est Information escription					
structions					
ultiple Attempts This test a	llows 3 attempts. This is atte	empt number 1.			
rce Completion This test c	an be saved and resumed la	ater.			
Question Completion Status					
QUESTION 1					1 points Sav
between GPA and IQ (i.e	., GPA*IQ), and X5 = Interact thousands of dollars). Sup	ction between GPA and Gen	er (1 for Female and 0 for Male), der (i.e., GPA*Gender). The resp to fit the model, and the intercep	onse is starting	
<ul> <li>b0 = 50; b1 = 20;</li> <li>b3 = 35; b4 = 0.0</li> </ul>					
			rediction using the same unit a, you should input it as 10.	as the response	
QUESTION 2					1 points Sav
Continue with the GPA a	nd Salary Data. Which of th	e following are correct			
	•	•	s provided that their GPA < 3.5.		
	•	9	•		
	,	less on average than males.			
5. For a fixed value of	IQ and GPA, females earn i	more on average than males	S.		
4. For a fixed value of	IQ and GPA, females earn	more on average than males	s provided that their GPA > 3.5.		
5. For a fixed value of	IQ and GPA, females earn	more on average than males	s provided that their GPA < 3.5.		
6. For a fixed value of	IQ and GPA, males earn me	ore on average than females	s provided that their GPA > 3.5.		
QUESTION 3					2 points Sav
Continue with the GPA a is recorded as 1.5), then	nd Salary Data. If we chang	ge the unit of IQ to 100 (i.e.,	IQ score 100 is recorded as IQ=	1, and IQ score 150	
<ul><li>The new LS estim</li><li>The new LS estim</li></ul>	ate for the intercept b0 is er ate for the intercept b2 is er ate for the intercept b3 is er ate for the intercept b4 is er	qual to 7 qual to 35			
					2 points Sav
QUESTION 4			and GPA, the prediction for the s	salary of a <b>female</b>	_
Continue with the GPA a	nd Salary Data. Based on the steed on the st				
	s the following linear combin				
Continue with the GPA a student can be written a	s the following linear combin		and w3=0.01		
Continue with the GPA a student can be written as w0 + w1*GPA + w2*IQ + where w0=85	s the following linear combin w3*(GPA*IQ), , w1=10	, w2= 0.07	and w3=0.01 orded as 100 and score 150 is re	ecorded as 150.	
Continue with the GPA a student can be written as w0 + w1*GPA + w2*IQ + where w0=85  Note: the IQ is still in the	s the following linear combin w3*(GPA*IQ), , w1=10	, w2= 0.07		ecorded as 150.	2 points Sav
Continue with the GPA a student can be written at w0 + w1*GPA + w2*IQ + where w0=85  Note: the IQ is still in the QUESTION 5  Continue with the GPA a	s the following linear combin w3*(GPA*IQ), w1=[10 original scale as in Problen	w2=0.07 n 1, i.e., IQ score 100 is reco			2 points Sav

and the an to as to d	PA and Salary Data. Suppose we enote the five coefficients assoc			denote the intercept	3 points	Saved
• a0 = 85	; a1 = 10	; $a2 = 0.07$	y. Then			
• a3 = -35	; a4 = 0.01	; a5 = 10				
Note: IQ is in the orig	ginal scale as in Problem 1, i.e., I	IQ score 100 is recorded as X2=	:100 and score 150 is record	ded as X2=150.		
QUESTION 7					1 points	Saved
Continue with the GP	PA and Salary Data. What does it	t mean if the LS coefficient for X	(1=GPA is significant?			
