

Like the time of compiler programming and Master programming List of things to be done on paper and cross out				
General SAS: best resource: online search	Format (right click column + attribute): a name; a w value, which specifies the width displaying; a period following the w value. Numeric SAS formats, such as the DOLLARw.d format:d value, which is the number of decimal places t		SAS Statement is free format: 1.can begin and end anywhere on a line 2.one statement can continue over several lines 3. several statements can be on a line. 4. SAS statements are not case sensitive. 5. Blanks or special characters separate the "words" in a SAS statement.	
31 st June 2009				
Any source/platform (data)→ management → analysis → presentation (any format)	table properties: Column (name, type, length, label)		data test; /*10252012*/ set Ts102512.productsales; run; procprintdata=test; run;	
Management: 1. Format 2. create variables (columns) 3. use operators to evaluate data values 4. use functions to create and recode data 5. subset data 6. perform conditional processing 7. merge a wide range of data sources 8. create, retrieve, and update database information.	An informat (input format) is the instruction that specifies how SAS reads raw data; the same as Format but start with \$		Defining one level name vs. multi level names	
Analysis: 1. produce tables, frequency counts, and cross-tabulation tables 2. create a variety of charts and plots 3. compute a variety of descriptive statistics, including the mean, sum, variance, standard deviation, etc. 4. compute correlations and other measures of association, multi-way cross-tabulations and inferential statistics.	SAS procedures help in: 1. manipulate data 2. store and retrieve information 3. perform statistical analysis 4. create reports.		data original; input x1-x5; cards; 9 8 7 8 . 8 7 6 . 9 . . 9 7 6 run; procprint; title'Original Data'; run; data modified; set original; array zero x1-x5; doover zero; if zero=.then zero=0; end; procprint; Title'Data modified with array and do loop'; run;	
	SAS program two steps: a DATA step and a PROC step .		* Create a SAS data set named distance; * Convert miles to kilometers; DATA distance; Miles = 26.22; Kilometers = 1.61 * Miles; RUN; * Print the results; PROCPRINTDATA = distance; RUN;	
output formats: 1. ML: HTML4 and XML 2. output that is formatted for a high-resolution printer, such PostScript, PDF, and PCL files, RTF 3. color graphs that you can make interactive using ActiveX controls or Java applets. 4. output these reports to a wide variety of locations and platforms	DATA steps typically create or modify SAS data sets, but they can also be used to produce custom-designed reports (e.g. put data in, compute value of new variable, check and correct data errors, produce new data by: subsetting, merging, and updating).		Import the data from excel/access/...	
Result window: You can view, save, and print individual items of output. (Recall that the Results Viewer window is the where you actually view HTML output.)	data clinic.admit2; set clinic.admit; run; creates a new SAS data set ADMIT2 in the CLINIC library using the existing SAS data set ADMIT in the CLINIC library.		* Read internal data into SAS data set uspresidents; DATA uspresidents; INPUT President \$ Party \$ Number; DATALINES; Adams F 2 Lincoln R 16 Grant R 18 Kennedy D 35 ; RUN;	
SAS data set (table): descriptor (info related data values): organized as table of observation (rows) and variables (columns)	SAS procedure (step): data sets, such as listing, sorting, and summarizing data 1.print a report. 2. produce descriptive statistics. 3. create a tabular report 4. produce plots and charts.		* Read data from external file into SAS data set; DATA presidentt; INFILE "c:\CDBookSurvey\president.dat"; INPUT President \$ Party \$ Number; RUN;	
SAS library, where file such as SAS data sets & catalogs stored: Files in the directory; To access a library, you assign it a name (also known as a libref, or library reference).	procprintdata=clinic.admit2; run; prints the data in a data set. The DATA= option tells SAS what data to use for the procedure.		DATA sales; INFILE "C:\CDBookSurvey\OnionRing.dat"; INPUT VisitingTeam \$ 1-20 ConcessionSales 21-24 BleacherSales 25-28 OurHits 29-31 TheirHits 32-34 OurRuns 35-37 TheirRuns 38-40; RUN;	
Sashelp: library for SAS (Read only); Sasuser: to store your files, Permantnt , work: temporary folder for not need to be saved	'SAS statement characteristics: 1. begins with a SAS keyword. 2. ends with a semicolon. A RUN statement tells SAS to process all the preceding lines of the step			
read other vendor's data directly using SAS/ACCESS.	DATA toads; INFILE "C:\CDBookSurvey\ToadJump.p.dat"; INPUT ToadName \$ Weight Jump1 Jump2 Jump3; RUN; * Print the data to make sure the file was read correctly;			
* Create a SAS data set named contest; * Read the file Pumpkin.dat using formatted input; DATA contest; INFILE "C:\CDBookSurvey\Pumpkin.dat"; INPUT Name \$16. Age 3. +1 Type \$1. +1 Date MMDDYY10. (Score1 Score2 Score3 Score4 Score5) (4.1); RUN; The +1 skips over one column. By putting the variables and the informat in separate sets of parentheses, you only have to list the informat once.				
