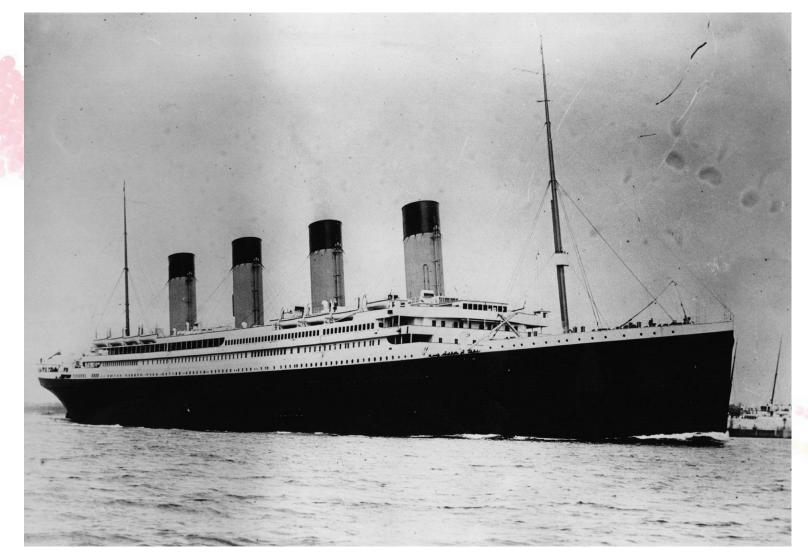
statgraphics®

Would you have survived the sinking of the Titanic?

Using the Statgraphics mosaic charts, multiple correspondence analysis and logistic regression procedures.

Presented by Dr. Neil W. Polhemus

RMS Titanic





Sinking of the Titanic

- Struck an iceberg on April 14, 1912 and sank the next morning.
- Carried over 2000 passengers and crew, many of whom did not survive.
- Controversies have centered on the lack of sufficient lifeboats and the differential treatment of individuals based on class of accommodation.



Sample Data File #1

		titan	2	3					
		Class	Age	Sex	Survived	4			
	1	1	1	1	1				
	2	1	1	1	1				
	3	1	1	1	1				
	4	1	1	1	1				
	5	1	1	1	1				
	6	1	1	1	1				
	7	1	1	1	1				
	8	1	1	1	1				
	9	1	1	1	1				
1	LO	1	1	1	1	-			
14	I I I I I I I I I I I I I I I I I I I								

VARIABLE DESCRIPTIONS: Class (1 = first, 2 = second, 3 = third, 0 = crew) Age (1 = adult, 0 = child) Sex (1 = male, 0 = female) Survived (1 = yes, 0 = no)

n=2,201 observations (all people at risk)

Source: Robert J. MacG. Dawson, Saint Mary's University http://www.amstat.org/publications/jse/datasets/titanic.txt



