## THE UNIVERSITY OF BRITISH COLUMBIA FINAL EXAMINATION, April 2016 STAT 306, Finding Relationships in Data

TIME:  $2 \ 1/2$  hours

## THIS EXAMINATION CONSISTS OF 4 PAGES. THERE ARE 6 PROBLEMS FOR A TOTAL OF 100 POINTS. PLEASE CHECK TO ENSURE THAT THIS PAPER IS COMPLETE.

**Special Instructions:** Please write your name and student number on the front page of the exam booklet. A non-programmable calculator and a two-sided sheet of notes are allowed. The problems can be done in any order. READ THE QUESTIONS CAREFULLY.

Suggestion: use the left-hand side of booklet for scratch work, and the right-hand side for submitted answers. For a large-sample z or t critical value for a 95% confidence interval, use 1.96 or 2. For smaller degrees of freedom,  $t_{35,0.975} = 2.030$ ,  $t_{24,0.975} = 2.064$ ,  $t_{21,0.975} = 2.080$ ,  $t_{10,0.975} = 2.228$ ,  $t_{3,0.975} = 3.182$ .

(16) 1. Scientists are assessing if the CO2 concentration at a single location can be used to estimate the mean global annual temperature. The variables are x=CO2 concentration (in parts per 10<sup>8</sup>) at top of Mauna Loa in Hawaii and y=mean annual temperature in Celsius (over land and water across the globe). Consider the data for 1959–1995 (n = 37 years); for example (x, y) = (2.452, 16.53) in 1959 and (x, y) = (6.412, 16.83) in 1995. Summary statistics are:  $\bar{x} = 4.639$ ,  $s_x = 1.226$ ,  $\bar{y} = 16.621$ ,  $s_y = 0.098$ ,  $r_{xy} = 0.508$ .

- (a) What are the slope  $\hat{\beta}_1$  and intercept  $\hat{\beta}_0$  of the least squares regression line for y as a function of x?
- (b) What is the residual SD :  $\hat{\sigma} = \sqrt{(n-2)^{-1} \sum_{i=1}^{n} (y_i \hat{\beta}_0 \hat{\beta}_1 x_i)^2}$ ?
- (c) Suppose  $(x_i, y_i)$  are indexed by increasing year. Describe a residual plot to check if the residuals are serially correlated.
- (d) Assuming that the residual plot (c) is acceptable, what is a prediction interval for mean temperature of the globe when the CO2 concentration is 6.5 parts per 10<sup>8</sup>?

(16) 2. An experiment was done to find the rate of bottle return for 6 different deposit levels from 2 to 30 cents. Data and logistic regression output are shown below.

	deposit #sold #returned	$   \begin{array}{c}     2 \\     100 \\     14   \end{array} $	5 100 20	$10\\100\\34$	$20 \\ 100 \\ 59$	$25 \\ 100 \\ 81$	$30 \\ 100 \\ 90$
Estimate Std (Intercept) -2.10829 deposit 0.13715 Null deviance: 831.7 Residual deviance: 609.4 AIC: 613.41	0.19107 -1 0.01072 1 5 on 599	1.03 2.79 degre	<2e <2e es of	-16 -16 free			

- (a) Obtain the sample proportions for the 6 deposit levels. Compare with the fitted proportions using the estimated  $\beta$ 's given above. Is the logistic model a good fit?
- (b) Assuming the logistic model is reasonable, what is an approximate 95% confidence interval for the slope parameter  $\beta_1$ ?
- (c) What is the estimated probability that the bottle will be returned if the deposit is 15 cents?
- (d) What is the estimated deposit level so that the expected return rate is 50%?

(18) 3. http://www.cdc.gov/nchs/nhanes.htm has data sets from the National Health and Nutrition Examination Survey. Some variables such as total (body) fat are based on dual-energy x-ray absorptiometry, so are harder to measure. Other body measurements such as weight (wt), height (ht), body mass index (bmi), waist circumference, triceps skin fold (tri) are easier to measure. So one might try to find a prediction equation for total fat (tofat) based on these other body measurements. Below are some summary statistics for a sample of 631 men between the ages of 20 and 40 in the 2003–2004 survey. The right-hand side has the sample correlation matrix.

