## Foundations of Data Science and Machine Learning – *Homework 3*Isaac Martin

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EXERCISE 2. Let  $\mathbf{X} \in \mathbb{R}^{n \times d}$  be a matrix whose n rows are the data points  $\mathbf{x}_1, ..., \mathbf{x}_n \in \mathbb{R}^d$ , and let  $\chi = \{\mathbf{x}_1, ..., \mathbf{x}_n\}$ . Consider the k-means optimization problem: find a partition  $C_1, ..., C_k$  which minimizes, among all partitions of [n] into k subsets,

Exercise 3. Find the mapping  $\varphi(\mathbf{x})$  that gives rise to the polynomial kernel

$$K(\mathbf{x}, \mathbf{y}) = (x_1 x_2 + y_1 y_2)^2.$$

*Proof:* Consider the map  $\varphi:\mathbb{R}^2\to\mathbb{R}^3$  defined  $\varphi(x_1,x_2)=(x_1^2,x_2^2,\sqrt{2}x_1x_2)$ . Interestingly, this is similar to the map one considers from a polynomial ring  $R[x_1,x_2]$  to its  $2^{\mathrm{nd}}$  Veronese subring  $R[x_1^2,x_1x_2,x_2^2]$ . We then have that

$$\varphi(\mathbf{x}) \cdot \varphi(\mathbf{y})^T = (x_1^2, x_2^2, \sqrt{2}x_1x_2) \cdot (y_1^2, y_2^2, \sqrt{2}y_1y_2)$$
$$= x_1^2 y_1^2 + x_2^2 y_2^2 + 2x_1 y_1 x_2 y_2$$
$$= (x_1 y_1 + x_2 y_2)^2,$$

hence  $\varphi$  gives rise to the desired kernel.