Data Mining Using Business Analytics: Data Descriptions

(for all editions: 3rd Edition, R, Python)

[Organized by alphabetical order of dataset name]

Accidents

These data, from the U.S. Bureau of Transportation Statistics, can be used to predict whether an accident will result in injuries or fatalities, based on predictors such as alcohol involvement, time of day, road condition, etc. Such a prediction system could be used to prioritize responder resources at the time of the report.

Source: US Dept. of Transportation, Bureau of Transportation Statistics, "TranStats," (www.transtats.bts.gov -- select "databases" then "General Estimate System (GES))

<a href="http://www.transtats.bts.gov/Fields.asp?Table_ID=1158&SYS_Table_Name=T_GES_ACCIDENT&User_Table_Name=Accident&Year_Info=1&First_Year=1999&Last_Year=2001&Rate_Info=1&Frequency=Annual&Data_Frequency=Annual,Monthly&Map_Info=&Is_Survey=1&Univ_Filter=&Latest_Available_Data=200_1_1

Note: TranStats reports both variables with missing data, and their derived counterparts with imputed values filled in, denoted by an "I" at the end. Only one variant (the original or the derived) is included here.

An "R" at the end of the variable name indicates that the Transtats variable has been collapsed into fewer categories for analysis purposes

Data are for the year 2001.

	Variables	
1	HOUR_I_R	1=rush hour, 0=not (rush = 6-9 am, 4-7 pm)
2	ALCOHOL_I	Alcohol involved = 1, not involved = 2
3	ALIGN_I	1 = straight, 2 = curve
4	STRATUM_R	1= NASS Crashes Involving At Least One Passenger Vehicle, i.e., A Passenger Car, Sport Utility Vehicle, Pickup Truck Or Van)
		Towed Due To Damage From The Crash Scene And No Medium
		Or Heavy Trucks Are Involved.
		0=not
5	WRK_ZONE	1= yes, 0= no
6	WKDY_I_R	1=weekday, 0=weekend
7	INT_HWY	Interstate? 1=yes, 0= no
8	LGTCON_I_R	Light conditions - 1=day, 2=dark (including dawn/dusk), 3=dark,
		but lighted,4=dawn or dusk
9	MAN_COL_I	0=no collision, 1=head-on, 2=other form of collision
10	PED_ACC_R	1=pedestrian/cyclist involved, 0=not

11	REL_JCT_I_R	1=accident at intersection/interchange, 0=not at intersection
12	REL_RWY_R	1=accident on roadway, 0=not on roadway
13	PROFIL_I_R	1= level, 0=other
14	SPD_LIM	Speed limit, miles per hour
15	SUR_CON	Surface conditions (1=dry, 2=wet, 3=snow/slush, 4=ice,
		5=sand/dirt/oil, 8=other, 9=unknown)
16	TRAF_CON_R	Traffic control device: 0=none, 1=signal, 2=other (sign, officer)
17	TRAF_WAY	1=two-way traffic, 2=divided hwy, 3=one-way road
18	VEH_INVL	Number of vehicles involved
19	WEATHER_R	1=no adverse conditions, 2=rain, snow or other adverse condition
20	NO_INJ_I	Number of injuries
21	PRPTYDMG_CRASH	H1=property damage, 2=no property damage
22	FATALITIES	1= yes, 0= no
23	MAX_SEV_IR	0=no injury, 1=non-fatal inj., 2=fatal inj.

AdSales

Hypothetical data about advertising expenditures in one time period and sales in a subsequent time period.

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Airfares

- 1. S_CODE: starting airport's code
- 2. S_CITY: starting city
- 3. E_CODE: ending airport's code
- 4. E CITY: ending city
- 5. COUPON: average number of coupons (a one-coupon flight is a non-stop flight, a two-coupon flight is a one stop flight, etc.) for that route
- 6. NEW: number of new carriers entering that route between Q3-96 and Q2-97
- 7. VACATION: whether a vacation route (Yes) or not (No); Florida and Las Vegas routes are generally considered vacation routes
- 8. SW: whether Southwest Airlines serves that route (Yes) or not (No)
- 9. HI: Herfindel Index measure of market concentration (refer to BMGT 681)
- 10. S_INCOME: starting city's average personal income
- 11. E_INCOME: ending city's average personal income
- 12. S_POP: starting city's population
- 13. E POP: ending city's population
- 14. SLOT: whether either endpoint airport is slot controlled or not; this is a measure of airport congestion
- 15. GATE: whether either endpoint airport has gate constraints or not; this is another measure of airport congestion

- 16. DISTANCE: distance between two endpoint airports in miles
- 17. PAX: number of passengers on that route during period of data collection
- 18. FARE: average fare on that route

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Amtrak

Ridership = Amtrak Ridership Number of Passengers (in thousands)

ApplianceShipments

Source:Data courtesy Ken Black

The series of quarterly shipments (in millions of dollars) of US household appliances between 1985 and 1989.

AustralianWines

Source: Website

Monthly Australian sales of wine Jan 1980 - Jul 1995

Bankruptcy

Source: "Predicting Corporate Bankruptcy"

Darden Business Publishing

Case authors Mark E. Haskins (HASKINSM@Darden.virginia.edu) and Phillip E. Pfeifer

(PFEIFERP@Darden.virginia.edu)

NO Arbitrary ID number for each firm.