summary	wt	ht	bmi	waist	tri	to fat	cor	wt	ht	hmi	waist	tri	tofat
Min.	42.80	154.1	16.01	63.70	3.50	6.785							0
1st-Qu.	70.60	170.8	23 36	84.30	9.30	15.854	wt	1.000	0.407		0.885		
Median				92.30			ht	0.407	1.000	-0.046	0.109	0.010	0.149
			-				bmi	0.891	-0.046	1.000	0.916	0.665	0.887
3rd-Qu.	91.50	180.9	29.41	100.45	18.40	26.079	maist	0.885	0.109	0.916	1.000	0 705	0.942
Max.	132.40	193.0	41.97	138.30	38.80	55.052			0.200				
Mean	81.75	175.5	26.51	92.82	14.19	21.622	tri	0.610	0.010		0.705		
SD	15.23		4.47	12.22	-	8.097	to fat	0.880	0.149	0.887	0.942	0.795	1.000
SD	10.20	1.40	4.47	12.22	0.04	0.097							

Next are some summaries of the best fitting 2-, 3-, 4- and 5-variable models from regsubsets() in R.

2-variable	Estimate	SE	tvalue	I	3-variable	Estimate	e SE	tvalue
(Intercept)	-29.657	0.803	-36.94	Ι	(Intercept)	-26.985	0.787	-34.28
waist	0.503	0.010	48.25	Ι	wt	0.121	0.012	10.29
tri	0.321	0.019	16.49	Ι	waist	0.366	0.016	22.30
				Ι	tri	0.329	0.018	18.23
4-variable	Estimate	SE	tvalue	Ι	5-variable	Estimate	SE	tvalue
(Intercept)	-24.373	2.886	-8.45	Ι	(Intercept)	20.794	12.322	1.69
wt	0.131	0.016	8.26	Ι	wt	0.408	0.075	5.43
ht	-0.014	0.015	-0.94	Ι	ht	-0.271	0.070	-3.88
waist	0.357	0.019	18.47	Ι	bmi	-0.892	0.237	-3.77
tri	0.327	0.018	18.06	Ι	waist	0.367	0.019	19.00
				Ι	tri	0.329	0.018	18.33

Other summaries of the above fits are:

#vars	residSD	df	$R^2$	$\mathrm{adj}R^2$	$C_p$	CVRMSE
2	2.268	628	0.9218	0.9216	123.3	2.277
3	2.099	627	0.9331	0.9328	17.1	2.111
4	2.099	626	0.9332	0.9328	18.2	2.113
5	2.077	625	0.9347	0.9342	6.0	2.094

- (a) Which of the 5 explanatory variables leads to the best 1-variable simple linear regression model? Why?
- (b) What is the adjusted  $R^2$  value for the 1-variable model in (a)?
- (c) What is the partial correlation  $r_{\text{tofat,tri;waist}}$ ?
- (d) Consider the partial correlation  $r_{\text{tofat,ht;wt,bmi,waist,tri}}$ . Which of the following is correct based on the above summaries: (i)  $r_{\text{tofat,ht;wt,bmi,waist,tri}} > 0$ ; (ii)  $r_{\text{tofat,ht;wt,bmi,waist,tri}} < 0$ ; (iii) insufficient information to determine the sign. Explain your choice.
- (e) Which of the four models would you prefer? Explain your reasoning in one sentence.
- (f) Why does the coefficient for wt change so much among the models with 3,4,5 explanatory variables?

(21) 4. A paper helicopter experiment (http://www.paperhelicopterexperiment.com/) was run to find some optimal dimensions. Explanatory variables are body length and body width (both in cm) of a piece of paper before the folding/cutting is done to produce the helicopter. The response variable y is the flight time (to land on floor) after release from a height of 2.5 m. With variables abbreviated as len and wid, the data are:

len	wid	flighttime									
5.6	1.4	1.98	6.6	1.4	1.85	6.1	2.1	1.82	6.6	2.1	2.00
6.1	1.4	1.89	6.1	1.4	1.87	6.1	0.7	1.43	5.6	0.7	1.25
6.6	2.1	2.08	6.6	2.1	1.86	6.1	1.4	1.73	6.6	0.7	1.00
6.1	0.7	1.41	5.6	0.7	1.10	6.6	0.7	0.93	6.1	2.1	1.85
5.6	1.4	1.87	5.6	2.1	1.69	5.6	1.4	1.83	5.6	0.7	1.21
6.1	0.7	1.30	5.6	2.1	1.69	6.6	1.4	1.98	5.6	2.1	1.61
6.6	0.7	0.93	6.6	1.4	1.94	6.1	2.1	1.94			

Summaries from fitting a linear equation, a quadratic with original variables and a quadratic with centred variables (clen=len-6, cwid=wid-1.5) are given below.

