Supervised
input \((x, y)\) \rightarrow \text{labels}

Supervised

Unsupervised
input \(X\) \rightarrow \text{Clustering}

Output
real values
\(\mathbb{R}\)

Output set

Regression

Dimensionality Reduction

Classification

Cross-validation

Cross-validation
1. Consider the random variables $X$ and $Y$ described by the joint probability table.

<table>
<thead>
<tr>
<th></th>
<th>$X = 1$</th>
<th>$X = 2$</th>
<th>$X = 3$</th>
</tr>
</thead>
<tbody>
<tr>
<td>$Y = 1$</td>
<td>0.10</td>
<td>0.05</td>
<td>0.10</td>
</tr>
<tr>
<td>$Y = 2$</td>
<td>0.30</td>
<td>0.25</td>
<td>0.20</td>
</tr>
</tbody>
</table>

Derive the following values

(a) $\Pr(X = 1) = 0.1 + 0.3 = 0.4$
(b) $\Pr(X = 2 \cap Y = 1) = 0.05$
(c) $\Pr(X = 3 \mid Y = 2) = \frac{0.2}{0.3 + 0.25 + 0.2} = \frac{0.2}{0.75}$

Compute the following probability distributions.

(d) What is the marginal distribution for $X$?
(e) What is the conditional probability for $Y$, given that $X = 2$?

Answer the following question about the joint distribution.

(f) Are random variables $X$ and $Y$ independent?
(g) Is $\Pr(X = 1)$ independent of $\Pr(Y = 1)$?

\[ P_{X,Y} \begin{pmatrix} 0.1 & 0.25 \\ 0.25 & 0.3 \end{pmatrix} \]

\[ P_{X=2,Y=1} = \frac{0.25}{0.25 + 0.25} = \frac{0.25}{0.5} = 0.5 \]

\[ \Pr(X = 1, Y = 1) = \frac{0.10}{0.10 + 0.25 + 0.10} = \frac{0.10}{0.45} = \frac{2}{9} \]

\[ \Pr(X = 2, Y = 1) = \frac{0.05}{0.25} = 0.2 \]

\[ \Pr(X = 3, Y = 1) = \frac{0.10}{0.3} = \frac{1}{3} \]

\[ \Pr(X = 1, Y = 1) \times \Pr(Y = 1) = \frac{2}{9} \times 0.5 = \frac{1}{9} \]

\[ \Pr(X = 1, Y = 1) \times \Pr(Y = 1) = \frac{1}{9} \neq \frac{2}{9} \]

**No.**
2. Consider two models $M_1$ and $M_2$, where from prior knowledge we believe that $\Pr(M_1) = 0.25$ and $\Pr(M_2) = 0.75$. We then observe a data set $D$. Given each model we assess the likelihood of seeing that data given the model as $\Pr(D \mid M_1) = 0.5$ and $\Pr(D \mid M_2) = 0.01$. Now that we have the data, which model is has a higher probability of being correct?
3. Assume I observe 3 data points $x_1$, $x_2$, and $x_3$ drawn iid from an unknown distribution. Given a model $M$, I can calculate the likelihood this each data point as $\Pr(x_1 \mid M) = 0.5$, $\Pr(x_2 \mid M) = 0.1$, and $\Pr(x_3 \mid M) = 0.2$. What is the likelihood of seeing all of these data points, given the model $M$: $\Pr(x_1, x_2, x_3 \mid M)$?

$$\Pr(x_1, x_2, x_3 \mid M) = \Pr(x_1 \mid M) \cdot \Pr(x_2 \mid M) \cdot \Pr(x_3 \mid M) = 0.5 \cdot 0.1 \cdot 0.2 = 0.0$$
4. Consider a pdf $f$ so that a random variable $X \sim f$ has expected value $E[X] = 3$ and variance $\text{Var}[X] = 10$. Now consider $n = 10$ iid random variables $X_1, X_2, \ldots, X_{10}$ drawn from $f$. Let $\bar{X} = \frac{1}{10} \sum_{i=1}^{10} X_i$.

