## Problem 1

*Probability.* If you throw two fair, six-sided dice, what is the probability that there will be at least one 5?

## Problem 2

Bayes Rule. There is a 0.1% chance that I have a certain disease. The test for this disease is 90% accurate for positive test results (i.e., p(test positive | have disease) = 0.9) and 80% accurate for negative test results (i.e., p(test negative | don't have disease) = 0.8). What is the probability that I have the disease given that I have tested positive?

## Problem 3

Uniform Distribution. Let X be an independent and identically distributed (i.i.d.) collection of random variables from a Uniform distribution with parameters a and b,  $X \sim \text{Uniform}(x \mid a, b)$ , where a = 0 and  $b = \frac{1}{2}$ .

- (a) What out the probability density function (pdf) of X?
- (b) What is the  $p(X = 0.00027 \mid a, b)$  ?
- (c) What is the  $Pr(X = 0.00027 \mid a, b)$  (the probability that X = 0.00027)?

#### Problem 4

Poisson Distribution. Let X be an i.i.d. collection of random variables form a Poisson distribution with parameter  $\lambda$ ,  $X \sim \text{Poi}(\lambda)$ , where  $\lambda > 0$ .

- (a) Write out the exponential family form of X.
- (b) Determine the sufficient statistic of X for the Poisson distribution (T(x)).
- (c) Write out the log-partition function of  $X(A(\eta))$ .
- (d) Determine the response function (*hint:* find the inverse of the link function).
- (e) Determine  $\mathbb{E}(X)$  and  $\operatorname{var}(X)$  (you can either derive from the Poisson distribution or use the log-partition function).
- (f) Write out the maximum likelihood estimate (MLE) of  $\lambda$  for data  $X = \{x_1, ..., x_n\}$ .
- (g) Now, let  $\lambda \sim \text{Ga}(\alpha, \beta)$ , where  $\alpha, \beta > 0$ . The gamma distribution is conjugate to the Poisson distribution,  $\text{Ga}(x|\alpha, \beta) \propto x^{\alpha-1}/e^{\beta x}$ . Write out the MAP of  $\lambda$  given  $X = \{x_1, ..., x_n\}$ .

#### Problem 5

MAP and MLE simulation. For this problem, please download the data labelled HW1.txt. This file contains a column array of 10,000 integer values, where each value represents the number of customers that entered a 24 hour laundromat in one hour time intervals, over 10,000 hours. Let  $X = \{x_1, ..., x_t, ..., x_n\}$  represent this column array, where  $x_t$  is the number of customers that entered during the  $t^{th}$  hour. The hourly arrival of customers can be modeled as a collection of i.i.d. random variables drawn from a Poisson distribution,  $X \sim \text{Poi}(\lambda)$ , where  $\lambda$  is the hourly arrival rate.

- (a) Plot a histogram of X using 25 bins.
- (b) Using your answer from Problem 4(f), compute the MLE of  $\lambda$  for the observed data X.

For parts c - e, model the *hourly arrival rate*  $\lambda$  as having a Gamma distribution,  $\lambda \sim \text{Ga}(\alpha, \beta)$ , where  $\alpha, \beta > 0$ . Use your answer from Problem 4(g) to:

- (c) Compute the MAP of  $\lambda$  given X for  $\alpha = 1$  and  $\beta = 1$ .
- (d) Compute the MAP of  $\lambda$  given X for  $\alpha = 100$  and  $\beta = 1$ .
- (e) Compute the MAP of  $\lambda$  given X for  $\alpha = 10$  and  $\beta = 1$ .
- (f) Which approximation of  $\lambda$  in parts b e do you think is the best? How much does the prior distribution, and parameterizations of the prior in particular, impact the MAP estimates of  $\lambda$ ? (one or two sentences)

#### Problem 6

*MAP and MLE regression.* For this problem, please download the data sets: test.txt, train.txt, samples.txt. This data consists of training data and independent data of gene expression for two types of leukemia (AML and ALL) the samples.txt describes the class labels.

- (a) Build a standard linear model on the training data and make predictions on the test data.
- (b) Build a linear shrinkage model on the training data and make predictions on the test data. Report results for various values of  $\lambda$  the regularization parameter.
- (c) Use cross-validation to set the regularization parameter.

# Problem 7

 $MAP\ logistic\ regression.$  Derive a MAP estimator for a Logistic regression model using the standard shrinkage idea.