Estimate Std.Error tvalue Pr(>|t|) fit1 (Intercept) 0.736 0.702 1.049 0.305 0.038 0.113 0.334 0.742 len wid 0.475 0.081 5.866 5e-06 \*\*\* Residual SD: 0.2403 on 24 df; Multiple R2: 0.5899, Adjusted R2: 0.5557

fit2 Estimate SE tvalue Pr(>|t|) | fit2c Estimate SE tvalue pval Intercept -10.994 6.722 -1.635 0.117 1.973 0.046 42.720 2e-16 \*\*\* | Intercept len 4.042 2.200 1.837 0.080 | clen 0.152 0.064 2.376 0.027 \* wid 0.231 0.612 0.376 0.710 cwid 0.282 0.042 6.641 1e-06 \*\*\* -0.373 0.180 -2.074 0.051 . | I(clen<sup>2</sup>) -0.373 0.180 -2.074 0.051 .  $I(len^2)$ -0.769 0.092 -8.370 4e-08 \*\*\* | I(cwid^2) I(wid<sup>2</sup>) -0.769 0.092 -8.370 4e-08 \*\*\* 0.393 0.091 4.321 .0003 \*\*\* | clen:cwid 0.393 0.091 4.321 .0003 \*\*\* len:wid Residual SD : 0.1102 on 21 df | Residual SD: 0.1102 on 21 df Multiple R2: 0.9245, AdjustedR2: 0.9065 | Multiple R2: 0.9245, AdjustedR2: 0.9065

 $(\mathbf{X}^T \mathbf{X})^{-1}$  for the third regression (labelled fit2c) is given below.

0.1755	0.0607	-0.0319	-0.4178	-0.2198	-0.0068
0.0607	0.3357	-0.0068	-0.5333	-0.0000	0.0680
-0.0319	-0.0068	0.1479	0.0000	0.1388	-0.0680
-0.4178	-0.5333	0.0000	2.6667	0.0000	0.0000
-0.2198	-0.0000	0.1388	0.0000	0.6942	0.0000
-0.0068	0.0680	-0.0680	0.0000	0.0000	0.6803

- (a) What are three things in the regression summaries that indicate that the quadratic function fit is better than the linear function?
- (b) For the quadratic model, which  $\beta$ 's are invariant to the shifting of length and width by constants?
- (c) What are the best residual plots to check that the quadratic model adequately handles the curved surface for flight time as a function of length and width?
- (d) For the **X** matrix for quadratic with centred variables, what are the values of the 6 columns in row 1?
- (e) Using the above summaries, obtain a point estimate, prediction SE and a prediction interval from the quadratic fit when length=6 cm, width=1.5 cm or clen=0, cwid=0. (Note: little arithmetic is needed).
- (f) Based on the fitted quadratic, what are the estimated optimal length and width to maximize flight time?

(14) 5. Consider a data set to daily log returns of 8 stocks in the American market in 2010–2011 (504 trading days); the stocks are: LO (Lorillard Tobacco), MO (Altria Tobacco), PM (Philip Morris Tobacco), RAI (Reynolds Tobacco), DPS (Dr Pepper Snapple Beverage), KO (Coca-Cola), MNST (Monster Beverage), PEP (Pepsico). Principal component analysis was applied to both the sample covariance matrix and sample correlation matrix. A summary of some results are below.