D D=0 for failed firms, D=1 for healthy firms.

YR Year of Bankruptcy for failed firm in matched pair

R1 CASH/CURDEBT

R2 CASH/SALES

R3 CASH/ASSETS

- R4 CASH/DEBTS
- R5 CFF0/SALES
- R6 CFFO/ASSETS
- R7 CFFO/DEBTS
- R8 COGS/INV
- R9 CURASS/CURDEBT
- R10 CURASS/SALES
- R11 CURRASS/ASSETS
- R12 CURDEBT/DEBTS
- R13 INC/SALES
- R14 INC/ASSETS
- R15 INC/DEBTS
- R16 UBCDEP/SALES
- R17 INCDEP/ASSETS
- R18 INCDEP/DEBTS
- R19 SALES/REC
- R20 SALES/ASSETS
- R21 ASSETS/DEBTS
- R22 WCFO/SALES
- R23 WCFO/ASSETS
- R24 WCFO/DEBTS
- (c) 1988 University of Virginia Darden School Foundation

banks

Financial Condition 1 = financially weak

0 = financially strong

BathSoapHousehold

Demographic Data

MEM Member ID

SEC Socio economic class (1 = high, 4 = low)

- 1 A
- 2 B
- 3 C
- 4 D/E

FEH **Food Eating Habits** 1 Pure Vegetarian 2 Veg.But Serve Eggs 3 Non Vegetarian 0 Not Specified ΜT Native Language (mother tongue) 1 Assamese 2 Bengali 3 English 4 Gujarati 5 Hindi 6 Kannada 7 Kashmiri 8 Konkani 9 Malayalam 10 Marathi 11 Oriya 12 Punjabi 13 Rajasthani 14 Sindhi 15 Tamil Telugu 16 17 Urdu 18 Sanskrit 19 Other 0 Not Specified SEX Sex of homemaker 1 Male 2 Female AGE Age of homemaker 1 Up to 24 2 25-34 3 35-44 4 45+ EDU Education of homemaker 1 Illiterate 2

Literate, but no formal schooling

Up to 4 years of school

5-9 years of school

3

4

- 5 10-12 years of school
- 6 Some college
- 7 College graduate
- 8 Some graduate school
- 9 Graduate or professional school degree
- 0 Not specified
- HS Household size

Number of people in the household

CHILD Presence of children in household

- 1 Children up to age 6 present (only)
- 2 Children 7-14 present (only)
- 3 Both
- 4 None
- 5 Not specified
- CS Television
- 1 Cable or broadcast TV available
- 2 Unavailable

Affluence Index

Calculated from **Durables** sheet.

Purchase Summary Data Labels What they stand for

No. Brands Number of brands purchased

Brand Runs Number of runs (streaks) of purchasing the same brand

Total volume Volume of product purchased (grams)

No. of trans. Number of transactions

Value in paise (100 paise = 1 rupee)

Avg. Price Avg. price (rupees per 100 gram cake); computed from total volume and

value

Purch. Vol. no promo Percent of volume purchased not on promotion

Purch Vol. promo 6 Percent of volume purchased on promo code 6

Purch. Vol other promo Percent of volume purchased on promo code other than 6

Brand Codelist (click <u>here</u>)

Price	Codelist
1 1100	Outlist

- 2 ANY POPULAR SOAP
- 3 ANY ECONOMY/CARBOLIC
- 4 ANY SUB-POPULAR

Promotion Codelist

- 1 Price off
- 2 Exchange Offer
- 3 Coupons
- 4 Extra grammage
- 5 Value added Pack
- 6 Banded Offer
- 7 Free gift
- 8 Others

Proposition Codelist

- 5 ANY BEAUTY
- 6 ANY HEALTH
- 7 ANY HERBAL
- 8 ANY FRESHNESS
- 9 ANY HAIR
- 10 ANY SKIN CARE
- 11 ANY FAIRNESS
- 12 ANY BABY
- 13 ANY GLYCERINE
- 14 ANY CARBOLIC
- 15 ANY OTHERS

Durable Ownership

Code	Durables	Affluence Weights
1	Radio/Transistor with FM	1
2	Radio/Transistor without FM	1
3	Stereo/Mono Tape Recorder	1
4	Two-in-one	2
5	Hi-Fi System/Music System	
	without Compact disk	3

6	Hi-Fi System/Music System	
	with Comapct disk	4
7	Walkman with FM	2
8	Walkman without FM	2
9	Discman with FM	3
10	Discman without FM	3
11	Video (VCP/VCR)	3
12	Laser Discs VCD/LD/DVD	5
13	TV - Black & White	2
14	Colour TV with remote	3
15	Colour TV without remote	3
16	Bicycle	1
17	Moped	2
18	Motorcycle	8
19	Scooter	5
20	Electric/Immersion Water heater	1
21	LPG/Bio-Gas stove	1
22	Mixer/Grinder	2
23	Pressure Cooker	1
24	Toaster	1
25	Cooking Range	4
26	Refrigerator - Non Frost free	3
27	Refrigerator - Frost free	5
28	Automatic dish washer	6
29	Oven - Electric	4
30	Electric Pressure Cooker	2
31	Microwave Oven	5
32	Rice Cooker	2
33	Electric Irons	1
34	Geyser	1
35	Cameras (still)	2
36	Telephones (with NSD/STD/ISD)	3
37	Telephones (Local only)	2
38	"Air Coolers"	2
39	Vacuum cleaner	2
40	Air Conditioners	5
41	Water purifier (Aquaguard etc.)	1
42	Washing Machines (Rs.5000+)	
	Semi Automatic	4
43	Washing Machines (Rs.5000+)	
	Fully Automatic	5
44	Washing Machines (Rs.5000+)	
	Front Loading	6