Sample Data File #2

	pclass	survived	name	sex	age	sibsp	parch	ticket	fare	cabin	embarked	boat	body	home.dest
	perass	Surviveu	Induce	SCA	aye	SIDSD	paron	CICKED	Tare	Cabin	enubarkeu	Duat	bouy	nome.dest
-	1	1.	111 Mi	f	29	0	0	24160	011 0075	B5	S	2		Co. Tauri a MO
1	±	<u>_</u>	Allen, Miss.			-	-		211.3375		-	-		St Louis, MO
2	1 🗟	1	Allison, Mast		0.9167	1	2	113781	151.5500	C22 C26	S	11		Montreal, PQ
3	1	0	Allison, Miss	female	2	1	2	113781	151.5500	C22 C26	S			Montreal, PQ
4	1	0	Allison, Mr.	male	30	1	2	113781	151.5500	C22 C26	S		135	Montreal, PQ
5	1	0	Allison, Mrs.	female	25	1	2	113781	151.5500	C22 C26	S			Montreal, PQ
6	1	1	Anderson, Mr.	male	48	0	0	19952	26.5500	E12	S	3		New York, NY
7	1	1	Andrews, Miss	female	63	1	0	13502	77.9583	D7	S	10		Hudson, NY
8	1	0	Andrews, Mr.	male	39	0	0	112050	0.0000	A36	S			Belfast, NI
9	1	1	Appleton, Mrs	female	53	2	0	11769	51.4792	C101	S	D		Bayside, Que
10	1	0	Artagaveytia,	male	71	0	0	PC 17609	49.5042		С		22	Montevideo,
11	1	0	Astor, Col. J	male	47	1	0	PC 17757	227.5250	C62 C64	С		124	New York, NY
12	1	1	Astor, Mrs. J	female	18	1	0	PC 17757	227.5250	C62 C64	С	4		New York, NY
13	1	1	Aubart, Mme.	female	24	0	0	PC 17477	69.3000	B35	С	9		Paris, France
14	1	1	Barber, Miss.	female	26	0	0	19877	78.8500		S	6		
15	1	1	Barkworth, Mr	male	80	0	0	27042	30.0000	A23	S	в		Hessle, York
16	1	0	Baumann, Mr.	male		0	0	PC 17318	25.9250		S			New York, NY
17	1	0	Baxter, Mr. Ç	male	24	0	1	PC 17558	247.5208	B58 B60	с			Montreal, PQ
18	1	1	Baxter, Mrs.	female	50	0	1	PC 17558	247.5208	B58 B60	С	6		Montreal, PQ
19	1	1	Bazzani, Miss	female	32	0	0	11813	76.2917	D15	С	8		
	1 Îtitanic3 B	1	Bazzani, Miss		32	0	0	11813	76.2917	D15	С	8		

n=1,309 observations (passengers only)

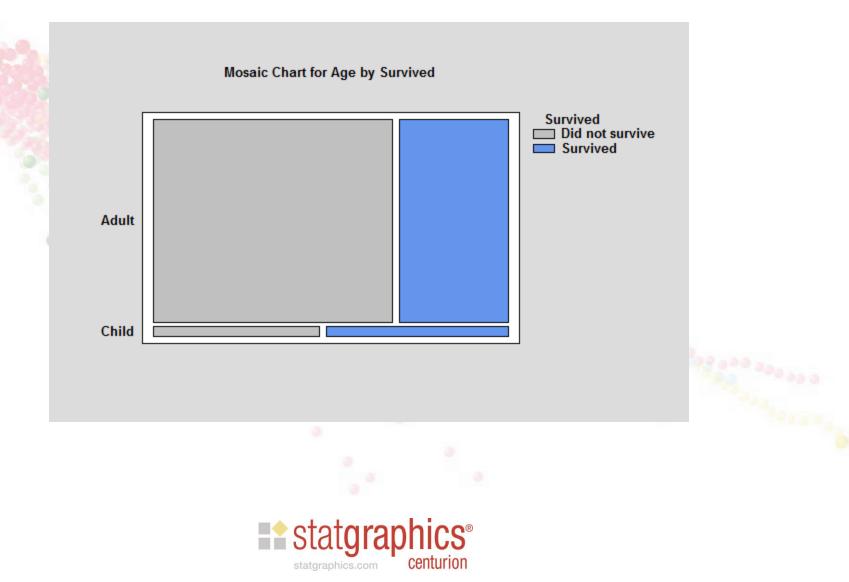
Note that age is numeric.