		LO	MO		PM	RAI	DPS	KO	MNST	PEP
sample m	ean 0	.0009	0.0010	0.0	011  0.	0011 (	).0007	0.0005	0.0017	0.0003
sample SI	D   0	.0149	0.0100	0.0	125  0.	0119 (	0.0158	0.0104	0.0202	0.0101
row 1	-0	.0272	0.0006	-0.0	088 0.	0014 -0	0.0116	-0.0122	0.0041	0.0120
sample covariance matrix l sample correlation matrix										
	-		Comp	1	Comp1	Compo	Comp3	Comp		
-	-	-	-		-	-	-	-		
				-						
/ar 0.515	0.175	0.117	0.071			0.104	0.087	0.080		
0.515	0.690	0.807	0.878		0.552	0.657	0.743	0.824		
gs: sampl	Le cova:	riance	matrix		sample	e corre	lation	natrix		
Comp1	Comp2	Comp3	Comp4		Comp1	Comp2	Comp3	Comp4		
-0.363	-0.389	0.383	0.689		-0.331	-0.504	_	-0.401		
-0.274	-0.173		-0.139		-0.395	-0.216				
-0.337	-0.206	0.187	-0.347	1	-0.381	-0.194	-0.102	0.163		
-0.335	-0.218	0.145	-0.122	i İ	-0.400	-0.250				
				•			0 600	-0 544		
				-						
				-						
-0.246	-0.105		-0.378		-0.356	0.271	0.255	0.483		
	sample SI row 1 samp ance of co Comp1 0.028 Jar 0.515 0 0.515 gs: samp Comp1 -0.363 -0.274 -0.335 -0.362 -0.278 -0.546	sample mean sample SD       0 0 row 1         sample SD       0         row 1       -0         sample cove       0         ance of component Comp1 Comp2       0.028         0.028       0.016         Var 0.515       0.175         0       0.515         0       0.515         0       0.515         0       0.515         0       0.515         0.515       0.690         gs:       sample cover         Comp1 Comp2       -0.363         -0.363       -0.389         -0.274       -0.173         -0.335       -0.218         -0.362       -0.174         -0.278       -0.102         -0.546       0.823	sample mean sample SD     0.0009       row 1     0.0149       row 1     -0.0272       sample covariance       ance of components:       Comp1     Comp2       Comp1     Comp3       0.028     0.016     0.013       Jar 0.515     0.175     0.117       o     0.515     0.690     0.807       gs:     sample covariance       Comp1     Comp2     Comp3       -0.363     -0.389     0.383       -0.274     -0.173       -0.337     -0.206     0.187       -0.335     -0.218     0.145	sample mean sample SD       0.0009       0.0010         sample SD       0.0149       0.0100         row 1       -0.0272       0.0006         sample covariance matrix       ance of components:       Comp1       Comp2       Comp3       Comp4         0.028       0.016       0.013       0.010         Var       0.515       0.175       0.117       0.071         0       0.515       0.690       0.807       0.878         gs:       sample covariance matrix       Comp1       Comp2       Comp3       Comp4         -0.363       -0.389       0.383       0.689       -0.274       -0.173       -0.139         -0.337       -0.206       0.187       -0.347       -0.335       -0.218       0.145       -0.122         -0.362       -0.174       -0.879       0.223       -0.278       -0.402       -0.402         -0.546       0.823       0.126       0.126       0.126       0.126	sample mean sample SD     0.0009     0.0010     0.0       sample SD     0.0149     0.0100     0.0       row 1     -0.0272     0.0006     -0.0       sample covariance matrix             ance of components:     Comp1     Comp2     Comp3     Com4             0.028     0.016     0.013     0.010             Var     0.515     0.175     0.117     0.071             po     0.515     0.690     0.807     0.878             gs:     sample covariance matrix                gs:     sample covariance matrix                gs:     sample covariance matrix                gs:     sample covariance matrix                 gs:     sample covariance matrix                    gs:     sample covariance matrix	sample mean sample SD     0.0009     0.0010     0.0011     0.0       row 1     0.0149     0.0100     0.0125     0.0       sample SD     0.0149     0.0100     0.0125     0.0       row 1     -0.0272     0.0006     -0.0088     0.0       sample covariance matrix     sample       ance of components:     Comp1     Comp2     Comp3     Comp4     Comp1       0.028     0.016     0.013     0.010     2.102       Var 0.515     0.175     0.117     0.071     0.552       po     0.515     0.690     0.807     0.878     0.552       gs:     sample covariance matrix     sample       Comp1     Comp2     Comp3     Comp4     Comp1       -0.363     -0.389     0.383     0.689     -0.331       -0.274     -0.173     -0.139     -0.395     -0.337       -0.335     -0.218     0.145     -0.122     -0.400       -0.362     -0.174     -0.879     0.223     -0.285       -0.278     -0.102     -0.402     -0.384       -0.546	sample mean sample SD     0.0009     0.0010     0.0011 <t< td=""><td>sample mean sample SD     0.0009     0.0010     0.0011     0.0011     0.0007       now 1     0.0149     0.0100     0.0125     0.0119     0.0158       now 1     -0.0272     0.0006     -0.0088     0.0014     -0.0116       sample covariance matrix   sample correlation       ance of components:       Comp1     Comp2     Comp3     Comp4       Comp1     Comp2     Comp3       0.028     0.016     0.013     0.010       2.102     0.913     0.832       Var     0.515     0.175     0.117     0.071       0.552     0.104     0.087       0     0.515     0.690     0.807     0.878       