MITx: 15.071x The Analytics Edge

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READING TEST SCORES

The Programme for International Student Assessment (PISA) is a test given every three years to 15-year-old students from around the world to evaluate their performance in mathematics, reading, and science. This test provides a quantitative way to compare the performance of students from different parts of the world. In this homework assignment, we will predict the reading scores of students from the United States of America on the 2009 PISA exam.

The datasets pisa2009train.csv (/c4x/MITx/15.071x/asset/pisa2009train.csv) and pisa2009test.csv (/c4x/MITx/15.071x/asset/pisa2009test.csv) contain information about the demographics and schools for American students taking the exam, derived from 2009 PISA Public-Use Data Files (http://nces.ed.gov/pubsearch/pubsinfo.asp?pubid=2011038) distributed by the United States National Center for Education Statistics (NCES). While the datasets are not supposed to contain identifying information about students taking the test, by using the data you are bound by the NCES data use agreement (/c4x/MITx/15.071x/asset/NCES_Data_Use_Agreement.txt), which prohibits any attempt to determine the identity of any student in the datasets.

Each row in the datasets pisa2009train.csv and pisa2009test.csv represents one student taking the exam. The datasets have the following variables:

grade: The grade in school of the student (most 15-year-olds in America are in 10th grade)

male: Whether the student is male (1/0)

raceeth: The race/ethnicity composite of the student

preschool: Whether the student attended preschool (1/0)

expectBachelors: Whether the student expects to obtain a bachelor's degree (1/0)

motherHS: Whether the student's mother ompleted high school (1/0)

motherBachelors: Whether the student's mother obtained a bachelor's degree (1/0)

motherWork: Whether the student's mother has part-time or full-time work (1/0)

fatherHS: Whether the student's father completed high school (1/0)

fatherBachelors: Whether the student's father obtained a bachelor's degree (1/0)

fatherWork: Whether the student's father has part-time or full-time work (1/0)

selfBornUS: Whether the student was born in the United States of America (1/0)

motherBornUS: Whether the student's mother was born in the United States of America (1/0)

fatherBornUS: Whether the student's father was born in the United States of America (1/0)

englishAtHome: Whether the student speaks English at home (1/0)

computerForSchoolwork: Whether the student has access to a computer for schoolwork (1/0)
read30MinsADay: Whether the student reads for pleasure for 30 minutes/day (1/0)
minutesPerWeekEnglish: The number of minutes per week the student spend in English class
studentsInEnglish: The number of students in this student's English class at school
schoolHasLibrary: Whether this student's school has a library (1/0)
publicSchool: Whether this student attends a public school (1/0)
urban: Whether this student's school is in an urban area (1/0)
schoolSize: The number of students in this student's school
readingScore: The student's reading score, on a 1000-point scale
PROBLEM 1.1 - DATASET SIZE (1 point possible)
Load the training and testing sets using the read.csv() function, and save them as variables with the names pisaTrain and pisaTest.
How many students are there in the training set?
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PROBLEM 1.2 - SUMMARIZING THE DATASET (2 points possible)
Using tapply() on pisaTrain, what is the average reading test score of males?
Of females?
Show Answer You have used 0 of 3 submissions

PROBLEM 1.3 - LOCATING MISSING VALUES (1 point possible)

Which variables are missing data in at least one observation in the training set?

\square male
□ raceeth
□ preschool
□ expectBachelors
□ motherHS
□ motherBachelors
□ motherWork
□ fatherHS
☐ fatherBachelors
☐ fatherWork
□ selfBornUS
□ motherBornUS
☐ fatherBornUS
\square englishAtHome
□ computerForSchoolwork
□ read30MinsADay
☐ minutesPerWeekEnglish
□ studentsInEnglish
□ schoolHasLibrary
□ publicSchool
□ urban
□ schoolSize
□ readingScore
□ readingscore
□ readingscore
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Show Answer	You have used 0 of 3 submissions
PROBLEM 2.	I - FACTOR VARIABLES (2 points possible)
homework prob	are variables that take on a discrete set of values, like the "Region" variable in the state dataset from the first lem this week. This is an unordered factor because there isn't any natural ordering between the levels. An ordered iral ordering between the levels (an example would be the classifications "large," "medium," and "small").
Which of the foll	owing variables is an unordered factor with at least 3 levels?
□ grade □ male □ racee	
Which of the foll	owing variables is an ordered factor with at least 3 levels?
□ grade □ male □ racee	
Show Answer	You have used 0 of 2 submissions
PROBLEM 2.2	2 - UNORDERED FACTORS IN REGRESSION MODELS (1 point possible)
each of the rema	dered factors in a linear regression model, we define one level as the "reference level" and add a binary variable for aining levels. In this way, a factor with n levels is replaced by n-1 binary variables. The reference level is typically be most frequently occurring level in the dataset.
then we would a	onsider the unordered factor variable "color", with levels "red", "green", and "blue". If "green" were the reference level, dd binary variables "colorred" and "colorblue" to a linear regression problem. All red examples would have colorred=1. All blue examples would have colorred=0 and
"More than one	ne variable "raceeth" in our problem, which has levels "American Indian/Alaska Native", "Asian", "Black", "Hispanic", race", "Native Hawaiian/Other Pacific Islander", and "White". Because it is the most common in our population, we will he reference level.
Which binary var	riables will be included in the regression model?
□ racee □ racee □ racee	
□ racee	thNative Hawaiian/Other Pacific Islander thWhite