45	Washing Machines (Rs.5000+)	
	Top Loading	5
46	Mobil/Cellular phone	4
47	Pager	2
48	Personal/Home Computers	8
49	Computer Printers	6
50	Fax Machine	6
51	Video camera/Handycam	6
52	Radio Clock	2
53	Deep Freezer	5
54	Electirc Kettle	1
55	Dish Washing Machine	5
56	Kitchen Sink	1
57	Floor Polisher	1
58	Cars/Jeeps/Vans	8
59	Auto Rickshaw	3
60	Tractors	5
61	Oven-In Built Range	5
62	Oven Ordinary Box (Gas)	3
63	Electric Table Fan	1
64	Electric Ceiling Fan	1
65	Torch	1
66	Sewing Machine	2
67	Generator	5
68	Pump Set/Water Pump	5

Not used:

Product Codelist

- 02 Toilet Soaps
- 05 Tooth Paste/Powder
- 01 Washing Soaps/Detergents
- 21 Washing Powder
- 45 Skin Creams
- 20 Edible Oils/Ghee/Vanaspati

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Bicup2006

Source: Oct. 2006 public business intelligence competition

http://www.tis.cl/2007/futurosTalleres/_2006/Taller_1/BICUP2006-ENGLISH/

Data are the number of customers appearing at a bus terminal during 15 minute periods beginning at the specified time periods

Book Purchases

Columns indicate book categories, cells indicate whether a book in that category was purchased.

BostonHousing

This dataset contains information collected by the US Census Service concerning housing in the area of Boston Massachusetts. It was obtained from the StatLib archive (http://lib.stat.cmu.edu/datasets/boston). The dataset has 506 cases.

Source: The data was originally published by Harrison, D. and Rubinfeld, D.L. 'Hedonic prices and the demand for clean air', J. Environ. Economics & Management, vol.5, 81-102, 1978.

There are 14 attributes in each case of the dataset. They are:

CRIM per capita crime rate by town

ZN proportion of residential land zoned for lots over 25,000 sq.ft.

INDUS proportion of non-retail business acres per town.

CHAS Charles River dummy variable (1 if tract bounds river; 0 otherwise)

NOX nitric oxides concentration (parts per 10 million)

RM average number of rooms per dwelling

AGE proportion of owner-occupied units built prior to 1940
DIS weighted distances to five Boston employment centres

RAD index of accessibility to radial highways TAX full-value property-tax rate per \$10,000

PTRATIO pupil-teacher ratio by town

LSTAT % lower status of the population

MEDV Median value of owner-occupied homes in \$1000

CanadianWorkHours

average annual number of weekly hours spent by Canadian manufacturing workers

Source: Ken Black (used by permission)

CatalogCrossSell

Multi-Division Catalog Company

Scenario - A random sample of customers is shown in the Data sheet. A "1" indicates a purchase has been made from a catalog in that division, a "0" indicates no purchase.

Source: Adapted from a set of cases provided for educational purposes by the Direct Marketing Education Foundation; used with permission.

Cereals

Source: DATA ANALYSIS FOR STUDENT LEARNING (DASL)

1. Name: Name of cereal

- 2. mfr: Manufacturer of cereal where A = American Home Food Products; G = General Mills; K = Kelloggs; N = Nabisco; P = Post; Q = Quaker Oats; R = Ralston Purina
- 3. type: cold or hot
- 4. calories: calories per serving
- 5. protein: grams of protein
- 6. fat: grams of fat
- 7. sodium: milligrams of sodium
- 8. fiber: grams of dietary fiber
- 9. carbo: grams of complex carbohydrates
- 10. sugars: grams of sugars
- 11. potass: milligrams of potassium
- 12. vitamins: vitamins and minerals 0, 25, or 100, indicating the typical percentage of FDA recommended
- 13. shelf: display shelf (1, 2, or 3, counting from the floor)
- 14. weight: weight in ounces of one serving
- 15. cups: number of cups in one serving
- 16. rating: a rating of the cereals calculated by Consumer Reports

CharlesBookClub

Source: Adapted with permission from The Bookbinders Club, prepared by Nissan Levin and Jacob Zahavi.

Variable Description

Seq# Sequence number in the sample

ID# in the full dataset
Gender 0=male, 1=female

M Monetary - total money spent on books
R Recency - Months since last purchase
F Frequency - Total number of purchases

FirstPurch Months since first purchase

Col H - R book categories

Related Purchase Number of related books purchased

Mcode, Rcode, Fcode Recoding of M, R and F - see case description in DMBA

Cosmetics

Source: Statistics.com

A drug store chain wants to learn more about cosmetics buyers purchase patterns. Specifically, they want to know what items are purchased in conjunction with each other, for purposes of display, point of sale special offers, and to eventually implement a real time recommender system to cross-sell items at time of purchase.

