictions:

```

[172] window_size = [3,5,10]
for window in window_size:
    print("For window_size: ", window)
    [optimal_values_RHC_B_LR, result_RHC_B_LR] = control_algo_online_op(homeB_LR, p_t = 0.40, a = 4, b = 4, predictionHorizon = window)
    print("For Home B, RHC Optimization Problem solution is: ", result_RHC_B_LR)
    [optimal_values_RHC_C_LR, result_RHC_C_LR] = control_algo_online_op(homeC_LR, p_t = 0.40, a = 4, b = 4, predictionHorizon = window)
    print("For Home C, RHC Optimization Problem solution is: ", result_RHC_C_LR)
    [optimal_values_RHC_F_LR, result_RHC_F_LR] = control_algo_online_op(homeF_LR, p_t = 0.40, a = 4, b = 4, predictionHorizon = window)
    print("For Home F, RHC Optimization Problem solution is: ", result_RHC_F_LR)

For window_size: 3
For Home B, RHC Optimization Problem solution is: 721.894394948552
For Home C, RHC Optimization Problem solution is: 1005.4272601940075
For Home F, RHC Optimization Problem solution is: 4414.920796928689
For window_size: 5
For Home B, RHC Optimization Problem solution is: 721.894394948552
For Home C, RHC Optimization Problem solution is: 1005.4272601940075
For Home F, RHC Optimization Problem solution is: 4414.920796928689
For window_size: 10
For Home B, RHC Optimization Problem solution is: 721.894394948552
For Home C, RHC Optimization Problem solution is: 1005.4272601940075
For Home F, RHC Optimization Problem solution is: 4414.920796928689

```

- Best window size with different prediction algorithms:

## Comment about Best window\_size of RHC:

For ARIMA Predictions:

For window\_size: 5

For Home B, RHC Optimization Problem solution is: 750.3907458132975

For Home C, RHC Optimization Problem solution is: 989.8302788430145

For Home F, RHC Optimization Problem solution is: 4551.13984721234

For Linear Regression Predictions:

For window\_size: 5

For Home B, RHC Optimization Problem solution is: 721.894394948552

For Home C, RHC Optimization Problem solution is: 1005.4272601940075

For Home F, RHC Optimization Problem solution is: 4414.920796928689

### Task 2(iii): Commitment Horizon control (with different commitment levels):

- Additional to RHC parameters, CHC has another parameter which we handle with i.e **commitment level, v**. We have tried different commitment levels for all the home.
- **Prediction Algorithm 1: ARIMA**

For ARIMA Model Predictions:

```
commitmentHorizon = [3,4,5]
for commitment in commitmentHorizon:
    print("For v: ", commitment)
    [optimal_values_CHC_B, result_CHC_B] = control_algo_online_op(homeB, p_t = 0.40, a = 4, b = 4, predictionHorizon = 5, commitmentHorizon= commitment)
    print("For Home B, CHC Optimization Problem solution is: ", result_CHC_B)
    [optimal_values_CHC_C, result_CHC_C] = control_algo_online_op(homeC, p_t = 0.40, a = 4, b = 4, predictionHorizon = 5, commitmentHorizon= commitment)
    print("For Home C, CHC Optimization Problem solution is: ", result_CHC_C)
    [optimal_values_CHC_F, result_CHC_F] = control_algo_online_op(homeF, p_t = 0.40, a = 4, b = 4, predictionHorizon = 5, commitmentHorizon= commitment)
    print("For Home F, CHC Optimization Problem solution is: ", result_CHC_F)
```

```
For v: 3
For Home B, CHC Optimization Problem solution is: 750.3907458133029
For Home C, CHC Optimization Problem solution is: 989.8302788430127
For Home F, CHC Optimization Problem solution is: 4551.139847212367
For v: 4
For Home B, CHC Optimization Problem solution is: 750.3907458133042
For Home C, CHC Optimization Problem solution is: 989.8302788430152
For Home F, CHC Optimization Problem solution is: 4551.139847212363
For v: 5
For Home B, CHC Optimization Problem solution is: 750.3907458133061
For Home C, CHC Optimization Problem solution is: 989.8302788430143
For Home F, CHC Optimization Problem solution is: 4551.139847212366
```