(a) What is $E[\bar{X}]$?
(b) What is $\text{Var}[\bar{X}]$?
(c) What is the standard deviation of $\bar{X}$?
(d) Which is larger $\Pr[X > 4]$ or $\Pr[\bar{X} > 4]$?
(e) Which is larger $\Pr[X > 2]$ or $\Pr[\bar{X} > 2]$?

\[ \text{Central Limit Theorem} \]
\[ E[\bar{X}] = E[X] \]
\[ \text{Var}(\bar{X}) = \frac{\text{Var}(X)}{n} \]
\[ \lim_{n \to \infty} \text{Normal} \]
5. Let \( X \) be a random variable that you know is in the range \([-1, 2]\) and you know has expected value of \( \text{E}[X] = 0 \). Use the Markov Inequality to upper bound \( \text{Pr}[X > 1.5] \)?

(Hint: you will need to use a change of variables.)

\[
\begin{align*}
Z &= X + 1 \\
E[Z] &= E[X + 1] = 0 + 1 = 1 \\
E[Z^2] &= 2 \\
\text{Pr}[X > 1.5] &= \text{Pr}[Z > 2.5] \\
&= \frac{\text{Pr}[Z > 3.5]}{2.5} \\
&\leq \frac{2}{3.5} = \frac{4}{7} = 0.57
\end{align*}
\]
6. Consider a matrix

\[ A = \begin{bmatrix} 2 & 2 & 3 \\ -2 & 7 & 4 \\ -3 & -3 & -4 \\ -8 & 2 & 3 \end{bmatrix} \]

(a) Add a column to \( A \) so that it is invertable.
(b) Remove a row from \( A \) so that it is invertable.
(c) Is \( AA^T \) invertable? **No**
(d) Is \( A^T A \) invertable? **Yes**

\[ A = U S V^T \]
\[ A^T A = U S^2 U^T \]

\[ A \in \mathbb{R}^{4 \times 3} \]
\[ A A^T = (4 \times 3)(3 \times 4) \]
\[ \text{columns} \quad \text{lin. indp.} \]
\[ A^T A \in \mathbb{R}^{3 \times 3} \]
\[ \text{full rank} \]
7. Consider two vectors \( u = (0.5, 0.4, 0.4, 0.5, 0.1, 0.4, 0.1) \) and \( v = (-1, -2, 1, -2, 3, 1, -5) \).

(a) Check if \( u \) or \( v \) is a unit vector.

(b) Calculate the dot product \( \langle u, v \rangle \).

(c) Are \( u \) and \( v \) orthogonal?

\[
\sum_{i=1}^{n} u_i v_i = 0.5 \cdot 0.4 + 0.4 \cdot 0.4 + 0.5 \cdot 0.4 + 0.1 \cdot 0.1 + 0.4 \cdot 0.3 + 0.1 \cdot 1 = 1
\]

\[
\langle u, v \rangle = \sum_{j=1}^{7} u_j v_j = (0.5)(-1) + (0.4)(-2) + \ldots = 0
\]
8. Consider a matrix \( A \in \mathbb{R}^{n \times 4} \). Each row represents a customer (there are \( n \) customers in the database). The first column is the age of the customer in years, the second column is the number of days since the customer entered the database, the third column is the total cost of all purchases ever by the customer in dollars, and the last column is the total profit in dollars generated by the customer.

For each of the following operations, decide if it is reasonable or unreasonable.

(a) Run simple linear regression using the first three columns to build a model to predict the fourth column. \( <x, y> \) \( \text{reasonable} \)

(b) Use k-means clustering to group the customers into 4 types using Euclidean distance between rows as the distance. \( \text{UnR} \)

(c) Use PCA to find the best 2-dimensional subspace, so we can draw the customers in a \( \mathbb{R}^2 \) in way that has the least projection error. \( \text{UnR} \text{ Reason} \)

(d) Use the linear classification to build a model based on the first three columns to predict if the customer will make a profit +1 or not −1.