The data (synthetic) are in the form of a matrix in which each column represents a product group, and each row a customer transaction.

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Cosmetics-small

Source: Statistics.com

A drug store chain wants to learn more about cosmetics buyers purchase patterns. Specifically, they want to know what items are purchased in conjunction with each other, for purposes of display, point of sale special offers, and to eventually implement a real time recommender system to cross-sell items at time of purchase.

The data are in the form of a matrix in which each column represents a product group, and each row a customer transaction.

Note: Data are from Peter Bruce, partially drawn from a real source unrelated to cosmetics and partially generated.

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Courserating

Source: Statistics.com

Student ratings of online statistics courses at Statistics.com

© 2016 Statistics.com

Coursetopics

Source: Statistics.com

Course topics at statistics.com (each row is a customer, column heads are topics taken [1] or not taken

[0] by that customer)

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DepartmentStoreSales

Data on the quarterly sales for a department store over a 6-year period. Source = Chris Albright, used with permission

© 2016 Statistics.com

drug

EastWestAirlines or EastWestAirlinesCluster

East-West Airlines is trying to learn more about its customers. Key issues are their flying patterns, earning and use of frequent flyer rewards, and use of the airline credit card. The task is to identify customer segments via clustering.

Source: Based upon real business data; company names have been changed.

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Field Name	Data Type	Max Data Length	Raw Dat Telcom (Field?	
ID#	NUMBER		Telcom	Unique ID
Balance	NUMBER	8	Raw	Number of miles eligible for award travel
Qual_miles	NUMBER	8	Raw	Number of miles counted as qualifying for Topflight status
cc1_miles	CHAR	1	Raw	Number of miles earned with freq. flyer credit card in the past 12 months:
cc2_miles	CHAR	1	Raw	Number of miles earned with Rewards credit card in the past 12 months:
cc3 miles	CHAR	1	Raw	Number of miles earned with Small Business credit card in the past 12 months:
note: miles bins:				1 = under 5,000
				0 5000 40000

2 = 5,000 - 10,000

3 = 10,001 - 25,000

4 = 25,001 - 50,000

5 = over 5	0	.0	0	C
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Bonus_miles	NUMBER	Raw	Number of miles earned from non-flight bonus transactions in the past
			12 months
Bonus_trans	NUMBER	Raw	Number of non-flight bonus transactions in the past 12 months
Flight_miles_12md	NUMBER	Raw	Number of flight miles in the past 12 months
Flight_trans_12	NUMBER	Raw	Number of flight transactions in the past 12 months
Days_since_enroll	NUMBER	Telcom	Number of days since Enroll_date
Award? NUMBEI	R Telcor	n	Dummy variable for Last_award (1=not null, 0=null)

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EastWestAirlineNN

East-West Airlines has entered into a partnership with the wireless phone company Telcon to sell the latter's service via direct mail. These are a sample of data, provided so that the analyst can develop a model to classify East-West customers as to whether they purchase a wireless phone service contract (target variable Phone_sale).

Source: Based upon a real business case and real data; company names have been changed. © 2016 Galit Shmueli and Peter Bruce

Field Name	Data Type	Max Data Length	Raw Data Telcom C Field?		Description
ID#		NUMBER		Telcor	m Unique ID
Balance		NUMBER	8	Raw	Number of miles eligible for award travel
Qual_miles		NUMBER	8	Raw	Number of miles counted as qualifying for Topflight status
cc1_miles		CHAR	1	Raw	Number of miles earned with freq. flyer credit card in the past 12 months:
cc2_miles		CHAR	1	Raw	Number of miles earned with Rewards credit card in the past 12 months:
cc3_miles		CHAR	1	Raw	Number of miles earned with Small Business credit card in the past 12 months:
note: miles bins:					1 = under 5,000
					2 = 5,000 - 10,000
					3 = 10,001 - 25,000
					4 = 25,001 - 50,000
					5 = over 50,000
Bonus_miles		NUMBER		Raw	Number of miles earned from non-flight bonus transactions in the
					past 12 months
Bonus_trans		NUMBER		Raw	Number of non-flight bonus transactions in the past 12 months
Flight_miles_12mo)	NUMBER		Raw	Number of flight miles in the past 12 months
Flight_trans_12		NUMBER		Raw	Number of flight transactions in the past 12 months
Email		CHAR	1	Raw	E-mail address on file. 1= yes, 0 =no?
Club_member		NUMBER		Telcor	m Member of the airline's club (paid membership), 1=yes, 0=no
Any_cc_miles_12n	no	NUMBER		Telcor	Dummy variable indicating whether member added miles on any credit card type within the past 12 months (1='Y', 0='N')
Phone_sale		NUMBER		Telcor	Dummy variable indicating whether member purchased Telcom service as a result of the direct mail campaign (1=sale, 0=no sale)

eBayAuctions

Source: Compiled from eBay.com for the period May-June 2004.

Variable descriptions

Category: Category of the auctioned item.

currancy:

sellerRating: a rating by eBay, as a function of the number of "good" and "bad"

transactions the seller had on eBay.