P65				

Regular Expression	Char. Pattern matching perl	Multiple substitution in a string in one step	1. Match 2. Substitute 3. Split	Parse
Pattern matching via several steps	//Sstr =~ /this/i : i makes case insensitive	\$str =~ /-?23/; # i.e. 0 or 1 -'s	\$str =~ /T*/; # matches 0 or more T's	

<pre>\$str = "This is a string"; \$pattern = "ing"; if (\$str =~ /\$pattern/) { print "match"; } else { print "no match"; }</pre>		<pre>+ means 1 or more \$str =~ /dog{4}y/; # matches 4 g's \$str =~ /dog{1,5}y/; # matches 1,2, 3, 4, or 5 g's</pre>	<pre>/(cat){3}/ # catcatcat /cat{3}/ # cattt /+ : to match + \\: to match \</pre>	<pre>".": any char. ./*: any string even empty</pre>	<pre>\. : literal period escape it "." doesn't match a newline List of characters that need to be escaped: \[\ (\) \ [\] \ { \ } \ ^ \ \$ \ * \ ? \ .</pre>	
<pre>my \$str = "The dog"; \$str =~ /dog/; # matches \$str =~ /\^dog/; # doesn't work: "d" must be the first character \$str =~ /dog\$/; # works: "g" is the last character</pre>	<pre>\$str = "There is a dog"; \$str =~ /The/; # matches \$str =~ /The\b/; # doesn't match because the "e" isn't at the end of the word ("b": bgn or end) [135]: char. class match any 1, 3, or 5 [0-7]: matches digit 0 to 7; if hyphen first: [-0-9]</pre>	<pre>[^0-9]/ [^\0-9] matches any char which is not in class [0-9]]: match square bracket as well [135]?: quantifier could also be used</pre>	<pre>Unless want include other chars.: \d = any digit = [0-9] \s = whitespace (spaces, tabs, newlines) = [\t\n] \w - word character = [a-zA-Z0-9_] For negation capital: \D, \S, \W</pre>			
<pre>Memory: saved in parentheses. The matching string is saved in scalar variables starting with \$1. \$str = "The z number is z567890"; \$str =~ /is z(d+)/; print \$1; # prints "567890"</pre>	<pre>Memory inside regex: \1 instead of \$1: /([AG]{3})CCGG\1/ matches AGGCCGGAGG But it is match exactly although it is char class When several match, only actual match captur. \$str = "My pet is a cat"; \$str =~ /\b(cat dog)\b/; print \$1; will print "cat". Finding blank lines (space or tab): /^s*\$/ matching letters only: /^A-Za-z\$/ or /^[\W\d_]\$/</pre>	<pre>Greedy (+) vs. Lazy (?) matching: \$str = "The dogggg"; \$str =~ /The (dog+)/; #greedy print \$1; # prints dogggg \$str =~ /The (dog+?)/; # lazy print \$1; # prints "dog" /ATG(.*?)TAG/: everything b/w 1st ATG and 1st TAG /ATG(*)TAG/: everything b/w 1st ATG and last TAG Warning: \w does not match every words since they include: (^), 1st, (-) Concat: my \$foo .= \$bar;</pre>	<pre>Words can be found within words: "there goes the cat" =~ m/the/ matches the first word, not the fourth, but =~ m/bthe\b/ matches the word "the" only. # Remove all HTML except "p" tags \$html =~ s/{<(??/?)(?:[^\pP] \pP[^\s>])>}/g; time of day: For example. 11:30. [01][0-9]:[0- 5][0-9] won't work well, because it would allow such impossible times as 19:00 and 00:30. A more complicated construction works better: (1[012] [1-9]):[0-5][0-9]. That is, a 1 followed by 0, 1, or 2, OR any digit 1-9.</pre>			
<pre>Different variables are counted from left to right by the position of the opening parenthesis: /(the ((cat) (runs)))/; captures: \$1 = the cat runs; \$2 = cat runs; \$3 = cat; \$4 = runs.</pre>	<pre>Multiconcat.: my \$string=join "",\$this, \$that, \$the, \$other;</pre>	<pre>Substitute operator works directly on string</pre>	<pre>Concatenating string: my \$s=\$s.\$b;or ="\$s1\$s2"</pre>			
<pre>Things within a delimiter: e.g. You want what's within the tags, everything between <a and >, so you try m/<a(.*)>/. Need for lazy evaluation here, as there is probably more than one tag on the page. But still, this expression picks up the ending >, because it matches *. So, use m/<a[^>]*>/ That is, zero or more characters that are not >. Usually you don't really want .*</pre>		<pre>BLAST scores (e-values) can be either decimals or exponents: 0.05 and 2e-40 and both typical values. To capture these numbers, use ([-e.d]+). Also, sometimes BLAST scores start with e: e-35, for example. Perl doesn't recognize this as a num., so you have to add a leading "1": \$score = "1" . \$score if substr(\$score, 0, 1) eq "e";</pre>				
<pre>For nested html/xml: non-word characters as delimiters: proceed with "m" inst. of "/" e.g. m<cat>, m@cat@, m\cat\ (obfuscation)</pre>		<pre>the use of memory for capturing parts of the match: \$str = "I have 2 cats and 3 cars at home"; @arr = split /a[rt]/, \$str; @arr has 3 elements: \$arr[0] = "I have 2 c" , \$arr[1] = "s and 3 c" , and \$arr[2] = "s at home".</pre>				
