

# IE 613: Assignment 2

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Question 1:

To Show: Follow-the-Regularized-Leader (FoReL) with linear loss functions and negative entropy as the regularizer is equivalent to Weighted Majority algorithm.

$$\begin{aligned} w_t &= \underset{w \in K}{\operatorname{argmin}} \sum_{s=1}^{t-1} \langle w, v_s \rangle + \frac{1}{\eta} \langle w, \log w \rangle \\ &= \underset{w \in K}{\operatorname{argmin}} \sum_{s=1}^{t-1} \sum_{j=1}^d w_j v_{s,j} + \frac{1}{\eta} \sum_{j=1}^d w_j \log w_j \\ &= \varphi(w) \text{ [Say]} \end{aligned}$$

Now,

$$\begin{aligned} \frac{\partial \varphi(w)}{\partial w_j} &= 0 \\ \Rightarrow \sum_{s=1}^t v_{s,j} + \frac{1}{\eta} \left[ \log w_j + w_j \frac{1}{w_j} \right] &= 0 \\ \Rightarrow \log w_j &= -\eta \sum_{s=1}^t v_{s,j} - 1 \\ \Rightarrow w_j &= \exp^{-\eta \left( \sum_{s=1}^t v_{s,j} - 1 \right)} \end{aligned}$$

which is the  $w_j$  that we use to update in the Weighted Majority Algorithm.

Question 2:

**File: ex2a.py**

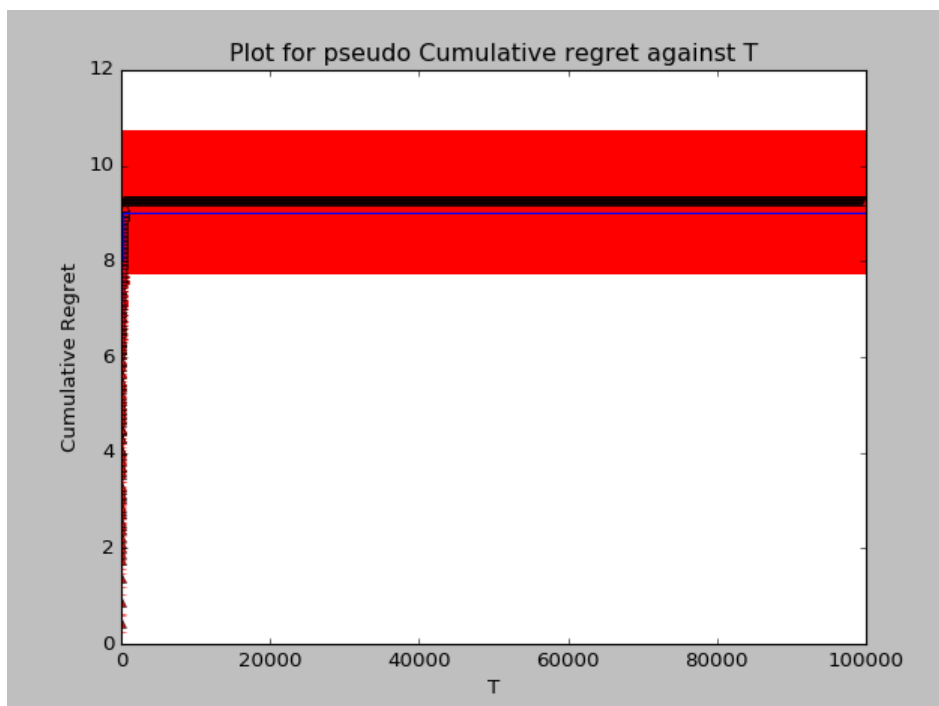


Figure 1: Plot for (pseudo) cumulative regret vs T for FTL

Question 3:

**File: ex3.py**

Since we proved the equivalence of WM algorithm and FoReL, we used it to get the following output.