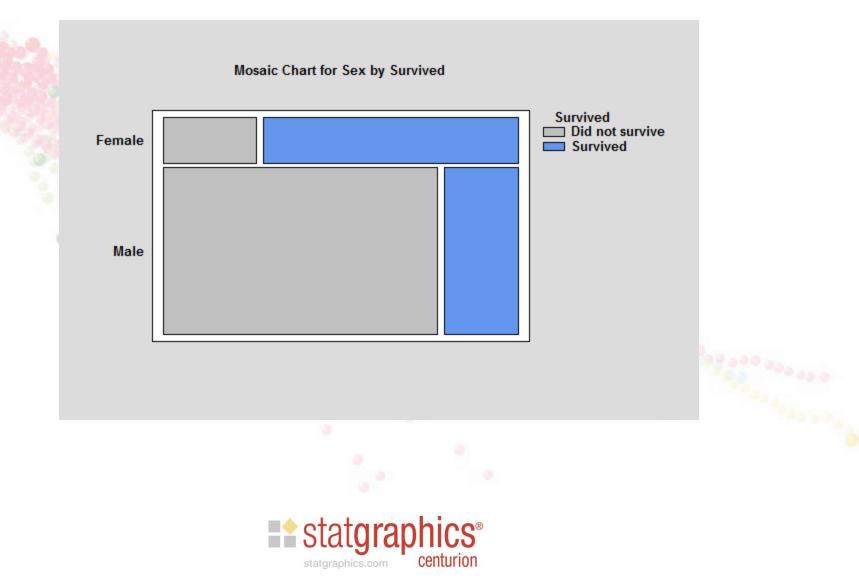
Source: Frank Harrell and Thomas Cason, University of Virginia http://biostat.mc.vanderbilt.edu/twiki/pub/Main/DataSets/titanic.html