<pre>\$str = "A cat ia a nice pet"; \$str =~ s/cat/dog/; print \$str; # prints "A dog is a nice pet"</pre>		<pre>/s args: 1st: regular expression, met characters, assertions, alternatives, character classes and parentheses, capturing group within () also work</pre>	<pre>/s 2nd arg: NOT a regular expression, but one specific quoted string or variable (\$1, \$2, ..) Substitution to remove characters: s/[^\ACGT]//g</pre>			
<pre>Substitute all: add a "g" to the end \$str = "A cat is a cat is a cat"; \$str =~ s/cat/dog/; # gives "A dog is a cat is a cat" \$str =~ s/cat/dog/g; # gives "A dog is a dog is a dog"</pre>	<pre>Making case insensitive: adding "i" to the end: \$str = "Cat"; \$str =~ s/cat/dog/; # no changes made; "cat" doesn't match "Cat" \$str =~ s/cat/dog/i; # \$str is now "dog"</pre>	<pre>Substitution and assignment: (\$newstr = \$oldstr) =~ s/cat/dog/; # == > = keep Translate tr//: pram: list of indiv. char \$str = "ACCGTTAC"; \$str =~ tr/ACGT/TGCA/; \$str is now TGGCAATG</pre>	<pre>Converting all to uppercase: \$str =~ tr/a-zA-Z/; Count: \$str = "AGCCTNNNCGTTANTA"; \$num = (\$str =~ tr/ACGT//); # returns the number of A, C, G and T, without counting the N's.</pre>			
<pre>find all instances of a pattern and give their positions. while (\$str =~ /ATG/g) { my \$position = pos \$str; my \$start_position = \$position - 4; print "\$start_position, "; }</pre>		<pre>position of the next character after the end of the matched string is found using "pos" followed by the name of the string variable being matched. (10/31/2012) SAS regular expr:</pre>		<pre>Single step: PROC SQL and the %SYSFUNC macro command. SAS(RX) multiple step pattern: program editor Perl (PRX)</pre>	<pre>Perl (PRX)</pre>	<pre>Description</pre>
<pre>RXparse(f)</pre>	<pre>PRXparse(f)</pre>	<pre>Compile RX 4 pattern matching</pre>	<pre>proc sql; create table work.MarkTabTest as select S.*, prxparse("/w*chips/") as re, ifc(prxmatch(calculated re, S.product), prxposn(calculated re, 0, S.product), " ") as PRX_Return_String from sashelp.snacks as S; alter table work.MarkTabTest drop re; quit;</pre>	<pre>CALL PRXPOSN(r)</pre>	<pre>Rtn start pos & length for capt. Buffer 2 substr</pre>	
<pre>RXMatch(f)</pre>	<pre>PRXMatch(f)</pre>	<pre>Search pptrn match & rtn Pos. mch fnd</pre>		<pre>PRXPOSN(f)</pre>	<pre>Rtrn the val. 4 a capt. buff</pre>	
<pre>Call RXSUBSTR(R)</pre>	<pre>Cal P..(r)</pre>	<pre>Rtrn POS & lngth of substr Match ptrn (RX includes scores</pre>		<pre>PRXPAREN (f)</pre>	<pre>Rtrn the last brack. Match 4 which there is a match in pttrn</pre>	
<pre>Call RXChange(r)</pre>	<pre>...P...(r) PRXchange (f)</pre>	<pre>Pttrn matching replacement</pre>		<pre>CALL PRXNEXT (r)</pre>	<pre>Rtrn the pos. & leng. Of a substr that matches a pptrn & iterate over multip match withn str</pre>	
<pre>Call RXFree(r)</pre>	<pre>...P...@</pre>	<pre>Free unneeded mem. Alloc 4 RX</pre>		<pre>CALL PRXDEBUG (r)</pre>	<pre>Enables perl reg. exp. In a DATA step to send debug output to the SAS log</pre>	
<pre>The recommended best practice is to use the CALL routines in a data step;</pre>						
<pre>Use any expression in this list and search in SAS search</pre>						
<pre>data_null_; /* Use a pattern to replace all occurrences of cat, */ /* rat, or bat with the value TREE. */ length text \$ 46; RegularExpressionId = prxparse('s/[crb]at/tree/'); text = 'The woods have a bat, cat, bat, and a rat!'; /* Use CALL PRXCHANGE to perform the search and replace. */ /* Because the argument times has a value of -1, the */ /* replacement is performed as many times as possible. */</pre>		<pre>data_null_; ExpressionID = prxparse('/[crb]at/'); text = 'The woods have a bat, cat, and a rat!'; start = 1; stop = length(text); /* Use PRXNEXT to find the first instance of the pattern, */ /* then use DO WHILE to find all further instances */ /* PRXNEXT changes start param. so that searching */ /* begins again after the last match. */ call prxnext(ExpressionID, start, stop, text, position, length); dowhile (position > 0);</pre>				

<pre>call prxchange(RegularExpressionId, -1, text); put text; run;</pre>		<pre>found = substr(text, position, length); put found= position= length=; if start > stop then position = 0; else call prxnxt(ExpressionID, start, stop, text, position, length); end; run;</pre>	
RXparse : creating parsing function Use this: To match this: \$a or \$A : a-z A-Z \$c or \$C : 0-9 a-z A-Z \$d or \$D : 0-9 \$l or \$l : a-z A-Z (only if first character in string) \$o or \$L : a-z \$u or \$U : A-Z \$w or \$W : whitespace	Communicate with other functions by passing value: Rx = rxparse (\$character class.);	<pre>position, length); end; run;</pre>	
	uge performance degradation will occur if this line is executed multiple times.		
