

CHE384, From Data to Decisions: Measurement, Uncertainty, Analysis, and Modeling

Lecture 35

The Wrong Model, part 2

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Coefficient of Determination, R^2

- The Coefficient of Determination (R^2) is a measure of how much of the variation in Y is explained by the model

Regression Sum of Squares: $SSR = \sum (\hat{y}_i - \bar{y})^2$

Error Sum of Squares: $SSE = \sum (y_i - \hat{y}_i)^2$

Total Sum of Squares: $SSTO = \sum (y_i - \bar{y})^2$

$$R^2 = \frac{SSR}{SSTO} = 1 - \frac{SSE}{SSTO} \quad SSTO = SSR + SSE \quad (\text{only true for linear regression})$$

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Goodness of Fit

- Goodness of fit metric: $R^2 = r^2$

$$r = \frac{cov(X, Y)}{\sqrt{var(X)var(Y)}} \quad R^2 = \frac{cov^2(X, Y)}{var(X)var(Y)}$$

- Also, we can show that

$$R^2 = 1 - \frac{var(\epsilon)}{var(Y)} \quad R^2 \text{ is the fraction of the variance of } Y \text{ that is explained by the linear fit}$$

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Overall F-Test for Regression

- An overall test for model significance:
 - $H_0: \beta_1 = \beta_2 = \dots = \beta_{p-1} = 0$ (not testing intercept)
 - $H_A: \beta_j \neq 0$, for at least one value of j
 - p = number of parameters in the model
- If H_0 is true, then the model is not useful for explaining the variation in y
 - SSR is much smaller than SSE
 - SSE is about the same as $SSTO$

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Overall F-Test for Regression

- Is $SSTO$ about the same as SSE ?
- Compare $SSTO - SSE$ to SSE , considering the degrees of freedom

$$SSTO = SSR + SSE \quad (\text{only true for linear regression})$$

$$F = \frac{(SSTO - SSE) / (DF_{SSTO} - D_{SSE})}{SSE / DF_{SSE}} = \frac{SSR / (p - 1)}{SSE / (n - p)}$$

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Build an ANOVA Table

Source	SS	df	MS	F	p-value
Regression	SSR	p-1	SSR/(p-1)	MSR/MSE	from F-distribution
Error	SSE	n-p	SSE/(n-p)		
Total	SSTO	n-1			

$$F = \frac{SSR / (p - 1)}{SSE / (n - p)} = \frac{MSR}{MSE}$$

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For Weighted Regression

- We must include the weights in our sum of squares calculations

Regression Sum of Squares: $SSR = \sum w_i (\hat{y}_i - \bar{y}_w)^2$

Error Sum of Squares: $SSE = \sum w_i (y_i - \hat{y}_i)^2$

Total Sum of Squares: $SSTO = \sum w_i (y_i - \bar{y}_w)^2$

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Overall F-Test for Regression

- Note that the coefficient of determination (R^2) is related to this F-statistic:

$$R^2 = \frac{\sum_{i=1}^n (\hat{y}_i - \bar{y})^2}{\sum_{i=1}^n (y_i - \bar{y})^2} = 1 - \frac{\sum_{i=1}^n (y_i - \hat{y}_i)^2}{\sum_{i=1}^n (y_i - \bar{y})^2}$$

$$F = \frac{R^2 / (p - 1)}{(1 - R^2) / (n - p)}$$

F distribution with 4 and 50 d.f.

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Overall F-Test for Regression

- The F-statistic is F-distributed with $(p-1, n-p)$ degrees of freedom ($F_{p-1, n-p}$)
 - An F-distribution is the ratio of two independent χ^2 distributions
 - Assumes normal distribution of iid residuals with constant σ_e
 - F = explained variance/unexplained variance
- Calculate the p-value from the F-distribution and compare it with your significance level α
 - P-value = probability that the model explains the variance in y no better than by chance

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Overall F-Test for Regression

- For a two-parameter model, the F-test is the same as asking if the confidence interval of the slope includes zero (t-test with $DF = n-2$)
 - $F = [b_1 / SE(b_1)]^2$ for this case
- Don't over-interpret the test results
 - A small p-value doesn't necessarily imply a good fit of model to data, only that at least one model parameter is non-zero
 - A large p-value doesn't necessarily mean that the response variable is not dependent on the predictor variables, only that *this* model is not significant

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Training vs. Predicting

- We build a model by fitting it to data
 - We "train" or calibrate a model using a given data set
 - The residual standard deviation is a measure of how well the model fits this data set
 - If we add more fitting parameters to the model, we always get a better fit
- The real test, however, is how well we match a *new* data set, one not used in the training
 - Called **validation** of the model
 - The residual standard deviation for validation data will almost always be higher than for the training data

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Overfitting

- Adding new model terms always makes the fit better, but can result in fitting noise

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Model Scope

- Every model is built and validated over a range of input conditions, called the **scope** of the model
 - When a model is developed, its scope should be clearly specified
 - Prediction and interpretation (the two goals of modeling) should generally be limited to within the scope
 - Extrapolations are sometimes done, but know that uncertainty estimates are no longer valid

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Lecture 35: What have we learned?

- What is the coefficient of determination and how is it calculated?
- What is the overall regression F-test and how is it used?
- What are the assumptions inherent in the F-test?
- What is model validation?
- Define model scope

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