

## Homework 3: Logistic Regression with Regularization

Implement logistic regression as defined in Lecture 10 with feature transformation and regularization:

- Transform the feature space to higher dimension with various  $\Phi_n(x)$ :  $n = 2, 3$ . You may call the built-in function in sklearn for polynomial feature transform. (from `sklearn.preprocessing import PolynomialFeatures`)
- Apply regularization on the problem with input space of  $\Phi_n(x)$ .
- Train and validate your implementation with the same breast cancer dataset for Homework 2: 2 classes, 30 features, 569 data points
- 5-fold cross validation.

### To Submit:

Code your work with Python 3. You are supposed to submit both the well-documented .py python files (20pt) and the report. In the report, the following sections are required:

1. **Solution:** (10 pts) Clearly state your algorithm for:
  - Logistic regression with regularization.
  - Computation of the error measure with cross validation.
2. **5-fold Cross Validation:** (20 pts)
  - **Experiment:** Description of the setup of the experiments and computation of  $E_{val}$ .
  - **Result:** Show the performance plots for various  $\lambda$  (regularization) for  $E_{val}$  for various  $\Phi_n(x)$ ,  $n = 1, 2, 3$  (similar to slide 17 in lecture 14)
  - **Discussion:** Discuss the result. What did you observe? Does feature transformation to higher order help? What is the effect of  $\lambda$ ? What is the best choice of  $\lambda$ ? What did you learn from this experiment?

Put all files together and submit a zipped file. Include a readme, explaining which problem(s) you have finished. So I know how to grade. Content in the readme file:

1. What did you finish?
2. What python version (2.7? 3.6?)
3. What platform did you use (linux? Mac? windows?)
4. Resources that helped me.