	User defined characters: Rx = rxparse (\$.A-Z.);		
	Character class complement: Rx = rxparse (^ .AEIOU.);	To change (To keyword): in place Rx = rxparse (.St. to Street.); Call rxchange(rx, 1000, addr_1); # max number of times str to change	
Use This: To Match This: \$F or \$F Floating Point Number \$N or \$N SAS Name \$P or \$P Prefix (User Specified) \$Q or \$Q User specified String \$S or \$S Suffix (User Specified)	Statistic: RXSUBSTR addr_1 = '.123 West St..' rx2 = rxparse(" St. to Street"); call rxsubstr(rx2,addr_1,position,length); Will result in position = 10 and length = 3.	Statistical characteristics of the data are examined using PROC UNIVARIATE. (normal tests normality hypothesis) <pre>proc univariate data=steam plot normal ; var steamuse temp ; title2 'Univariate Descriptive Statistics' ; run;</pre>	scatter plots of the raw data in regression problems. <pre>proc plot data=steam ; title2 'Scatterplot of Raw Data' ; plot steamuse*temp ; run;</pre>
	statex = "[\$ #'North' #9] [\$ #'South' #4] [\$ #'East' #2] [\$ #'West' #1] "; rx = rxparse (statex); call rxsubstr(rx, addr_1, start, len, state);		The PROC GPLOT : high-resolution graph of the raw data with the reg line superimposed. Graph form is specified in the SYMBOL statement, here: least squares reg line should be used to "interpolate" between data points and that raw data points should be indicated by plus signs.
<pre>proc reg data=steam ; title2 'Least Squares Analysis' ; model steamuse = temp ; run;</pre>	PROC MEANS or PROC CORR Defining SAS library to work with data: Libname cdbkts "C:\\"; /* the place the files would be put */ run;	<pre>proc gplot data=steam ; symbol1 = r1 value = PLUS ; plot steamuse*temp ; title2 'Observed Values and Estimated Regression Line' ; run ;</pre>	
<pre>proc sort data=test ; by id ; run ; #patient diagnose proc means data=test noprint ; by id ; var dx1 dx2 dx3 ; output out=results max=; run ; proc print data=results ; run ;</pre>	PROC IMPORT DATAFILE = " c:\CDBookSurvey\cdbkbook. xls " DBMS=XLS OUT=cdbkts.cdbk; RUN; * Read an Excel spreadsheet using PROC IMPORT; Proc contents data= cdbkts.cdbk position; Run; *content of the file will be shown; proc means data= cdbkts.cdbk; var age; run; *mean of specific data;	<pre>data cdbkts.cdbkusage; *create new data set ; set cdbkts.cdbk; if BGHTCDST<4 then offline=0; else offline=1; if BGHTCDON<4 then online=0; else online=1; run;</pre>	<pre>proc freq data=cdbkts.cdbkusage; * cross tab ; tables offline*CDTRNS online*CDTRNS/chisq; run; *test of association b/w CDTRNS and online offline ;</pre>
High resolution plot of normal and exponential: data randata; drop i; label normal_x = 'Normal Random Variable' exponential_x = 'Exponential Random Variable'; do i = 1 to 100; normal_x = 10*rannor(53124) + 50; exponential_x = ranexp(18746363); output; end; run; title '100 Obs Sampled from a Normal Distribution'; title2 'Normal Q-Q Plot'; proc univariate data=randata noprint; qqplot normal_x / normal(mu=est sigma=est); inset meanstd / format=3.0 header = 'Normal parameters' position = se; run;	proctimeseries data=sashelp.air out=series outtrend=trend outseason=season print=seasons; id date interval=qtr accumulate=avg; var air; run; procentropy data = cdbkts.cdbk; model BGHTCDST=RSKFRAUD RSKSHIP RSKPERF RSKINFO AVERISK HOURSONWEBVISTPRODINFONU MPRUCH AVERUSE ASRTRNK CONVRNK HASSRNK ENJRKNKINFORK SERVRNKPRICRNK SEXAGEEDUC INCM; run;	<pre>data melanoma ; *analysis of unobserved component struct. models; input Incidences @@ ; year = intnx('year','1jan1936'd,_n_-1) ; format year year4. ; label Incidences = 'Age Adjusted Incidences of Melanoma per 100,000'; datalines ; 0.9 0.8 0.8 1.3 1.4 1.2 1.7 1.8 1.6 1.5 1.5 2.0 2.5 2.7 2.9 2.5 3.1 2.4 2.2 2.9 2.5 2.6 3.2 3.8 4.2 3.9 3.7 3.3 3.7 3.9 4.1 3.8 4.7 4.4 4.8 4.8 4.8 ; run ; src: http://support.sas.com/rnd/app/examples/ets/melanoma/index.htm /</pre>	<pre>proc ucm data = melanoma; id year interval = year; model Incidences ; irregular ; level ; slope ; cycle ; run ; odshtml ; ods graphics on ; proc ucm data = melanoma; id year interval = year; model Incidences ; irregular ; level variance=0 noest ; slope variance=0 noest ; cycle plot=smooth ; estimate back=5 plot=(normal acf); forecast lead=10 back=5 plot=decomp; run ; ods graphics off ; odshtml close ;</pre>
<pre>proc sort data = cdbkts.cdbkusage; by BGHTCDST; run;</pre>	<pre>proc print data=cdbkts.cd bkusage(obs = 7); run; * only print 7 observation ;</pre>	<pre>data origdata; *Logit; input ttime1 ttime2 ttime3 choice @@; datalines;</pre>	<pre>data enso(drop=pi); *nonparametric model; set enso; pi = 4*atan(1); sin1=sin(2*pi*Month/12);</pre>