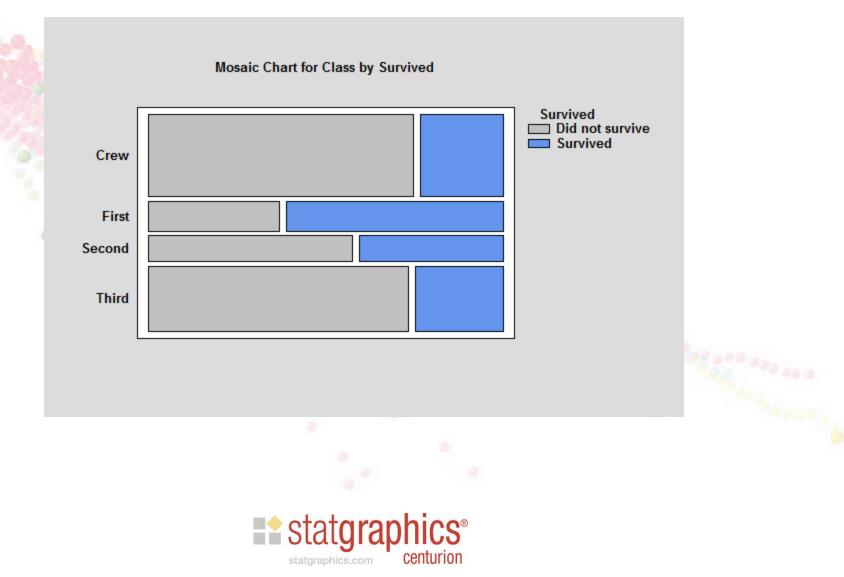


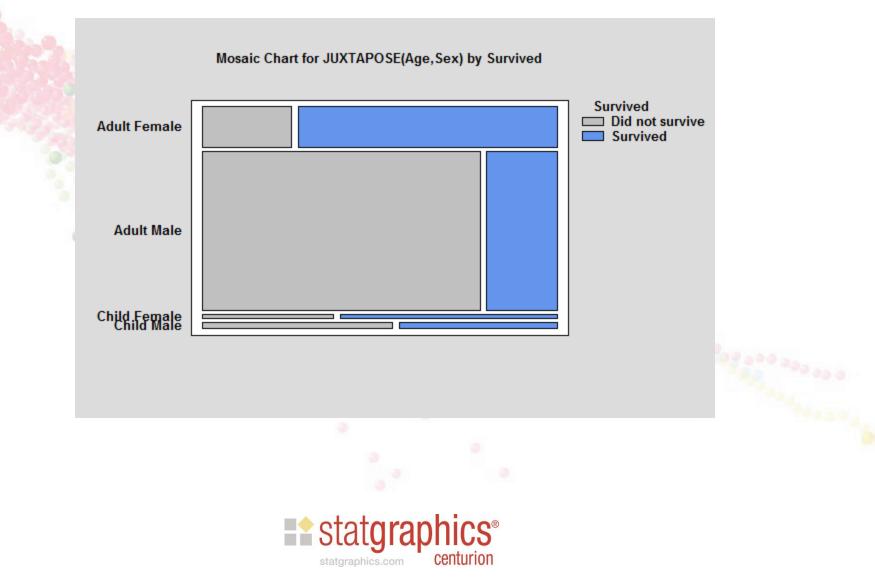
Crosstabulation

		Crosstabulation	
	Class Age Sex Survived	Row Variable:	
		Column Variable:	
•		(Counts:)	
		(Select:)	P & P
	Sort column names		0.000
	OK Cancel	Delete Transform Help	
	State	tatgraphics®	









Multiple Correspondence Analysis

- Creates a map of the associations amongst the categories of 2 or more variables.
- Seeks a small number of dimensions that describe most of the variability or "inertia" amongst the categories.
- Very useful in demonstrating the interrelationships amongst the variables.



Data Input

Calles	Multiple	Correspondence Analysis	×
	Class Age Sex Survived	Columns: Class Age Sex Survived	
· · · · · · · · · · · · · · · · · · ·		(Counts:) (Select:)	
	Sort column names	Delete Transform Help	
	statg	atgraphics®	

Burt Table

Burt Table

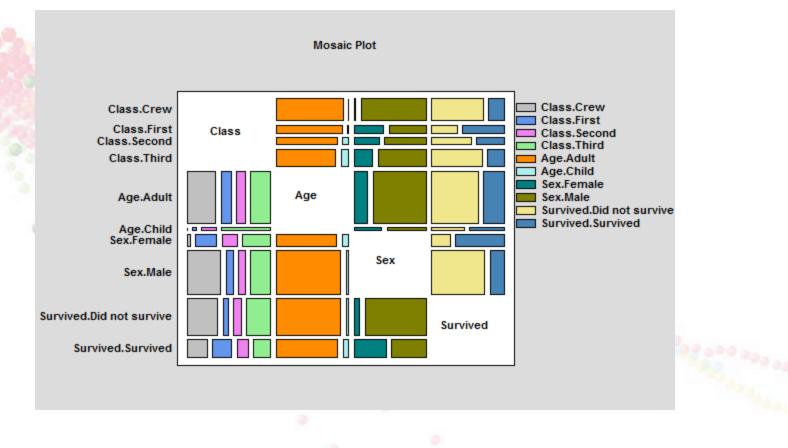
	Class.Crew	Class.First	Class.Second	Class.Third	Age.Adult	Age.Child
Class.Crew	885	0	0	0	885	0
Class.First	0	325	0	0	319	6
Class.Second	0	0	285	0	261	24
Class.Third	0	0	0	706	627	79
Age.Adult	885	319	261	627	2092	0
Age.Child	0	6	24	79	0	109
Sex.Female	23	145	106	196	425	45
Sex.Male	862	180	179	510	1667	64
Survived.Did not survive	673	122	167	528	1438	52
Survived.Survived	212	203	118	178	654	57

	Sex.Female	Sex.Male	Survived.Did not survive	Survived.Survived
Class.Crew	23	862	673	212
Class.First	145	180	122	203
Class.Second	106	179	167	118
Class.Third	196	510	528	178
Age.Adult	425	1667	1438	654
Age.Child	45	64	52	57
Sex.Female	470	0	126	344
Sex.Male	0	1731	1364	367
Survived.Did not survive	126	1364	1490	0
Survived.Survived	344	367	0	711