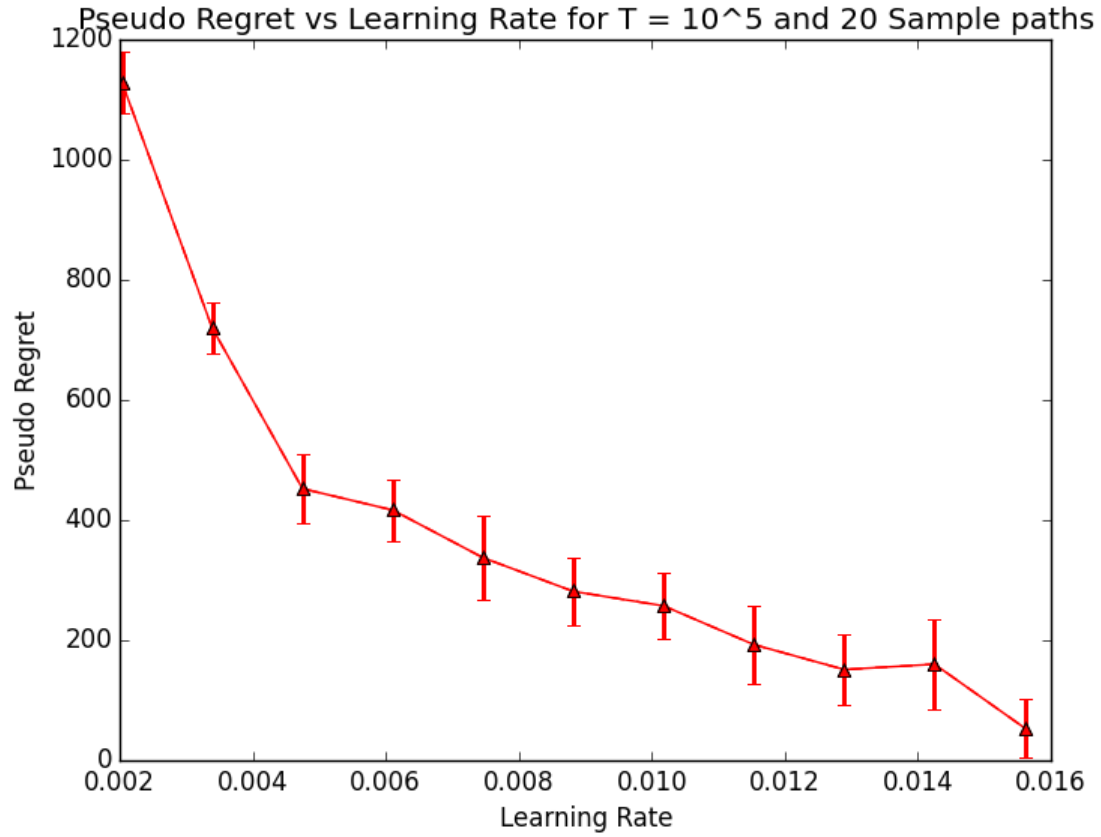


Figure 2: Pseudo Regret of FoReL for  $T = 10^5$  and 20 Sample paths

Question 4:

File: ex4a.py,ex4b.py,ex4c.py

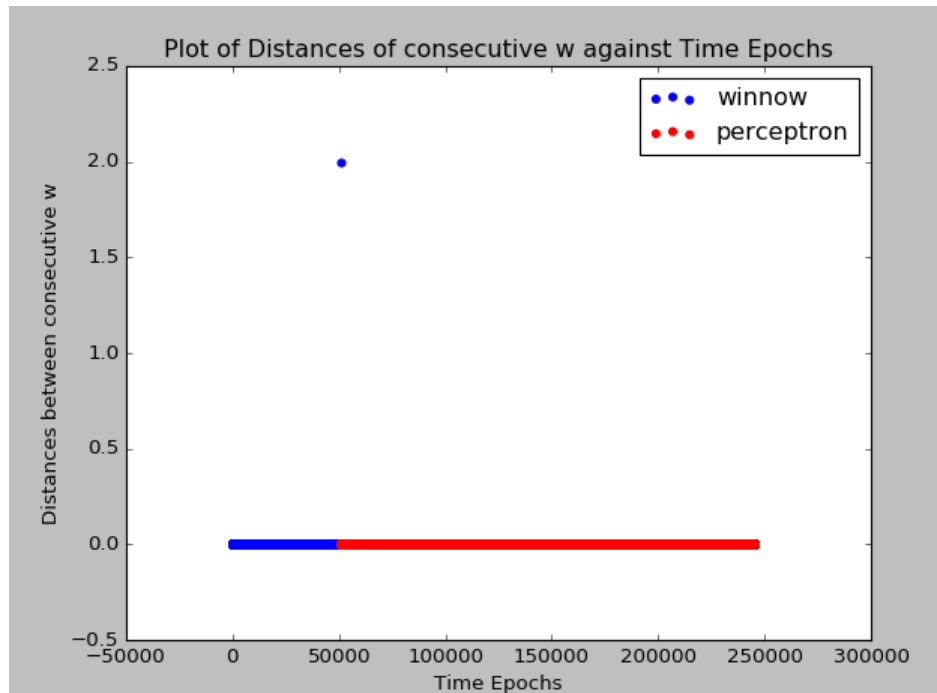


Figure 3: Plot of Distances of consecutive w against Time Epochs for Skin Segmentation Dataset