Conjoint sample title 'Nonmetric Conjoint Analysis of Ranks'; procformat ; value BrandF 1 = 'Goodstone' 2 = 'Pirogi' 3 = 'Machismo'; value PriceF 1 = '\$69.99' 2 = '\$74.99' 3 = '\$79.99'; value LifeF 1 = '50,000' 2 = '60,000' 3 = '70,000'; value HazardF 1 = 'Yes' 2 = 'No'; run ; data Tires; input Brand Price Life Hazard Rank; format Brand BrandF9. Price PriceF9. Life LifeF6. Hazard HazardF3.; datalines ; 1 1 2 1 3 1 1 3 2 2 1 2 1 2 14 1 2 2 2 10 1 3 1 1 17 1 3 3 1 12 2 1 1 2 7 2 1 3 2 1 2 2 1 1 8 2 2 3 1 5 2 3 2 1 13 2 3 2 2 16 3 1 1 1 6 3 1 2 1 4 3 2 2 2 15 3 2 3 1 9 3 3 1 2 18 3 3 3 2 11; proc transregmaxiter=50utiliti esshort; odsselect TestsNote ConvergenceStatus FitStatistics Utilities; model monotone(Rank / reflect) = class(Brand Price Life Hazard / zero=sum); output replacepredicted; run ; proc printlabel; var Rank TRank PRank Brand Price Life Hazard; label PRank = 'Predicted Ranks'; run ; MCMC Logistic Bayesian: data prior; input _type_ \$ Intercept x; datalines ; Var 25 25 Mean 0 0 ; run ; odsgraphicson; title "Bayes with normal prior"; proc genmoddescendingdata= testmcmc; model count/n = x / dist =binomial link =logit; bayesseed =10231995 nbi =100 0 nmc =21000 coeffprior =normal(input =p rior) diagnostics =all statistics =(summaryinterv al) plot =all; run ;	Clustering: Proc fastclusdata=cdbkfst .cdbkusage maxc =4 out=cdbkfst.cdbcluster; Var BGHTCDON BGHTBKST; Run ; proc plot; plot BGHTCDON*BGHTBKST=cluste r; run ; data endometrial; * bayesian estimation ; input nv pi eh hg ; nv2 = nv - 0.5; pi2 = (pi-17.3797)/9.9978; eh2 = (eh-1.6616)/0.6621; datalines ; 0 13 1.64 0 0 16 2.26 0 0 8 3.14 0 0 34 2.68 0 0 20 1.28 0 0 5 2.31 0 0 17 1.80 0 0 10 1.68 0 1 11 1.01 1 1 21 0.98 1 0 5 0.35 1 1 19 1.02 1 0 33 0.85 1; run ; proc genmoddescending; model hg = nv2 pi2 eh2 / dist =bin link =logit; bayescoeffprior =normal (var=1.0) diagnostics =mcerrornmc=1 000000; run ; proc genmoddescending; model hg = nv2 pi2 eh2 / dist =bin link =logit; bayescoeffprior =normal (var=100) diagnostics =mcerrornmc=1 000000; run ; * src : http://support.sas.com/d ocumentation/cdl/en/stat ug/63347/HTML/default/vi ewer.htm#statug_genmod_s ect007.htm ;	16.481 16.196 23.89 2 15.123 11.373 14.182 2 19.469 8.822 20.819 2 18.847 15.649 21.28 2 12.578 10.671 18.335 2 11.513 20.582 27.838 1 11.852 12.147 15.672 2 15.557 8.307 22.286 2 ; run ; data newdata(keep=pid decision mode ttime); set origdata; array tvec{3} ttime1 - ttime3; * travel time (ttime1..ttime3); retain pid 0; pid + 1; do i = 1to3; * extract whether the choice is chosen ; mode = i; ttime = tvec{i}; decision = (choice = i); * not chosen =0, chosen =1; output ; end ; * data format: panel dataset where, in this case, the variable pid indexes the cross-section dimension and the variable mode indexes the time dimension. ; run ; proc mdc data=newdata; * conditional logit using maximum likelihood ; model decision = ttime / type=clogit nchoice =3 optmethod =qn covest =hess; id pid; run ; DATA dads; * father of family ; INPUT famid name \$ inc ; CARDS ; 2 Art 22000 1 Bill 30000 3 Paul 25000 ; RUN ; DATA faminc; * income of family ; INPUT famid faminc96 faminc97 faminc98 ; CARDS ; 3 75000 76000 77000 1 40000 40500 41000 2 45000 45400 45800 run ; PROCSORTDATA =dads OUT =dads2; * you must always sort before merge BY famid; RUN ; PROCSORTDATA =faminc OUT =faminc2; BY famid; RUN ; DATA dadfam ; MERGE dads2 faminc2; BY famid; RUN ; PROCPRINTDATA =dadfam; RUN ;	cos1=cos(2*pi*Month/12); run ; *12 cycle time and estimate residuals ; proc regdata=enso; model Pressure=sin1 cos1; output out=enso1 r=FilteredPressure; run ; odsoutput OutputStatistics=enso1Stats FitSummary=enso1Summary; proc loessdata=enso1; model FilteredPressure=Month/smooth =0.12 dfmethod =exact; run ; title "Filtered ENSO Data"; symbol color=black value =dot i =none h =3.5 pct ; symbol 2color=blue interpol =join value =none width =2; proc gplotdata=enso1Stats; format DepVar f2.0; format Month f3.0; plot (DepVar Pred)*Month/ overlay hminor = 0 vminor = 0 vaxis = axis1 href = 4587129 frame ; axis label = (r=0a=90) order =(-6to6by2); run ; Factor analysis: perception map: positioning & advertis. Factor analysis (loading) Data corrmatr (type=corr); input M P C E H F; Type = "CORR"; Cards ; 0.37 0.62 0.54 0.32 0.284 0.37 0.62 1.00 0.51 0.38 0.351 0.43 0.54 0.51 1.00 0.36 0.336 0.405 0.32 0.38 0.36 1.00 0.686 0.73 0.284 0.351 0.336 0.686 1.00 0.7345 0.37 0.43 0.405 0.73 0.7345 1 run ; proc factormethod=prinrit rotate =v corr sascreeresidualspreplotp lot; var M P C E H F; run ;
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R Programming Language	read.csv function: rds comma-delimited	>txt.csv <- read.csv("c:/temp/GLAccDesc.csv");	R is case sensitive
Tm package for text mining	read.table function, which reads data that is tab-, comma-, or space-delimited [getReader() to see other]	"Suggests: filehash, proxy, Rgraphviz, Rmpi, RWeka, snow, Snowball, XML", Update other packages as well	up-to-date list of available datasources and readers available within the tm package: getSources()
cdbook.csv<- read.csv("c:/CDBookSurvey/cdbook.csv")	In order to analyze text it needs to be converted to Corpus. (readable by tm)	txt <- Corpus(DataframeSource(txt.csv))	Note in csv: the file needs to be index since duplicate index is not allowed

firefox.csv<-read.csv("c:/CDBookSurvey/firefoxanalyzed.txt")	firefox <- Corpus(DataframeSource(firefox.csv))	summary(firefox): summarize and inspect
inspect(firefox[1:5])	getTransformations(): available func. For preprocessing txt	firefox <- tm_map(firefox, tolower): convert to lower case
firefox <- tm_map(firefox, removeWords, stopwords("english"))#remove stop words	>txt <- tm_map(txt, removeNumbers)#remove numbers	>txt <- tm_map(txt, removePunctuation) #remove punctuations
for (j in 1:length(txt)) txt[[j]] <- gsub("enterprise risk management", "erm",txt[[j]]) #replace wth abbr.		