- **Prediction Algorithm 2: Linear Regression**

For Linear Regression Model Predictions:

```
[174] commitmentHorizon = [3,4,5]
for commitment in commitmentHorizon:
    print("For v: ", commitment)
    [optimal_values_CHC_B_LR, result_CHC_B_LR] = control_algo_online_op(homeB_LR, p_t = 0.40, a = 4, b = 4, predictionHorizon = 5, commitmentHorizon= 3)
    print("For Home B, CHC Optimization Problem solution is: ", result_CHC_B_LR)
    [optimal_values_CHC_C_LR, result_CHC_C_LR] = control_algo_online_op(homeC_LR, p_t = 0.40, a = 4, b = 4, predictionHorizon = 5, commitmentHorizon= 3)
    print("For Home C, CHC Optimization Problem solution is: ", result_CHC_C_LR)
    [optimal_values_CHC_F_LR, result_CHC_F_LR] = control_algo_online_op(homeF_LR, p_t = 0.40, a = 4, b = 4, predictionHorizon = 5, commitmentHorizon= 3)
    print("For Home F, CHC Optimization Problem solution is: ", result_CHC_F_LR)
```

```
For v: 3
For Home B, CHC Optimization Problem solution is: 721.8943949485557
For Home C, CHC Optimization Problem solution is: 1005.4272601940361
For Home F, CHC Optimization Problem solution is: 4414.920796928767
For v: 4
For Home B, CHC Optimization Problem solution is: 721.8943949485557
For Home C, CHC Optimization Problem solution is: 1005.4272601940361
For Home F, CHC Optimization Problem solution is: 4414.920796928767
For v: 5
For Home B, CHC Optimization Problem solution is: 721.8943949485557
For Home C, CHC Optimization Problem solution is: 1005.4272601940361
For Home F, CHC Optimization Problem solution is: 4414.920796928767
```

- CHC is better than RHC as it keeps an average at a commitment level which helps this control algorithm to handle dynamic data better.

### **Comment about Best Commitment Horizon for CHC:**

#### **For ARIMA Predictions:**

For v: 3

For Home B, CHC Optimization Problem solution is: 750.3907458133029

For Home C, CHC Optimization Problem solution is: 989.8302788430127

For Home F, CHC Optimization Problem solution is: 4551.139847212367

#### **For Linear Regression Predictions:**

For v: 3

For Home B, CHC Optimization Problem solution is: 721.8943949485557

For Home C, CHC Optimization Problem solution is: 1005.4272601940361

For Home F, CHC Optimization Problem solution is: 4414.920796928767

### **Task 3: Compare the costs of these algorithms to those of the offline static and dynamic solutions:**

#### **Calculating Regret Factor**

- We have used Regret value function to compare different algorithms which is calculate by as following:

**Static Regret = Cost of the Algorithm - Static Offline Cost**

**Dynamic Regret = Cost of the Algorithm - Dynamic offline Cost**

- **Regrets for OGD:**

For Home B, static Regret: 226.43140302971392

For Home C, static Regret: 248.94962061758133

For Home F, static Regret: 1235.1928123868756

For Home B, dynamic Regret: 408.822247336905

For Home C, dynamic Regret: 477.13380730096264

For Home F, dynamic Regret: 2289.3948658866875

- **Regrets for RHC for ARIMA:**

For Home B, static Regret: -268.01056454898924

For Home C, static Regret: -375.8402211369697

For Home F, static Regret: -1735.2033788881663

For Home B, dynamic Regret: -85.61972024179818

For Home C, dynamic Regret: -147.6560344535884

For Home F, dynamic Regret: -681.0013253883544

- **Regrets for RHC for Linear Regression:**

```
For Home B, static Regret: -296.50691541373476
For Home C, static Regret: -360.2432397859767
For Home F, static Regret: -1871.4224291718174
For Home B, dynamic Regret: -114.1160711065437
For Home C, dynamic Regret: -132.05905310259539
For Home F, dynamic Regret: -817.2203756720055
```

- **Regrets for CHC for ARIMA:**

```
For Home B, static Regret: -268.0105645489806
For Home C, static Regret: -375.84022113696994
For Home F, static Regret: -1735.2033788881408
For Home B, dynamic Regret: -85.61972024178954
For Home C, dynamic Regret: -147.65603445358863
For Home F, dynamic Regret: -681.001325388329
```

- **Regrets for CHC for Linear Regression:**

```
For Home B, static Regret: -296.506915413731
For Home C, static Regret: -360.24323978594816
For Home F, static Regret: -1871.4224291717392
For Home B, dynamic Regret: -114.11607110653995
For Home C, dynamic Regret: -132.05905310256685
For Home F, dynamic Regret: -817.2203756719273
```

**Task 4: For the best combination of control algorithm and prediction algorithm, vary a and b to see the impacts:**

- Best combination of control algorithm and prediction algorithm is determined in this task and then vary the a and b values to calculate the cost function.

```
A = [4, 0.1, 1, 5, 10]
B = [4, 0.1, 1, 5, 10]

for ai in A:
    for bi in B:
        print("For a = ",ai,"and b = ",bi,":")
        [_, result_B] = control_algo_online_op(homeB_LR, p_t = 0.40, a = ai, b = bi, predictionHorizon = 5, commitmentHorizon=3) #Linear Regression predictions with CHC
        print("For Home B, Result of CHC with LR(w=5,v=3): ", result_B)
        [_, result_C] = control_algo_online_op(homeC, p_t = 0.40, a = ai, b = bi, predictionHorizon = 5, commitmentHorizon=3) #ARIMA predictions with CHC
        print("For Home C, Result of CHC with ARIMA(w=5,v=3): ", result_C)
        [_, result_F] = control_algo_online_op(homeF_LR, p_t = 0.40, a = ai, b = bi, predictionHorizon = 5, commitmentHorizon=3) #Linear Regression predictions with CHC
        print("For Home F, Result of CHC with LR(w=5,v=3): ", result_F)
```

