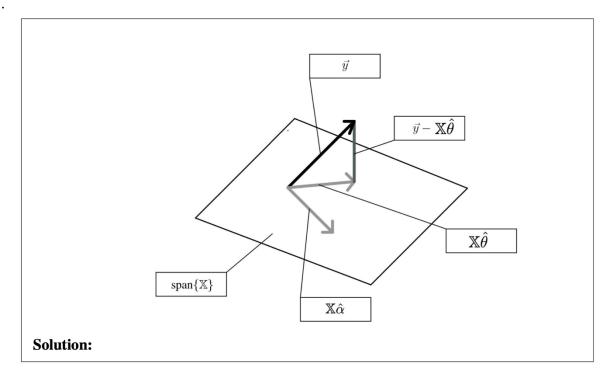
1.



- (a) What is always true about the residuals in least squares regression? Select all that apply.
 - \square A. They are orthogonal to the column space of the design matrix.
 - \square B. They represent the errors of the predictions.
 - \square C. Their sum is equal to the mean squared error.
 - \square D. Their sum is equal to zero.
 - \square E. None of the above.

Solution: (A), (B)

- (C): (C) is wrong because the mean squared error is the *mean* of the sum of the *squares* of the residuals.
- (D): A counter-example is: $\mathbb{X} = \begin{bmatrix} 2 & 3 \\ 1 & 5 \\ 2 & 4 \end{bmatrix}$, $\mathbb{Y} = \begin{bmatrix} 1 \\ 3 \\ 2 \end{bmatrix}$. After solving the least squares

problem, the sum of the residuals is -0.0247, which is not equal to zero. However, note that this statement is in general true if every feature contains the same constant intercept term.

(E): is wrong since (A) and (B) are correct.

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- (c) We fit a simple linear regression to our data (x_i, y_i) , i = 1, 2, 3, where x_i is the independent variable and y_i is the dependent variable. Our regression line is of the form $\hat{y} = \hat{\theta}_0 + \hat{\theta}_1 x$. Suppose we plot the relationship between the residuals of the model and the $\hat{y}s$, and find that there is a curve. What does this tell us about our model?
 - \square A. The relationship between our dependent and independent variables is well represented by a line.
 - ☐ B. The accuracy of the regression line varies with the size of the dependent variable.
 - ☐ C. The variables need to be transformed, or additional independent variables are needed.

Solution:

If we see a curve in our residual plot, then the relationship is not well represented by a line. Either more independent variables are needed, or transformations of the current variables are necessary.

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