Multiple Mosaic Plot





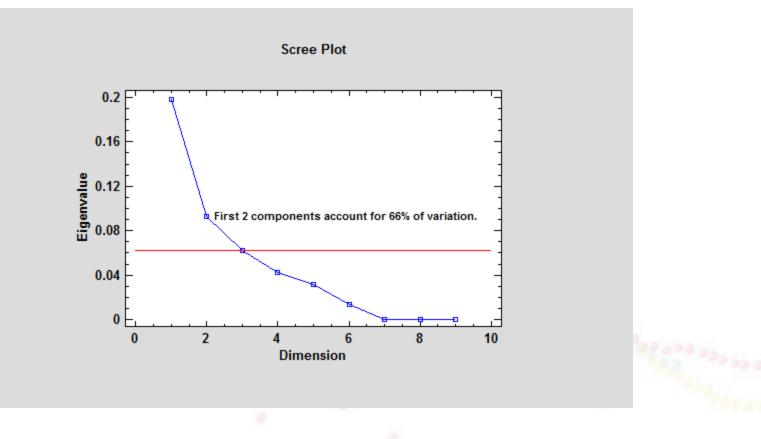
Inertia and Chi-Square Decomposition

Inertia and Chi-Square Decomposition

	Singular		Chi-		Cumulative	
Dimension	Value	Inertia	Square	Percentage	Percentage	Histogram
1	0.4451	0.1981	6976.1395	44.9107	44.9107	*****
2	0.3050	0.0931	3276.9079	21.0959	66.0066	*****
3	0.2500	0.0625	2201.1057	14.1702	80.1768	****
4	0.2050	0.0420	1480.4911	9.5310	89.7078	***
5	0.1785	0.0319	1122.2516	7.2248	96.9326	***
6	0.1163	0.0135	476.4708	3.0674	100.0000	*
TOTAL		0.4411	15533.366			

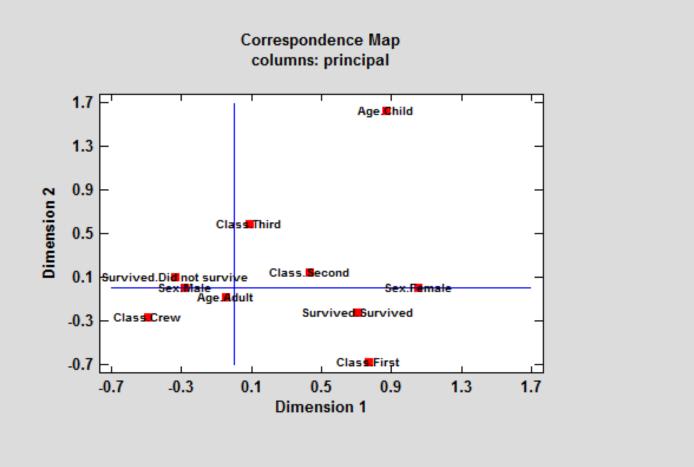


Scree Plot





Correspondence Map





Logistic Regression

- Binary logistic regression is designed to model situations in which there are only 2 possible outcomes, such as "survived" or "did not survive".
- It estimates a model that predicts the probability of those outcomes as a function of 1 or more predictor variables.
- The predictor variables may be either quantitative or categorical.



Bernoulli Events

- A Bernoulli event is an event in which 2 outcomes are possible.
- Let $Y_i = 1$ if ith person survived, 0 otherwise.
- The pmf for a Bernoulli event is

 $Prob(y) = p^{y}(1-p)^{1-y}$ where $0 \le p \le 1$



Predictor Variables

 We now want to let p be a function of one or more predictor variables X. In particular:

$$-X_1 = class$$

 $-X_2 = age$
 $-X_3 = sex$

Then:

$$p = f(X_1, X_2, X_3)$$



Logistic Model

- When modeling *p*, we need a model that remains between 0 and 1.
- Most common choice is the logistic model:

$$\log\left(\frac{p}{1-p}\right) = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \cdots$$

 It is a model for the log odds. If less than 0, odds are against the event. If greater than 0, odds are in favor of the event.