1. The absolute va	alue of the correlation between	Y and X1 is large				
2. The conditional	I correlation between Y and X1,	given other four predictors, is no	ot zero			
3. The correlation	between Y and X1 is not zero					
4. The absolute value	alue of the conditional correlation	n between Y and X1, given othe	er four predictors, is large			
QUESTION 8					1 points	Saveo
	PA and Salary Data. Suppose we	add one more predictor to the	model. Which of the following	ng statements are		
true?  1. R-square will in	ncrease					
2. Sigma-hat will						
✓ 3. RSS will decrea						
4. Adjusted R-squ	uare will increase					
QUESTION 9					1 points	Saved
Suppose you want to	o fit a multiple linear regression n	model on a data set with n=100 o	observations. You mistaken	ly duplicate the		
data set, that is, you	end up fitting a multiple linear re	egression model on n=200 obser				
1. Sigma-hat	of the following estimates stay t	unchangeu:				
2. RSS						
✓ 3. R-square						
4 LS coefficients						
					1 points	Saved
QUESTION 10						
	g a linear regression model with	ın samples and p predictors (inc	cluding the intercept). Circle	e correct		
Suppose we are fittin statements.	ng a linear regression model with		.,	correct		
Suppose we are fittin statements.  In the LS coefficients of the statements of the statements of the statement of the stateme		ercept predictor takes a constant	.,	correct		
Suppose we are fittin statements.  1. The LS coeffici	ents are not unique if a non-inter	ercept predictor takes a constant	t value	correct		
Suppose we are fittin statements.  1. The LS coeffici  2. The LS coeffici  3. The LS coeffici	ents are not unique if a non-inter	ercept predictor takes a constant r than or equal to n. nse variable Y takes a constant v	t value	correct		
Suppose we are fittin statements.  1. The LS coeffici. 2. The LS coeffici. 3. Tne LS coeffici. 4. The LS coeffici.	ents are not unique if a non-interents are not unique if p is biggerents are not unique if the respon	ercept predictor takes a constant r than or equal to n. nse variable Y takes a constant v ctors are uncorrelated with each	t value value n other.	correct		
Suppose we are fittin statements.  2 1. The LS coeffici 2. The LS coeffici 3. The LS coeffici 4. The LS coeffici 2 5. The LS coeffici	ents are not unique if a non-interents are not unique if p is biggerents are not unique if the responents are unique only if the predictions.	ercept predictor takes a constant r than or equal to n. nse variable Y takes a constant v ctors are uncorrelated with each	t value value n other.	correct	4 nainte	Rause
Suppose we are fittin statements.  1. The LS coeffici 2. The LS coeffici 3. The LS coeffici 4. The LS coeffici 5. The LS coeffici	ents are not unique if a non-interents are not unique if p is biggerents are not unique if the responents are unique only if the predictions.	ercept predictor takes a constant r than or equal to n. nse variable Y takes a constant v ctors are uncorrelated with each of the n-by-p design matrix X are	t value value n other. e not linear independent.	correct	4 points	Saved
Suppose we are fittin statements.  2 1. The LS coeffici 2. The LS coeffici 3. The LS coeffici 4. The LS coeffici 5. The LS coeffici	ents are not unique if a non-interents are not unique if p is biggerents are not unique if the responents are unique only if the predicents are not unique if columns o	ercept predictor takes a constant or than or equal to n.  The variable Y takes a constant of the n-by-p design matrix X are be lown Housing Data, data for last	t value value n other. e not linear independent.	correct	4 points	Saved
Suppose we are fittin statements.  2 1. The LS coeffici 2. The LS coeffici 3. The LS coeffici 4. The LS coeffici 5. The LS coeffici  The following question Load the trainic columns. The last column dollars (round	ents are not unique if a non-interents are not unique if p is bigger ents are not unique if the responsents are unique only if the predict ents are not unique if columns of ents are not unique if columns of ents are related to a subset of the ing data (train housing quiz2.cs	ercept predictor takes a constant or than or equal to n.  see variable Y takes a constant of the n-by-p design matrix X are be lowa Housing Data, data for last set.  by D. The data has 1460  price for each house. The cheap of the cheapest house the price for each house.	t value  value  n other.  e not linear independent.  st semester's project 1.  rows and 81  pest price is 35 se was sold for 2,400 dollars	thousand s, your answer	4 points	Savec
statements.  1 · The LS coeffici 2 · The LS coeffici 3 · The LS coeffici 4 · The LS coeffici 5 · The LS coeffici  1 · The LS coeffici 2 · The LS coeffici 2 · Load the trainic columns.  • The last columdollars (round)	ents are not unique if a non-interents are not unique if p is bigger ents are not unique if the responsents are unique only if the predicents are unique only if the predicents are not unique if columns of the sare related to a subset of the ing data (train housing quiz2.cs	ercept predictor takes a constant or than or equal to n.  Inse variable Y takes a constant of the n-by-p design matrix X are be lowa Housing Data, data for last set.  Description of the n-by-p design matrix X are be lowed Housing Data, data for last set.  Description of the n-by-p design matrix X are be lowed Housing Data, data for last set.  Description of the n-by-p design matrix X are belowed Housing Data, data for last set.	value value n other. e not linear independent. est semester's project 1. rows and 81	thousand s, your answer an housing price is	4 points	Saved

QUEST	TION 12				1 points	Saved
	nue with the lowa Housing Training Data. We will fit he housing price: OverallQual, X1stFlrSF, GrLivArea,		dictors that are highly con	related		
•	Since the evaluation is based on log of the sale pri	ice, we change SalePrice to log-scale.				