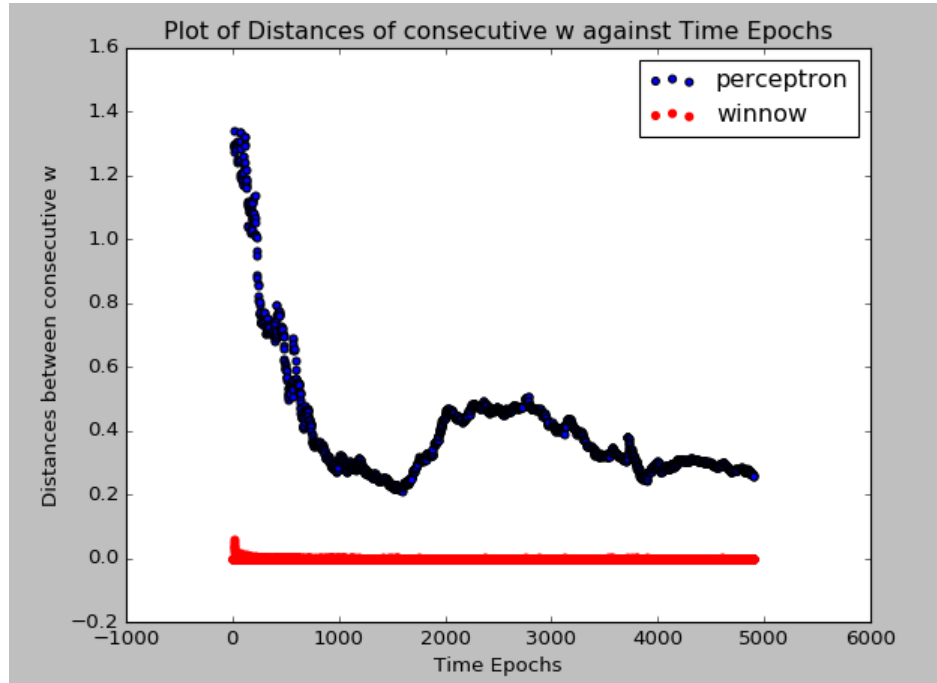


Figure 4: Plot of Distances of consecutive  $w$  against Time Epochs for Online News Popularity Dataset

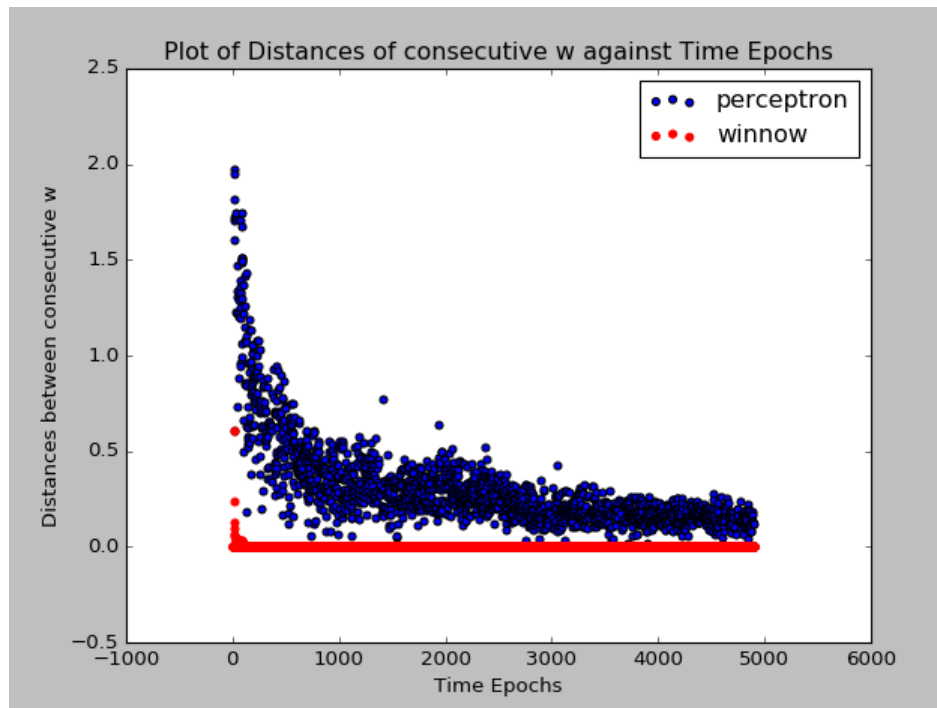


Figure 5: Plot of Distances of consecutive  $w$  against Time Epochs for Wine Quality Dataset

### Question 6:

File: ex6a.py

```
>>> runfile('C:/Users/Aak/Desktop/Sem IV/IE 613/Assignment 2/ex6a.py', wdir='C:/Users/Aak/Desktop/Sem IV/IE 613/Assignment 2')
No. of mistakes for Perceptron is: 5
Estimate of gamma is: 0.04151814921
Mistake Bound for perceptron is: 416.694159155
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.05
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.1
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.15
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.2
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.25
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.3
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.35
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.4
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.45
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.5
>>>
```

Figure 6: Output for question 6

No. of mistakes for Perceptron is: 5  
Estimate of gamma is: 0.04151814921  
Mistake Bound for perceptron is: 416.694159155  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.05  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.1  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.15  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.2  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.25  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.3  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.35  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.4  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.45  
No. of Mistakes for Winnow is: 1004 when value of eta is: 0.5