Data Input

Ŀ	ogistic Regression	
pclass survived name sex age sibsp parch ticket fare cabin embarked boat body home.dest	Dependent Variable: survived (Sample Sizes:) Quantitative Factors: age Categorical Factors: pclass sex (Select:) Deleta	
OK Cancel	Delete Transform Help	1. The set



Analysis Options

Method Maximum Likelihood C Weighted Least Squares	Fit All Variables Fit Fit Forward Sele C Backward S	ection	OK Cancel Exclude	
Model C First Order Second Order Include Constant	P-to-Enter: 0.01 Max. Steps: 50 Display	P-to-Remove:	Help	8° 0 0
	C All Steps	••••	*****	
S	tatgrap	hics		

Likelihood Ratio Test

Full second order model

Likelihood Ratio Tests

Factor	Chi-Square	Df	P-Value
age	9.7946	1	0.0017
pclass	11.6284	2	0.0030
sex	3.36405	1	0.0666
age^2	1.91558	1	0.1663
age*pclass	9.37229	2	0.0092
age*sex	3.89	1	0.0486
pclass*sex	38.6254	2	0.0000

Simplified model

Likelihood Ratio Tests							
Factor	Chi-Square	Df	P-Valu				
age	12.208	1	0.0005				
pclass	9.71294	2	0.0078				
sex	3.10055	1	0.0783				
age*pclass	10.0606	2	0.0065				
age*sex	4.33079	1	0.0374				

38.1605

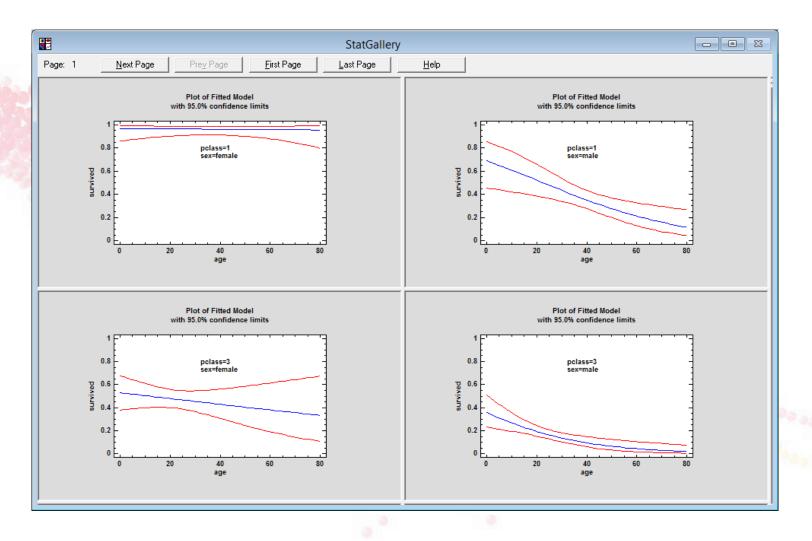
pclass*sex



0.0000

2

Plot of Fitted Model





Predictions

	C:\Data\webinar\titanic3.sgd							
	pclass	survived	name	sex	age 🔺			
-								
1303	3	0	Yousif, Mr. W	male				
1304	3	0	Yousseff, Mr.	male				
1305	3	0	Zabour, Miss.	female	14.5			
1306	3	0	Zabour, Miss.	female				
1307	3	0	Zakarian, Mr.	male	26.5			
1308	3	0	Zakarian, Mr.	male	27			
1309	3	0	Zimmerman, Mr	male	29			
1310	1		David	male	40			
$ \bullet \bullet \bullet \bullet$	titanic3 B C							

Predictions for survived

	name	Observed	Fitted	Lower 95.0%	Upper 95.0%
Row		Value	Value	Conf. Limit	Conf. Limit
1310	David		0.350919	0.276982	0.432779



Recommended Reading

- <u>Correspondence Analysis in Practice</u> (second edition) by Michael Greenacre (Chapman and Hall, 2007)
- <u>Applied Logistic Regression</u> (third edition) by David Hosmer and Stanley Lemeshow (Wiley, 2013)



Recorded Webinar

You may find the recorded webinar, PowerPoint slides and sample data at:

www.statgraphics.com

Look for "Instructional Videos".

