Name:

1. (7 points) A die is cast independently a number of times until a six appears on the up side. The probability such event happens after n casts is:

$$p(n) = (1-q)^{n-1}q$$

where q is un unknown probability of six showing on the up side.

- (a) Explain the form (geometric distribution) of p(n). (1 point)
- (b) Derive the maximum likelihood estimate for q. (3 points)
- (c) Let's assume that an upper limit on the number of casts n_{max} has been set. What conclusion can be made about q if n_{max} is reached, i.e. if $n > n_{max}$? (1 point).
- (d) What is the maximum likelihood estimate of q if you were just told that in n_{max} casts six never showed on the up side? (1 point)
- (e) Let's assume that it is know a priori that $q \in [0.1, 0.2]$. What would be the maximum likelihood estimate in this case? Consider e.g. the case when six shows up in the first cast. (1 point)
- 2. (5 points) Define the minimax decision problem (2 points). Find the optimal minimax strategy in the following two-class problem: the observations are real numbers x ∈ [0, 1]. Conditional probability densities p(x|k) are p(x|1) = x/2 and p(x|2) = 6x(1-x). (3 points)
- 3. (7 points). K-means. Describe the basic algorithm (2 points).

Consider the following problem. K wells can be drilled at any location (X_k, Y_k) at fixed cost C_w per well to provide water to n huts in positions (x_i, y_i) . A pipe can be laid in a straight line between a well and a hut at a cost C_p per meter.

- (a) Define the cost of the installation in terms of the number of wells, K, well locations, $\{(X_k, Y_k)\}_{k=1}^K$, and hut positions, $\{(x_i, y_i)\}_{i=1}^n$. (1 point)
- (b) Modify the K-means algorithm so that it minimizes the cost of drilling wells and laying the pipes, i.e. so that it finds a local minimum of the total cost of installation. (2 points)
- (c) Explain how to set K. (1 point)
- (d) Consider the case when pipes can be laid only in north-south and east-west directions along existing roads. Modify the K-means accordingly. (1 point)
- 4. (6 points). Describe the Support Vector Machine, its learning method and its properties (3 points), including the way it handles non-separable training data (1 point). Discuss the use of kernels and the differences between linear and kernel SVM in learning and in classification (2 points).

RPZ