0.552     0.657     0.743       gs:     sample covariance matrix   sample correlation n     Comp1     Comp2     Comp3       -0.363     -0.389     0.383     0.689       -0.331     -0.504       -0.274     -0.173     -0.139       -0.381     -0.194     -0.102       -0.335     -0.218     0.145     -0.122       -0.400     -0.250       -0.362     -0.174     -0.879</td><td>sample mean sample SD     0.0009     0.0010     0.0011     0.0011     0.0007     0.0005       row 1     -0.0272     0.0006     -0.0088     0.0014     -0.0116     -0.0122       sample covariance matrix   sample correlation matrix       ance of components:       Comp1     Comp2     Comp3     Comp4       Comp1     Comp2     Comp3     Comp4       0.028     0.016     0.013     0.010       2.102     0.913     0.832     0.802       Var     0.515     0.175     0.117     0.071       0.552     0.104     0.087     0.080       pb     0.515     0.690     0.807     0.878       0.552     0.657     0.743     0.824       gs:     sample covariance matrix   sample correlation matrix     Comp1     Comp2     Comp3     Comp4       -0.363     -0.389     0.383     0.689       -0.331     -0.504     -0.401       -0.274     -0.173     -0.139       -0.381     -0.194     -0.102     0.163       -0.335     -0.218     0.145     -0.122       -0.400     -0.250     -0.362     -</td><td>sample mean sample SD row 1     0.0009     0.0010     0.0011     0.0011     0.0007     0.0005     0.0017       sample SD row 1     0.0149     0.0100     0.0125     0.0119     0.0158     0.0104     0.0202       -0.0272     0.0006     -0.0088     0.0014     -0.0116     -0.0122     0.0041       sample covariance matrix   sample correlation matrix       ance of components:       Comp1 Comp2 Comp3 Comp4   Comp1 Comp2 Comp3 Comp4       0.028     0.016     0.013     0.010     2.102     0.913     0.832     0.802       Var 0.515     0.175     0.117     0.071     0.552     0.104     0.087     0.080       o     0.515     0.690     0.807     0.878     0.552     0.657     0.743     0.824       gs: sample covariance matrix   sample correlation matrix       Comp1     Comp2     Comp3     Comp4     -0.401       -0.363     -0.389     0.383     0.689     -0.331     -0.102     0.163       -0.377     -0.206     0.187     -0.381     -0.194     -0.102     0.163       -0.335</td></t<>	sample mean sample SD     0.0009     0.0010     0.0011     0.0011     0.0007       now 1     0.0149     0.0100     0.0125     0.0119     0.0158       now 1     -0.0272     0.0006     -0.0088     0.0014     -0.0116       sample covariance matrix   sample correlation       ance of components:       Comp1     Comp2     Comp3     Comp4       Comp1     Comp2     Comp3       0.028     0.016     0.013     0.010       2.102     0.913     0.832       Var     0.515     0.175     0.117     0.071       0.552     0.104     0.087       0     0.515     0.690     0.807     0.878       0.552     0.657     0.743       gs:     sample covariance matrix   sample correlation n     Comp1     Comp2     Comp3       -0.363     -0.389     0.383     0.689       -0.331     -0.504       -0.274     -0.173     -0.139       -0.381     -0.194     -0.102       -0.335     -0.218     0.145     -0.122       -0.400     -0.250       -0.362     -0.174     -0.879	sample mean sample SD     0.0009     0.0010     0.0011     0.0011     0.0007     0.0005       row 1     -0.0272     0.0006     -0.0088     0.0014     -0.0116     -0.0122       sample covariance matrix   sample correlation matrix       ance of components:       Comp1     Comp2     Comp3     Comp4       Comp1     Comp2     Comp3     Comp4       0.028     0.016     0.013     0.010       2.102     0.913     0.832     0.802       Var     0.515     0.175     0.117     0.071       0.552     0.104     0.087     0.080       pb     0.515     0.690     0.807     0.878       0.552     0.657     0.743     0.824       gs:     sample covariance matrix   sample correlation matrix     Comp1     Comp2     Comp3     Comp4       -0.363     -0.389     0.383     0.689       -0.331     -0.504     -0.401       -0.274     -0.173     -0.139       -0.381     -0.194     -0.102     0.163       -0.335     -0.218     0.145     -0.122       -0.400     -0.250     -0.362     -	sample mean sample SD row 1     0.0009     0.0010     0.0011     0.0011     0.0007     0.0005     0.0017       sample SD row 1     0.0149     0.0100     0.0125     0.0119     0.0158     0.0104     0.0202       -0.0272     0.0006     -0.0088     0.0014     -0.0116     -0.0122     0.0041       sample covariance matrix   sample correlation matrix       ance of components:       Comp1 Comp2 Comp3 Comp4   Comp1 Comp2 Comp3 Comp4       0.028     0.016     0.013     0.010     2.102     0.913     0.832     0.802       Var 0.515     0.175     0.117     0.071     0.552     0.104     0.087     0.080       o     0.515     0.690     0.807     0.878     0.552     0.657     0.743     0.824       gs: sample covariance matrix   sample correlation matrix       Comp1     Comp2     Comp3     Comp4     -0.401       -0.363     -0.389     0.383     0.689     -0.331     -0.102     0.163       -0.377     -0.206     0.187     -0.381     -0.194     -0.102     0.163       -0.335