tm_map(txt,tolower): convert to lowercase	for (j in 1:length(txt)) txt[[j]] <- gsub("/", " ",txt[[j]]): substituting "/" with " "	
newstopwords <- c("and", "for", "the", "to", "in", "when", "then", "he", "she", "than") // use customized stop word in addition to normally available one		
firefox <- tm_map(firefox, removeWords, newstopwords)		
tm package stemming function: Remove words such as: "es", "ed" and "s":	dtm <- DocumentTermMatrix(firefox) : create document term Matrix	
firefox <- tm_map(firefox, stemDocument)# stem words	idx <- which(dimnames(myTdm)\$Terms == "alexa")inspect(myTdm[idx+(0:5),1:10]) :	
firefox <- tm_map(firefox, stemCompletion, dictionary=firefoxcopy) #stem completion	show 5 terms in 10 document of dtm after term "alexa"	
dtm3 <- removeSparseTerms(dtm, 0.94): removing sparse items of document term matrix	library(tm): load tm library	
Remove numbers and punctuations: firefox <- tm_map(firefox, removeNumbers)	Read dtm that is created in perl: dtm.csv<-read.csv("c:/Directory/dtm.csv")	correlation is an indicator of how closely related two terms are (similarity measure)
firefox <- tm_map(firefox, removePunctuation)	Check the content of the dtm: inspect(myTdm[0:10,1:10])	finds all words with a correlation of at least: findAssocs(dtm, "nice", 0.2):
Similarity measures can be applied across rows as well as across columns of a database. When applied across rows the similarity indicates how similar two records are. In the case of text mining, a similarity measure would indicate how many words the two rows have in common .		Chi-Squared : measures how closely related two categorical variables are, Phi : measures the correlation b/w binary categorical variables
>library(proxy) # dissimilarity check >dissimilarity(dtm3, method = "cosine")[1:10]	Euclidian distance for dissimilarity (i, j): records, m:# variables) $d_{ij} = (\sum_{k=1}^m (x_{ik} - x_{jk})^2)^{\frac{1}{2}}$	Cluster analysis: Maximize
List of terms of dtm: rownames(myTdm)		
require(vegan)### some sample data data(dune) # draw cluster kclus <- kmeans(dune,centers= 4, # kmeans iter.max=1000, nstart=10000) dune_dist <- dist(dune) # distance matrix # Multidimensional scaling cmd <- cmdscale(dune_dist) # plot MDS, with colors by groups from kmeans groups <- levels(factor(kclus\$cluster)) ordiplot(cmd, type = "n") cols <- c("steelblue", "darkred", "darkgreen", "pink") for(i in seq_along(groups)){ points(cmd[factor(kclus\$cluster) == groups[i],], col = cols[i], pch = 16)} # add spider and hull ordispider(cmd, factor(kclus\$cluster), label = TRUE) ordihull(cmd, factor(kclus\$cluster), lty = "dotted")	data("crude")tdm <- TermDocumentMatrix(crude, control = list(removePunctuation = TRUE, removeNumbers = TRUE,stopwords = TRUE)) source("http://bioconductor.org/biocLite.R") biocLite("Rgraphviz") plot(dtm, terms = findFreqTerms(dtm, lowfreq = 1)[1:20], corThreshold = 0)# cluster draw library(fpc) # draw clusters plotcluster(dtm3, glKmeans\$cluster) Not removing 2 letter words myTdm <- TermDocumentMatrix(firefox, control = list(wordLengths=c(1,Inf))) myTdm findFreqTerms(myTdm, lowfreq=3): frequent words frequency not less than 3	x <- c(1,2,3,4,5,6) # Create ordered collection (vector) y <- x^2 # Square the elements of x print(y) # print (vector) y mean(y) # Calculate average (arithmetic mean) of (vector) y; result is scalar > var(y) # Calculate sample variance > lm_1 <- lm(y ~ x) # Fit a linear regression model "y = f(x)" or "y = B0 + (B1 * x)" # store the results as lm_1 > print(lm_1) # Print the model from the (linear model object) lm_1 > summary(lm_1) # Compute and print statistics for the fit # of the (linear model object) lm_1 > par(mfrow=c(2, 2)) # Request 2x2 plot layout > plot(lm_1) # Diagnostic plot of regression model
# plot of more frequent words [horizontal] termFrequency <- rowSums(as.matrix(myTdm)) termFrequency <- subset(termFrequency, termFrequency>=3) library(ggplot2) qplot(names(termFrequency), termFrequency, geom="bar") + coord_flip() # for vertical: barplot(termFrequency, las=2)	Find words that are highly associated with a word: findAssocs(myTdm, "love", 0.50) Page 28 text mining hand book	termFrequency <- rowSums(as.matrix(dtm)) highly used terms termFrequency <- subset(termFrequency, termFrequency>=2) library(ggplot2) qplot(names(termFrequency), termFrequency, geom="bar") + coord_flip()

Text mining			
Text mining phases: 1. Preprocessing and integration of unstructured data, 2) statistical analysis of the preprocessed data to extract content from the text	Parsing: 1. Array of the words to be parsed 2. Search for space and record the position 3. Extract string from first position to position before space 4. Go to step 2 [split func. Of perl]	Simple analysis: Indicator 0/1 for existence of word e.g. "home owner": if the lower case of word is same then 1, else 0 (lc: lower case) => code	\$target = "(homeowner)"; \$i=0; \$flag=0; foreach \$x (@words) { if (lc(\$x) =~ /\$target/) { \$flag=1; } }