Duration: Number of days the auction lasted (set by seller at auction start)

endDay: Day of week that the auction closed ClosePrice: Price item sold at (converted into USD)

OpenPrice: Initial price set by the seller (converted into USD)

Competitive?: whether the auction had a single bid (0) or more (1)

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EuropeanJobs

Data labels

1. Country: Name of country

2. Agr: Percentage employed in agriculture
3. Min: Percentage employed in mining
4. Man: Percentage employed in manufacturing

5. PS: Percentage employed in power supply industries

6. Con: Percentage employed in construction

7. SI: Percentage employed in service industries

8. Fin: Percentage employed in finance

9. SPS: Percentage employed in social and personal services10. TC: Percentage employed in transport and communications

Faceplate

Synthetic Data on Purchases of Phone Faceplates.

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Farm-ads

Data on advertisements posted at a website that caters to the needs of a specific farming community. Each ad is in a row, and each ad labeled as either -1 (not relevant) or 1 (relevant). The goal is to develop a predictive model that can classify ads automatically.

fiftytransactions

A small database of 50 transactions, where each of the nine items is assigned randomly to each transaction.

FlightDelays

Source: Bureau of Transportation Statistics

Variable explanations are in comments appended to column heads.

Note the data has both scheduled and actual departure time - pay attention to which you use!

All flights out of 3 DC airports (WAS) into 3 NYC airports not cancelled flights in January 2004

Data labels:

CRS_DEP_TIME scheduled departure time

CARRIER The airline

DEP_TIME Actual departure time

DEST Destination airport in NY: Kennedy (JFK), LaGuardia (LGA), Newark (EWR)

DISTANCE Flight distance in miles

FL_DATE Flight date
FL_NUM Flight number

ORIGIN Departure airport in Washingon DC: National (DCA), Baltimore-Washington (BWI),

Dulles (IAD)

Weather Whether the weather was inclement (1) or not (0)

DAY WEEK Day of week. 1=Mon, 2=Tues...

DAY_OF_MONTH

TAIL_NUM This number is airplane specific

Flight Status Whether the flight was delayed or on time (defined as arriving within 15 min of

scheduled time)

Carrier Code Carrier Name

AA American Airlines, Inc.
CO Continental Air Lines, Inc.
DH Atlantic Coast Airlines
DL Delta Air Lines, Inc.

EV Atlantic Southeast Airlines
FL Airtran Airways Corporation
MQ American Eagle Airlines,inc

OH Comair, Inc.

RU Continental Express Airline

UA United Air Lines, Inc. US US Airways, Inc.

Fundraising

ZIP: Zipcode group (zipcodes were grouped into 5 groups; only 4 are needed for analysis since if a potential donor falls into none of the four he or she must be in the other group. Inclusion of all five variables would be redundant and cause some modeling techniques to fail. A "1" indicates the potential donor belongs to this zip group.)

00000-19999 => 1 (omitted for above reason)

20000-39999 => zipconvert_2 40000-59999 => zipconvert_3 60000-79999 => zipconvert_4 80000-99999 => zipconvert_5

HOMEOWNER 1 = homeowner, 0 = not a homeowner

NUMCHLD Number of children

INCOME Household income

GENDER Gender: 0 = Male 1 = Female

WEALTH Wealth Rating (Wealth rating uses median family income and population statistics

from each area to index relative wealth within each state. The segments are denoted 0-9, with 9 being the highest wealth group and zero being the lowest.

Each rating has a different meaning within each state.)

HV Average Home Value in potential donor's neighborhood in \$ hundreds ICmed Median Family Income in potential donor's neighborhood in \$ hundreds

ICavg Average Family Income in potential donor's neighborhood in hundreds IC15

Percent earning less than 15K in potential donor's neighborhood

NUMPROM Lifetime number of promotions received to date

RAMNTALL Dollar amount of lifetime gifts to date MAXRAMNT Dollar amount of largest gift to date LASTGIFT Dollar amount of most recent gift

TOTALMONTHS Number of months from last donation to July 1998 (the last time the case was

updated)

Number of months between first and second gift TIMELAG

AVGGIFT Average dollar amount of gifts to date

TARGET B

1 = Donor 0 = Non-donor

TARGET_D Target Variable: Donation Amount (in \$). We will NOT use it.

gdp

DATA FROM VEENHOVEN'S WORLD DATABASE OF HAPPINESS.

http://data.worldbank.org/indicator/NY.GDP.MKTP.CD

World Development Indicators.

Gross domestic product of the countries.