PROBLEM 2	.3 - EXAMPLE UNORDERED FACTORS (2 points possible)
Consider again	adding our unordered factor race to the regression model with reference level "White".
For a student v	who is Asian, which binary variables would be set to 0? All remaining variables will be set to 1.
□ race	eethAmerican Indian/Alaska Native
□ race	eethAsian
□ race	eethBlack
□ race	ethHispanic
□ race	ethMore than one race
□ race	ethNative Hawaiian/Other Pacific Islander
For a student v	who is white, which binary variables would be set to 0? All remaining variables will be set to 1.
□ race	eethAmerican Indian/Alaska Native
□ race	eethAsian
□ race	eethBlack
□ race	ethHispanic
□ race	ethMore than one race
□ race	ethNative Hawaiian/Other Pacific Islander
Show Answer	You have used 0 of 2 submissions
PROBLEM 3	.1 - BUILDING A MODEL (1 point possible)
Because the ra see this when y Indian/Alaska N	ce variable takes on text values, it was loaded as a factor variable when we read in the dataset with read.csv() you can you run str(pisaTrain) or str(pisaTest). However, by default R selects the first level alphabetically ("American lative") as the reference level of our factor instead of the most common level ("White"). Set the reference level of the g the following two lines in your R console:
pisaTrain\$race	eth = relevel(pisaTrain\$raceeth, "White")

pisaTest\$raceeth = relevel(pisaTest\$raceeth, "White")

Now, build a linear regression model (call it lmScore) using the training set to predict readingScore using all the remaining variables.

It would be time-consuming to type all the variables, but R provides the shorthand notation "readingScore ~ ." to mean "predict readingScore using all the other variables in the data frame." The period is used to replace listing out all of the independent variables. As an example, if your dependent variable is called "Y", your independent variables are called "X1", "X2", and "X3", and your training data set is called "Train", instead of the regular notation:

LinReg = $Im(Y \sim X1 + X2 + X3, data = Train)$

You would use the following command to build your model:

 $LinReg = Im(Y \sim ., data = Train)$

What is the Multiple R-squared value of ImScore on the training set?



Note that this R-squared is lower than the ones for the models we saw in the lectures and recitation. This does not necessarily imply that the model is of poor quality. More often than not, it simply means that the prediction problem at hand (predicting a student's test score based on demographic and school-related variables) is more difficult than other prediction problems (like predicting a team's number of wins from their runs scored and allowed, or predicting the quality of wine from weather conditions).
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PROBLEM 3.2 - COMPUTING THE ROOT-MEAN SQUARED ERROR OF THE MODEL (1 point possible)
What is the training-set root-mean squared error (RMSE) of ImScore?
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PROBLEM 3.3 - COMPARING PREDICTIONS FOR SIMILAR STUDENTS (1 point possible)
Consider two students A and B. They have all variable values the same, except that student A is in grade 11 and student B is in grade 9. What is the predicted reading score of student A minus the predicted reading score of student B?
○ -59.09
O -29.54
○ 0 ○ 29.54
O 59.09
O The difference cannot be determined without more information about the two students
Show Answer You have used 0 of 2 submissions
PROBLEM 3.4 - INTERPRETING MODEL COEFFICIENTS (1 point possible)
What is the meaning of the coefficient associated with variable raceethAsian?
O Predicted average reading score of an Asian student
 Difference between the average reading score of an Asian student and the average reading score of a white student
Difference between the average reading score of an Asian student and the average reading score of all the students in the dataset
Predicted difference in the reading score between an Asian student and a white student who is otherwise
identical
Standard Vou being used 0 of 1 submissions
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Based on the signific	rance codes, which variables are candidates for removal from the model?
\square grade	
☐ male	
\square raceeth	
□ preschoo	I
□ expectBa	chelors
☐ motherH	5
☐ motherBa	achelors
☐ motherW	ork
\square fatherHS	
☐ fatherBac	helors
\square fatherWo	rk
☐ selfBornU	IS .
☐ motherBo	ornUS
☐ fatherBor	nUS
☐ englishAt	Home
	ForSchoolwork
☐ read30Mi	
	erWeekEnglish
☐ studentsI	
☐ schoolHa	
☐ publicSch	ool
□ urban	
☐ schoolSize	e
Show Answer Yo	u have used 0 of 2 submissions
PROBLEM 4.1 - I	PREDICTING ON UNSEEN DATA (1 point possible)
in pisaTest. Call this	unction and supplying the "newdata" argument, use the ImScore model to predict the reading scores of students vector of predictions "predTest". Do not change the variables in the model (for example, do not remove variables not significant in the previous part of this problem). Use the summary function to describe the test set
What is the range be	etween the maximum and minimum predicted reading score on the test set?
Show Answer Yo	u have used 0 of 3 submissions
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PROBLEM 3.5 - IDENTIFYING VARIABLES LACKING STATISTICAL SIGNIFICANCE (1 point possible)

PROBLEM 4.2 - TEST SET SSE AND RMSE (2 points possible)

What is the sum of squared errors (SSE) of lmScore on the testing set?

What is the root-mean squared error (RMSE) of ImScore on the testing set?
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PROBLEM 4.3 - BASELINE PREDICTION AND TEST-SET SSE (2 points possible)
What is the predicted test score used in the baseline model? Remember to compute this value using the training set and not the test set.
What is the sum of squared errors of the baseline model on the testing set? HINT: We call the sum of squared errors for the baseline model the total sum of squares (SST).
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PROBLEM 4.4 - TEST-SET R-SQUARED (1 point possible)
What is the test-set R-squared value of lmScore?
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