or w/ Perception
not ok w/ Loss fxn

miles z
etcu 1000
pop 700
1742
9. Consider a data set \((X, y)\) where \(X \in \mathbb{R}^{n \times 3}\) we decompose into a test and a training data set \((X_{\text{train}}, y_{\text{train}})\). Assume that \(X_{\text{train}}\) is not just a subset of \(X\), but also pads prepends a columns of all 1s. We build a linear model

\[
\alpha = (X_{\text{train}}^T X_{\text{train}})^{-1} X_{\text{train}}^T y_{\text{train}}.
\]

where \(\alpha \in \mathbb{R}^4\). The remaining two testing data points are \((x_1, y_1)\) and \((x_2, y_2)\), where \(x_1, x_2 \in \mathbb{R}^3\). Explain (write a mathematical expression) to use this test data to estimate the generalization error. That is, if one new data point arrives \(x\), how much squared error would we expect the model \(\alpha\) to have compared to the unknown true value \(y\)?

\[
M_\alpha(x) = \langle \alpha, (1, x) \rangle
\]

\[
\left( M_\alpha(x_1) - y_1 \right)^2 + \left( M_\alpha(x_2) - y_2 \right)^2
\]

\[
\hat{\epsilon}
\]
10. Consider a function $f(x, y)$ with gradient \( \nabla f(x, y) = (x - 1, 2y + x) \). Starting at a value \((x = 1, y = 2)\), and a learning rate of $\gamma = 1$, execute one step of gradient descent.

\[
\nabla f(x, y) = (x - 1, 2y + x)
\]

\( (1, 2) - (1) \begin{pmatrix} x - 1, 2y + x \end{pmatrix} = (0, 5) \)

\[
(1, 2) - (0, 5) = (1, -3)
\]
11. Consider running gradient descent with a fixed learning rate $\gamma$. For each of the following, we plot the function value over 10 steps (the function is different each time). Decide whether the learning rate is probably too high, too low, or about right.

(a) $f_1: 100, 99, 98, 97, 96, 95, 94, 93, 92, 91$
(b) $f_2: 100, 50, 75, 60, 65, 45, 75, 110, 90, 85$
(c) $f_3: 100, 80, 65, 50, 40, 35, 31, 29, 28, 27.5, 27.3$
(d) $f_4: 100, 80, 60, 40, 20, 0, -20, -40, -60, -80, -100$

\[\text{too low, too high, just right, too low}\]
12. Consider a matrix $A \in \mathbb{R}^{8 \times 4}$ with squared singular values $\sigma_1^2 = 10$, $\sigma_2^2 = 5$, $\sigma_3^2 = 2$, and $\sigma_4^2 = 1$.

(a) What is the rank of $A$?

(b) What is $\|A - A_2\|_F^2$, where $A_2$ is the best rank-2 approximation of $A$.

(c) What is $\|A - A_2\|_F^2$, where $A_2$ is the best rank-2 approximation of $A$.

(d) What is $\|A\|_2^2$?

(e) What is $\|A\|_F^2$?

Let $v_1, v_2, v_3, v_4$ be the right singular vectors of $A$.

(f) What is $\|Av_2\|_2^2$?

(g) What is $(v_1, v_3)$?

(h) What is $\|v_4\|$?

Let $a_1 \in \mathbb{R}^4$ be the first row of $A$.

(i) Write $a_1$ in the basis defined by the right singular vectors of $A$. 

$$a_1 = \sqrt{5}v_1$$
13. Draw the Voronoi diagram of the following set of points.
14. What should you do, if running Lloyd’s algorithm for $k$-means clustering ($k = 2$), and you reach this scenario, where the algorithm terminates? (The black circles are data points and red stars are the centers).
15. Consider the following “loss” function. \( \ell_i(z_i) = \frac{(1 - z_i)^2}{2} \), where for a data point \((x_i, y_i)\) and prediction function \(f\), then \(z_i = y_i \cdot f(x_i)\). Predict how this might work within a gradient descent algorithm for classification.
16. Consider a set of 1-dimensional data points

\[(x_1 = 0, y_1 = +1) \ (x_2 = 1, y_1 = -1) \ (x_3 = 2, y_1 = +1) \ (x_4 = 4, y_1 = +1)\]

\[(x_5 = 6, y_1 = -1) \ (x_6 = 7, y_1 = -1) \ (x_7 = 8, y_1 = +1) \ (x_8 = 9, y_1 = -1)\]

Predict -1 or +1 using a kNN (k-nearest neighbor) classifier with \(k = 3\) on the following queries.

(a) \(x = 3\)
(b) \(x = 9\)
(c) \(x = -1\)