GermanCredit

Codelist (available in the textbook)

Variable Name	Description	Variable Type	Code Description
OBS# CHK_ACCT	Observation No. Checking account status	Categorical Categorical	0: < 0 DM 1: 0 << 200 DM
			2: => 200 DM 3: no checking account
DURATION	Duration of credit in mon	ths Numerical	
HISTORY	Credit history	Categorical	0: no credits taken1: all credits at this bank paid back duly2: existing credits paid back duly till now3: delay in paying off in the past4: critical account
NEW_CAR USED_CAR FURNITURE RADIO/TV	Purpose of credit Purpose of credit Purpose of credit Purpose of credit	Binary Binary Binary Binary	<pre>car (new) 0: No, 1: Yes car (used) 0: No, 1: Yes furniture/equipment 0: No, 1: Yes radio/television 0: No, 1: Yes</pre>

EDUCATION	Purpose of credit	Binary	education 0: No, 1: Yes			
RETRAINING	Purpose of credit	Binary	retraining 0: No, 1: Yes			
AMOUNT	Credit amount	Numerio	cal			
SAV_ACCT	Average balance in	Categorical 0 : < 100 DM				
	savings account		1 : 100<= < 500 DM			
			2 : 500<= < 1000 DM			
			3 : =>1000 DM			
			4 : unknown/ no savings account			
EMPLOYMENT	Present employment since	Categorical	0 : unemployed			
			1 : < 1 year			
			2 : 1 <= < 4 years			
			3 : 4 <= < 7 years			
			4 : >= 7 years			
INSTALL_RATE	Installment rate as % of					
	disposable income	Numerical				
MALE_DIV	Applicant is male and divorce	d Binary	0: No, 1: Yes			
MALE_SINGLE	Applicant is male and single	Binary	0: No, 1: Yes			
MALE_MAR_WID	Applicant is male and married					
	or a widower	Binary	0: No, 1: Yes			
CO-APPLICANT	Application has a co-applican	t Binary	0: No, 1: Yes			
GUARANTOR	Applicant has a guarantor	Binary	0: No, 1: Yes			
PRESENT_RESIDENT	Present resident since-years	Categorica	1 0: <= 1 year			
			1<<=2 years			
			2<<=3 years			
			3:>4years			
REAL_ESTATE	Applicant owns real estate	Binary	0: No, 1: Yes			
PROP_UNKN_NONE	Applicant owns no property					
	(or unknown)	Binary	0: No, 1: Yes			
AGE	Age in years	Numerical				
OTHER_INSTALL	Applicant has other					
	installment plan credit	Binary	0: No, 1: Yes			
RENT	Applicant rents	Binary	0: No, 1: Yes			
OWN_RES	Applicant owns residence	Binary	0: No, 1: Yes			
NUM_CREDITS	Number of existing credits					
	at this bank	Numerical				
JOB	Nature of job	Categorical	0: unemployed/ unskilled -			
			non-resident			
			1: unskilled - resident			
			2: skilled employee / official			
1.6			3: management/			
self-employed/hi	duīh		1.6.			
DEDENIDENES			qualified employee/ officer			
NUM_DEPENDENTS	Number of people for whom	27				
met editore	liable to provide maintenance	Numerical				
TELEPHONE	Applicant has phone in his	Dinam-	0. No. 1. Voc			
FORTON	or her name	Binary				
FOREIGN	Foreign worker		0: No, 1: Yes			
RESPONSE	Credit rating is good	DIIId	0: No, 1: Yes			

Hair-Care-Product

Fictional data representing an uplift study. A promotion for a hair color product was sent out to a sample of potential customers.

Promotional literature about a hair care product was sent to members of a buyers club. The goal is to determine which groups are most likely to make increased purchases as a result of receiving the promotion.

Source: SAS Institute, used by permission.

Worksheets:

Hair Care Product_original - This worksheet contains original hair care product data of size 1,26,184. **Hair Care Product_sample** - This worksheet contains a sample dataset of size 10,000, sampled (without replacement) from the original dataset of size 1,26,184.

Data_for_analysis - This worksheet contains the sample dataset of size 10,000, but with variables Promotion(Yes/No), Gender(Male/Female) and Residence(Urban/Rural) recoded as Promotion(1/0), Gender(1/0) and Residence(1/0) respectively.

LaptopSales

Date purchase date

Configuration A numerical code representing a combination of screen size, battery life, RAM, etc.

Each code corresponds to a particular combination.

Customer Postcode postcode in London of the customer Store Postcode postcode in London of the store

Retail Price price of laptop in GBP

Screen Size screen size of laptop (Inches)
Battery Life battery life of laptop (Hours)
RAM RAM size of laptop(GB)

Processor Speeds processor speed of laptop (GHz)

Integrated Wireless? whether the laptop has integrated wireless or not

HD Size HD size of laptop (GB)

Bundled Applications? whether the laptop comes with bundled applications or not

customer X X geo coordinates for customer location.

customer Y Y geo coordinates for customer location.

store X X geo coordinates for store location

Y geo coordinates for store location

Y geo coordinates for store location

LaptopSalesJanuary2008

This is a subset of the Laptop sales dataset. It includes only the Jan 2008 sales (the complete dataset includes the entire 2008 sales).

Source: The laptop sales data were part of the ENBIS 2009 Challenge in Industrial Statistics

MortgageDefaulters

This data set contains data on mortgages that have been approved by bank underwriters.

Variable Explanation
Bo_Age Borrower age
Ln Orig Value of loan, USD

Orig_LTV_Ratio_Pct Ratio of loan to home purchase price

Credit_score Borrower's credit score

First_home First time home buyer? (Y/N)

Tot_mthly_debt_exp Borrower's total monthly debt expense

Tot_mthly_incm Borrower's total monthly income

pur_prc_amt Purchase price for house

DTI_ratio Borrower debt to income ratio (Tot_mthly_debt_exp/Tot_mthly_incm)

Status Current loan status

OUTCOME Binary version of "Status" (either default or non-default)

State US state in which home is located

Median state inc Median household income by state 2002-2004

UPB>Appraisal Loan amount (Ln_Orig) greater than appraisal (orig_apprd_val_amt) 0-no, 1=yes

Note that some of the above variables were derived from combinations of two others.