unstructured text data is converted into structured data			
Steps of preprocessing: 1.Parse the data. That is, extract the words from the data, typically discarding spaces and punctuation. 2. Eliminate articles and other words that convey little or no info . 3. Replace words that are synonyms , and plural and other variants of words with a single term. 4. Create the structured data , a table where each term in the text data becomes a variable with a numeric value for each record.	Parse : Extracting words: @words =split (/[\s+]/,\$Response3); \$len=@words; for(\$i=0;\$i<\$len;\$i++){ print "\$i'th word is: @words[\$i]\n"; } Removing the punctuations , and unwanted characters: \$Response3=~ s/[-,?!"'{}&;\ /]/g;	Counting frequencies : foreach \$word (@words){ ++\$counts{lc(\$word)}; } #sort counts for keys foreach \$value(sort{\$counts{\$a} cmp \$counts{\$b}} keys %counts){ #print the word and the count for the word print "\$value \$counts{\$value} \n"; }	Simple text statistics : 1. Length statistics of the word: @countlen[length(\$word)] +=1 Produce matrix of term indicator : 1) create a list of all words in the data base (which will be referred to as the grand dictionary), 2) check each record (either claim description or survey response) for the presence of each word, 3) create an indicator value of 1 if the word is present, otherwise a zero is recorded and 4) print the results to a file for further processing.
find a regular sentence structure that ends in a period (\. \.)	Bag of words concept: order discarded and only existence matter (could be per sentence or per comment)	Create term frequency DB	Elimination of Stop words : the articles "the" and "a" [not add info.]by substitut.
#!perl -w: Create the matrix of term # Program TermDocData.pl # This program computes the term-document matrix # a key part is to tabulate the indicator/count of every	#!perl -w eliminate stop words while creating matrix of terms # StopWords.pl # This program eliminates stop words and computes the term-document matrix		stemming:synonym & abbrev. handling important task and you can use normal DBs, yet, usually it needs to be tailor made based on context (substitute)

Perl (Quick and Nasty)	Perl (Quick and Nasty)
Scalar Variables # End of line comments begin with a # \$a = 17; # Scalar variables begin with a dollar symbol # The Perl assignment operator is = # Statements finish with a semicolon ; \$b = 0x11; # Hexadecimal (17 in decimal) \$c = 021; # Octal (17 in decimal) \$d = 0b10001; # Binary (17 in decimal) \$f = 3.142; # Floating point \$a = \$a + 1; # Add 1 to variable \$a \$a += 1; # Add 1 to variable \$a \$a++; # Add 1 to variable \$a \$b = \$b * 10; # Multiply variable \$b by 10; \$b *= 10; # Multiply variable \$b by 10; # Other arithmetic operators include: # ** Exponentiation # % Modulo division # ++ Auto increment # -- Auto decrement # < Numeric less than # > Numeric greater than # == Numeric equality # != Numeric inequality # <= Numeric less than or equal to	Scalar Variables Cont. \$a = 'Number of DFFs: '; # No interpolation with 'single quotes' \$b = "\$a\$c\n"; # Interpolation (variable substitution) with "double quotes" # \n is the newline character print \$b; # This makes "Number of DFFs: 17\n" appear on the standard output print \$a, \$c, "\n"; # As does this line because print takes # a comma separated list of arguments to print print "That's all\n"; # No commas means a list of one element # String operators include: # lt String less than # gt String greater than # le String less than or equal to # ge String greater than or equal to # cmp String compare: Returns -1 0 1 print 'one' lt 'two'; # Prints 1 # ASCII-betically 'o' is less than 't' print 'buf4' lt 'buf3'; # Prints nothing (that is undef, numerically zero) # Perl's undefined value is undef # ASCII-betically '4' is not less than '3'
	Logic and Truth 0; # Integer zero 0.0; # Decimal zero '0'; # String containing a single zero character ''; # Empty string

<pre># >= Numeric greater than or equal to # <=> Numeric compare: Returns -1 0 1</pre>		<pre>undef; # Undefined</pre>	