- (a) How many components are needed to explain 80% of the variation in the data using the sample covariance matrix?
- (b) Interpret the first two components from the coefficients of the loadings for both sets of output.
- (c) The first observation in the data set is given in the above table along with the sample means and SDs of the 8 variables. For observation 1, what is the value of comp1 for the left-hand side (in the new coordinate system centred at the vector of sample means). Write down an expression without doing the calculation.

(15) 6. Let data be  $(x_i, y_i)$ , i = 1, ..., n, where all of the  $x_i$  are positive. For simple linear regression with model  $Y_i = \beta_0 + \beta_1 x_i + \epsilon_i$ , i = 1, ..., n, the least squares estimate of  $\beta_1$  is  $\hat{\beta}_1 = \sum_{i=1}^n a_i y_i$  where  $a_i = (x_i - \overline{x})/[(n-1)s_x^2]$ , and the least squares intercept is  $\hat{\beta}_0 = \overline{y} - \hat{\beta}_1 \overline{x}$ . Let  $\hat{B}_1$  be the least squares slope when considered as a random variable. Determine the variance of  $\hat{B}_1$  under the two scenarios given below.

- (a) (heteroscedastic)  $\epsilon_i$  are independent  $N(0, \sigma_i^2)$  random variables, where  $\sigma_i^2 = \gamma_0 + \gamma_1 x_i$  with  $\gamma_0 > 0$  and  $\gamma_1 > 0$  being parameters. What is Var  $(Y_i)$ ? What is Var  $(\hat{B}_1)$ ?
- (b) (serial dependence with *i* an index for a time sequence)  $\epsilon_i$  are serially dependent  $N(0, \sigma^2)$  random variables such that  $\text{Cov}(\epsilon_i, \epsilon_j) = \sigma^2 \gamma^{|i-j|}$  for  $i, j \in \{1, \ldots, n\}$ , where  $0 < \gamma < 1$ . What is  $\text{Var}(\hat{B}_1)$ ?
- (c) Assume the model in (b), what is the correlation of  $\epsilon_1, \epsilon_2$  (equivalently, the correlation of  $\epsilon_i, \epsilon_{i+1}$ )?
- (d) [Harder.] Assume the model in (b). Let  $e_i = y_i \hat{y}_i$  and  $\hat{y}_i = \hat{\beta}_0 + \hat{\beta}_1 x_i$  for i = 1, ..., n. Based on your answers in (b) and (c), suggest an estimated standard error of  $\hat{\beta}_1$  that is a function of the  $e_i$  and  $x_i$ .