Pharmaceuticals

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Source: compiled from various web sources

RidingMowers

Source: Data courtesy of Dean Wichern.

Income: Annual income in \$000 Lot Size: In thousands of sq. feet

Ownership: Whether the resident owns a riding mower or not

Sept11Travel

Source: Bureau of Transportation Statistics - https://goo.gl/w2lJPV

AirRMP Air revenue passenger miles (1 RMP is one revenue passenger carried for one mile)

RailPM Rail passenger miles VMT Vehicle miles traveled

ShampooSales

Data on the monthly sales of a certain shampoo over a 3-year period.

Source: Time Series Data Library, http://data.is/TSDLdemo

SouvenirSales

Monthly sales for a souvenir shop at a beach resort town in Queensland, Australia, between 1995–2001. Source: Time Series Data Library, http://data.is/TSDLdemo

SP500

Close=Monthly closing prices of S&P500

Spambase

Source: UCI Machine Learning Repository, HP database of emails

Each of the words below are columns in the data and the values represent % of words in the e-mail that match that particular word. For example, make represent % of words in the e-mail that match "make".

make address all W_3d our over remove internet order mail receive will people report addresses free business email you credit your font W_000 money hp hpl george W_650 lab labs telnet W_857 data W_415 W_85 technology W_1999 parts pm direct cs meeting original project re: edu table conference C; C(C[C! C\$ C#

CAP_avg - average length of uninterrupted sequences of capital letters
CAP_long - length of longest uninterrupted sequence of capital letters

CAP_tot - total number of capital letters in the e-mail

Spam -1 = spam, 0 = not spam

SystemAdministrators

Source: Samprit Chatterjee

Variables

Experience - measures months of full-time system administrator experience

Training - measures the number of relevant training credits

Completed task - either Yes or No, according to whether or not the administrator

completed the tasks

Taxi-cancellation-case

The data are a randomly selected subset of the original data, with 10,000 rows, one row for each booking of a taxi. There are 17 input variables, including user (customer) ID, vehicle model, whether the booking was made online or via a mobile app, type of travel, type of booking package, geographic information, and the date and time of the scheduled trip. The target variable of interest is the binary indicator of whether a ride was canceled.

tinydata

Data includes information on a tasting score for a certain processed cheese. The two predictors are scores for fat and salt, indicating the relative presence of fat and salt in the particular cheese sample (where 0 is the minimum amount possible in the manufacturing process, and 1 the maximum). The outcome variable is the cheese sample's consumer taste preference, where like or dislike indicate whether the consumer likes the cheese or not.

Tayko

Codelist

Var. # 1.	Variable Name US	Description Is it a US address?		Variable Ty binary	/pe	Code D	•
2 - 16	Source_*	Source catalog for the record		binary		1: yes	0: no
		(15 identified sources plus one "c	othe	r source" ca	atego	ry; 15 d	ummies
		created with "other" as the refere	nce	, hence omi	itted.)	
17.	Freq.	Number of transactions in last					
		year at source catalog		numeric			
18.	last_update_days_ago	How many days ago was last					
		update to cust. record		numeric			
19.	1st_update_days_ago	How many days ago was 1st					
		update to cust. record		numeric			
20.	Web_order	Customer placed at least					
		1 order via web		binary		1: yes	0: no
21.	Gender=mal	Customer is male bir	nary	1:	yes	0: no	
22.	Address_is_res	Address is a residence		binary		1: yes	0: no
23.	Purchase	Person made purchase in					
		test mailing		binary		1: yes	0: no
24.	Spending	Amount spent by customer in					
		test mailing (\$)		numeric			

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Textiles

Codelist

ID	case number	
SILKWT	silk weight	
ZARIWT	zari weight	
SILKWT_Cat		categorical version of SILKWT
ZARIWT_Cat		categorical version of ZARIWT
BODYCOL	body color	
BODYCOL_*	body color	series of binary variables, 1 = body is that color
BRDCOL	border color	
BRDCOL_*	border color	series of binary variables, 1 = border is that color
BODYSHD_*	body shade	1 = pale, 4 = bright
BRDSHD	border shade	
BRDSHD_*	border shade	1 = pale, 4 = bright
SARSIDE	1 or 2 sided sari	1 = 1-sided, 2 = 2-sided
BODYDES	body design	
BODYDES_*	body design	series of binary variables, 1 = body is that design
BRDDES	border design	
BRDDES_*	border design	series of binary variables, 1 = border is that design
PALDES	pallav design	
PALDES_*	pallav design	series of binary variables, 1 = border is that design
BRDSZ	border size	
PALSZ	pallav size	
SALE	1 = sale, 0 = no sale	

Note: The colors and designs selected for the binary variables were those that were most common.