	Predictors that measure amounts (such as price, s tail, i.e., there are some very large values. Suggest dataset, we'll change X1stFlrSF, GrLivArea, and Ga	t to apply square-root or log-transform				
	Fit a linear regression model to predict the log(Sale log(GrLivArea+1), GarageCars, log(GarageArea+1)					
	The R-square of this model is equal to 0.80 0.56.).	(round to the 2nd digit	after the decimal point,	e.g.,		
	The median OverallQual is 6, the median X1stFlrSf					
	the median GarageArea is 480. Use the fitted mod prediction is equal to 1692 hund	lel to predict the SalePrice for a house red dollars (round to the nearest hund)				
	prediction is equal to 1692 hund				1.5 points	Save
QUES1	prediction is equal to 1692 hundi blank is an integer.)	red dollars (round to the nearest hundr	red, i.e., the number you fi	If in the	1.5 points	Save
QUEST Continu log(GrL	prediction is equal to 1692 hunds blank is an integer.)  TION 13  nue with the multiple linear regression model that pr LivArea+1), GarageCars, log(GarageArea+1). Now,  Is the R-square of this new model guaranteed to be	redicts log(SalePrice+1) using predictorefit the model treating GarageCars as	red, i.e., the number you fi ors: OverallQual, log(X1stFl s a categorical variable wit	rSF+1), h 5 levels.	1.5 points	Save
QUEST Continu log(GrL	prediction is equal to 1692 hunds blank is an integer.)  TION 13  nue with the multiple linear regression model that pr LivArea+1), GarageCars, log(GarageArea+1). Now,  Is the R-square of this new model guaranteed to be	redicts log(SalePrice+1) using predictorefit the model treating GarageCars as be bigger than the old linear model that "Yes" or "No".)	ors: OverallQual, log(X1stFls a categorical variable wit treats GarageCars as a n = 0, 1, 2, 3, 4:	rSF+1), h 5 levels.	1.5 points	Save
Continu log(GrL	prediction is equal to 1692 hunds blank is an integer.)  TION 13  nue with the multiple linear regression model that prediction for the prediction of the pr	redicts log(SalePrice+1) using predictorefit the model treating GarageCars as be bigger than the old linear model that "Yes" or "No".)	ors: OverallQual, log(X1stFls a categorical variable wit treats GarageCars as a n = 0, 1, 2, 3, 4: 94*log(GarageArea+1)	rSF+1), h 5 levels.	1.5 points	Save
Continui log(GrL	prediction is equal to 1692 hunds blank is an integer.)  TION 13  nue with the multiple linear regression model that produced the produced form of the produ	redicts log(SalePrice+1) using predictorefit the model treating GarageCars as the bigger than the old linear model that with "Yes" or "No".)  In models for houses with GarageCars (1stFlrSF+1) + b3"log(GrLivArea+1) + bgeCars = 1 (Round to the first digit at the that treats GarageCars as a numeric	ors: OverallQual, log(X1stFls a categorical variable wit treats GarageCars as a n = 0, 1, 2, 3, 4: 4*log(GarageArea+1) fter the decimal point)	rSF+1), h 5 levels. umerical	1.5 points	Save
Continui log(GrL	prediction is equal to 1692 hundiblank is an integer.)  TION 13  nue with the multiple linear regression model that prediction for the prediction of the pre	redicts log(SalePrice+1) using predictorefit the model treating GarageCars as the bigger than the old linear model that with "Yes" or "No".)  In models for houses with GarageCars (1stFlrSF+1) + b3"log(GrLivArea+1) + bgeCars = 1 (Round to the first digit at the that treats GarageCars as a numeric	ors: OverallQual, log(X1stFls a categorical variable wit treats GarageCars as a n = 0, 1, 2, 3, 4: 04*log(GarageArea+1)  fter the decimal point)  real variable and the new or	rSF+1), h 5 levels. umerical	1.5 points	Save
Continuing (GrL	prediction is equal to 1692 hundiblank is an integer.)  TION 13  nue with the multiple linear regression model that prediction for the prediction of the pre	redicts log(SalePrice+1) using predictorefit the model treating GarageCars as the bigger than the old linear model that with "Yes" or "No".)  In models for houses with GarageCars (1stFlrSF+1) + b3"log(GrLivArea+1) + bgeCars = 1 (Round to the first digit at the that treats GarageCars as a numerication of this test?	ors: OverallQual, log(X1stFls a categorical variable wit treats GarageCars as a n = 0, 1, 2, 3, 4: 04*log(GarageArea+1)  fter the decimal point)  real variable and the new or	rSF+1), h 5 levels. umerical	1.5 points	Savec