- Best for home B:

**For a = 4 and b = 0.1:**

**For Home B, Result of CHC with Linear Regression (w=5, v=3): 420.8331773104108**

- Best for home C:

For  $a = 4$  and  $b = 0.1$ :

For Home C, Result of CHC with ARIMA( $w=5, v=3$ ): 609.9011831912904

- Best for home F:

For  $a = 4$  and  $b = 0.1$ :

For Home F, Result of CHC with LR( $w=5, v=3$ ): 2238.5675645420356

**Task 5: Try at least two algorithm selection (one deterministic, one randomized) to see if their performance**

- **Deterministic Algorithm Selection: Weighted Majority with 4 tests**

- We have initially assigned 0.33 weight to all the three algorithms.
- We have taken four kind of test in that on different iterations.
- On every iteration we have increased the weight of algorithm with minimum cost function by 0.10 and decreased the weight of maximum cost with 0.10.
- Test 1 : Home B with static regret  
Test 2: Home B with dynamic regret  
Test 3: Inverted Home B with static regret  
Test 4: Inverted Home B with dynamic regret
- After 4 iterations we get the results as follows:

```
Algorithm Selection Using Weighted Majority:
0.63*OGD + 0.73*RHC + 0.33*CHC
We choose RHC as it has maximum weight: 0.73
```

---

- **Randomized Algorithm Selection: Most winning algorithm is chosen**

- Like Deterministic selection above, I have taken 4 tests but the test get chosen using random number generator.
- The Algorithm which has minimum regret won the round.
- We calculate the winnings of all the algorithms.

```
Algorithm Selection Using Randomized Maximum Winning
Winning of OGD 0
Winning of RHC 4
Winning of CHC 0
```

So, RHC wins this selection.

## Bonus Task 1: Online Balanced Descent Algorithm:

- OBD is one step ahead of OGD as its step sizes varies with time.

Reference: <https://arxiv.org/pdf/1803.10366.pdf>

We have already implemented OBD above.

```
result_OBD_B = OBD_online_op(homeB)
result_OBD_C = OBD_online_op(homeC)
result_OBD_F = OBD_online_op(homeF)
```

OGD Optimization Problem solution is: 1892.720581127459

OGD Optimization Problem solution is: 2143.127367233312

OGD Optimization Problem solution is: 7521.536038487382

- 
- Regrets of OBD:

### Regrets of OBD:

```
193] static_regret_OGD_B = result_OBD_B - offline_static_result_B
static_regret_OGD_C = result_OBD_C - offline_static_result_C
static_regret_OGD_F = result_OBD_F - offline_static_result_F
print("For Home B, static Regret: ", static_regret_OGD_B)
print("For Home C, static Regret: ", static_regret_OGD_C)
print("For Home F, static Regret: ", static_regret_OGD_F)
dynamic_regret_OGD_B = result_OBD_B - offline_dynamic_result_B
dynamic_regret_OGD_C = result_OBD_C - offline_dynamic_result_C
dynamic_regret_OGD_F = result_OBD_F - offline_dynamic_result_F
print("For Home B, dynamic Regret: ", dynamic_regret_OGD_B)
print("For Home C, dynamic Regret: ", dynamic_regret_OGD_C)
print("For Home F, dynamic Regret: ", dynamic_regret_OGD_F)
```

```
↳ For Home B, static Regret: 874.3192707651722
For Home C, static Regret: 777.4568672533276
For Home F, static Regret: 1235.1928123868756
For Home B, dynamic Regret: 1056.7101150723634
For Home C, dynamic Regret: 1005.6410539367089
For Home F, dynamic Regret: 2289.3948658866875
```



## Bonus Task2: Another Algorithm selection Method

- We have created a randomize selection algorithm for Algorithm selection which runs for 1000 trials.

### Another Algorithm selection Method

Based on Regret Factor as scale, we can generate new algorithm Selector. In this **Randomized method**, we allocate few keys to all the three algorithms and then we generate a random number if that matches with the key of the algorithm then we going to increase the wins for that algorithm.

```
[194] #OGD = 0, #OBD = 1 #RHC = 2 #CHC = 3  
      rand_dict = {'OGD': 0, 'OBD': 0, 'RHC':0,'CHC' : 0 }
```

```
▶ i = 1  
  while(i<1000):  
    random_number = np.random.randint(4)  
    if random_number == 0:  
      rand_dict['OGD'] += 1  
    elif random_number == 1:  
      rand_dict['OBD'] += 1  
    elif random_number == 2:  
      rand_dict['RHC'] += 1  
    else:  
      rand_dict['CHC'] += 1  
    i += 1  
  print(rand_dict)
```

```
📄 {'OGD': 287, 'OBD': 296, 'RHC': 330, 'CHC': 288}
```

So, after 1000 trials, RHC got maximum wins. Thus, it will be chosen randomly.