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ToyotaCorolla

Variable Description Id Record_ID

Model Model Description
Price Offer Price in EUROs

Age_08_04 Age in months as in August 2004 Mfg_Month Manufacturing month (1-12)

Mfg_Year Manufacturing Year

KM Accumulated Kilometers on odometer

Fuel_Type Fuel Type (Petrol, Diesel, CNG)

HP Horse Power

Met_Color Metallic Color? (Yes=1, No=0)

Color (Blue, Red, Grey, Silver, Black, etc.)

Automatic ((Yes=1, No=0)

CC Cylinder Volume in cubic centimeters

Doors Number of doors
Cylinders Number of cylinders
Gears Number of gear positions
Quarterly_Tax Quarterly road tax in EUROs

Weight Weight in Kilograms

Mfr_Guarantee Within Manufacturer's Guarantee period (Yes=1, No=0)
BOVAG_Guarantee BOVAG (Dutch dealer network) Guarantee (Yes=1, No=0)

Guarantee_Period Guarantee period in months

ABS Anti-Lock Brake System (Yes=1, No=0)

Airbag_1 Driver_Airbag (Yes=1, No=0)
Airbag_2 Passenger Airbag (Yes=1, No=0)
Airco Airconditioning (Yes=1, No=0)

Automatic airco Automatic Airconditioning (Yes=1, No=0)

Boardcomputer (Yes=1, No=0)

CD_Player CD Player (Yes=1, No=0)

Central_Lock Central Lock (Yes=1, No=0)

Powered_Windows Powered Windows (Yes=1, No=0)

Power_Steering Power Steering (Yes=1, No=0)

Radio (Yes=1, No=0)

Mistlamps (Yes=1, No=0)

Sport_Model (Yes=1, No=0)

Backseat_Divider (Yes=1, No=0)

Metallic Rim (Yes=1, No=0)

Metallic Rim (Yes=1, No=0)

Radio cassette Radio Cassette (Yes=1, No=0)

Parking Assistant Parking assistance system (Yes=1, No=0)

Tow_Bar Tow Bar (Yes=1, No=0)

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ToysRUsRevenues

The quarterly revenues of Toys "R" Us between 1992 and 1995

Source: Chris Albright

UniversalBank

Courtesy - Statistics.com

Data Description:

ID Customer ID

Age Customer's age in completed years

Experience #years of professional experience
Income Annual income of the customer (\$000)

ZIPCode Home Address ZIP code.
Family Family size of the customer

CCAvg Avg. spending on credit cards per month (\$000)

Education Education Level. 1: Undergrad; 2: Graduate; 3: Advanced/Professional

Mortgage Value of house mortgage if any. (\$000)

Personal Loan Did this customer accept the personal loan offered in the last campaign?

Securities Account Does the customer have a securities account with the bank?

CD Account Does the customer have a certificate of deposit (CD) account with the bank?

Online Does the customer use internet banking facilities?

CreditCard Does the customer use a credit card issued by UniversalBank?

Note: Data are synthetic © Cytel, Inc. 2005

Universities

The dataset on American college and university rankings (available from www.dataminingbook.com) contains information on 1302 American colleges and universities offering an undergraduate program. For each university, there are 17 measurements that include continuous measurements (such as tuition and graduation rate) and categorical measurements (such as location by state and whether it is a private or a public school).

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Source: Compiled from US News and World Report rankings on 1302 American Colleges and Universities

Utilities

Variable Description
Company Company name

Fixed_charge Fixed-charge coverage ratio (income/debt)

RoR Percent rate of return on capital Cost Cost per KW capacity in place

Load factor Annual load factor

Demand_growth Percent demand growth
Sales (KWH use per year)

Nuclear Percent nuclear

Fuel_Cost Total fuel costs (cents per KWH)

Veerhoven

Data measuring happiness of countries. according to a 2006 Gallup survey.

Voter-Persuasion

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Source: Ken Strasma and HaystaqDNA

See separate <u>dictionary sheet</u> for variable descriptions.

These data and this method are used in the Uplift Case in the Cases chapter.

WalMartStock

The series of Walmart daily closing prices between February 2001 and February 2002.publicly available, for example, at http://finance.yahoo.com.

These data are also used in "Data Analysis for Managers" by Albright, Winston & Zappe.

West Roxbury

Variable Description

TOTAL VALUE Total assessed value for property, in thousands of USD

TAX Tax bill amount based on total assessed value multiplied by the tax rate

LOT SQFT Total lot size of parcel in square feet

YR BUILT Year property was built

GROSS AREA Gross floor area

LIVING AREA Total living area for residential properties (ft2)

FLOORS Number of floors

ROOMS Total number of rooms
BEDROOMS Total number of bedrooms
FULL BATH Total number of full baths
HALF BATH Total number of half baths

KITCHEN Total number of kitchens

FIREPLACE Total number of fireplaces

REMODEL When house was remodeled (Recent/Old/None)

Wine

Wine dataset contains properties of wine captured from three different wineries in the same region. There are 13 variables describing various properties of wine and 3 classes. This dataset can be used for classification with Type as a output variable OR can be used to perform clustering to without using Type variable to see the accuracy of prediction.

This data set can be found in the UCI Machine Learning Repository (http://www.ics.uci.edu/~mlearn/MLSummary.html or ftp://ftp.ics.uci.edu/pub/machine-learning-databases/wine/)