Logic and Truth <pre>\$a = 0; \$b = 45; # More than one stmt per line possible print(\$a and \$b++); # prints 0 \$a = 22; print(\$a and \$b++); # prints 45 print \$b; # prints 46 # * \$b++ only evaluated when \$a was true # Some logic operators take shortcuts # Other logical operators include # or Logical OR # Logical OR # and Logical AND # && Logical AND # not Logical NOT # ! Logical NOT # Bitwise OR # & Bitwise AND # ~ Bitwise NOT print 6 & 5; # prints 4, 0b0110 & 0b0101 = 0b0100 print 6 5; # prints 7, 0b0110 0b0101 = 0b0111 print ! 0; # prints 1 print ! 5; # prints nothing (that is undef or false) print ~5; # prints 4294967290, same as: # 0b1111111111111111111111111111010</pre>		Arrays and Hashes <pre>@components = ('X_LUT4', 'X_AND2', 'X_BUFGMUX', 'X_BUF_PP', 'X_FF'); # or use qw''. Saves typing commas or quotes, gives the same result # qw stands for Quoted Words @components = qw'X_LUT4 X_AND2 X_BUFGMUX X_BUF_PP X_FF'; # or even put the data in columns, gives the same result again @components = qw' X_LUT4 X_AND2 X_BUFGMUX X_BUF_PP X_FF '; # Easier to read this way push(@components, 'X_MUX2'); # Push another item onto the top push(@components, 'X_ONE'); # And one more print \$components[0]; # Prints element 0, that is, 'X_LUT4' print \$components[5]; # Prints element 5, that is, 'X_MUX2' print "@components\n"; # Prints everything separated by spaces: # X_LUT4 X_AND2 X_BUFGMUX X_BUF_PP X_FF X_MUX2 X_ONE print @components; # No double quotes, no spaces: # X_LUT4X_AND2X_BUFGMUXX_BUF_PPX_FFX_MUX2X_ONE</pre>	
		<pre>while(@components) { # Array in scalar context \$next_component = shift(@components); print "\$next_component\n"; } # Array variable @components is now empty</pre>	
Command Line Arguments <pre># This script is called process_netlist.pl # Perl scripts often have the file extension .pl \$netlist_filename = \$ARGV[0]; \$report_filename = \$ARGV[1]; print " Processing \$netlist_filename\n"; print " Writing report to \$report_filename\n"; print " ARGV contains '@ARGV'\n"; # Use it in this way: #C:\perl process_netlist.pl chip_timesim.vhd report.txt #Processing chip_timesim.vhd # Writing report to report.txt #ARGV contains 'chip_timesim.vhd report.txt' # C:\</pre>		Arrays and Hashes <pre># Initialising several hash keys %components = qw' X_LUT4 0 X_AND2 0 X_BUFGMUX 0 X_BUF_PP 0 X_FF 0 '; # keys # ^ values \$components{'X_LUT4'} = 1; # Set key X_LUT4 to the value 1 \$components{'X_LUT4'}++; # Increment value associated with X_LUT4 print \$components{'X_FF'}; # Print value associated with X_FF @keys = keys %components; # Get a list of hash keys print "@keys\n"; # Print them - order is indeterminate \$components = (); # Emptying the components hash</pre>	
<pre>if(\$ff_count == 1) # Is this expression true or false? { # Do this action if it is true print "There is 1 flip flop\n"; } else { # Do this action if it is false print "There are \$ff_count flip flops\n"; } # More compact layout if(\$ff_count == 1) { print "There is 1 flip flop\n"; } else { print "There are \$ff_count flip flops\n"; }</pre>		<pre># Counting to one hundred while(\$count < 100) { \$count++; # Perl assumes \$count == 0 # the first time print "\$count\n"; }</pre>	
		Read from screen & print <pre>while(\$line = <STDIN>) { \$line_count++; print "\$line_count: \$line"; } # perl filter_netlist.pl < # chip_timesim.vhd> report.txt</pre>	
		<pre>foreach \$course ('VHDL', 'SystemVerilog', 'SystemC', 'Perl', 'Tcl/Tk', 'PSL') { print "There is a \$course Doulos training course\n"; } # \$course is the loop variable. # It takes the string value 'VHDL' for the first loop # and 'PSL' for the last loop. # Get a list from an array variable foreach \$component (@components) { print "Component is \$component\n"; }</pre>	
		<pre>cleaning strings safely: chomp(\$myvar); # changes \$myvar dropping the last character: chop(\$myvar); # changes \$myvar</pre>	
Files <pre>open(FILE1, '>file1.txt'); # ^ # > means open in write mode print FILE1 "The first line to file1.txt\n"; print FILE1 "The final line to file1.txt\n"; close(FILE1); # Don't have to explicitly close a file print STDOUT "This goes to the standard output\n"; print "So does this\n"; # ^^^^^^ # STDOUT is a file handle that always # refers to the standard output. # It is the default so doesn't have to be stated.</pre>		Files <pre>open(FILE2, 'file2.txt'); # Open in read mode - the default mode \$first_line = <FILE2>; #RD first line from file2.txt into \$first_line # Includes the newline character, \n. while(\$line = <FILE2>) { print \$line; # Read and print remaining lines from file2.txt. } # When every line has been read <FILE2> returns undef. \$standard_input = <STDIN>; # Read a line from the standard input. # Can be the keyboard if run from the command line. chomp(\$standard_input); # Remove the trailing newline character</pre>	
<pre>\$netlist_filename = \$ARGV[0]; # file names in the arg \$report_filename = \$ARGV[1]; #read from first file put open(FILE_IN, \$netlist_filename); open(FILE_OUT, ">\$report_filename"); while(\$line = <FILE_IN>) { \$line_count++; #number print in 2nd file print FILE_OUT "\$line_count: \$line"; } # perl filter_netlist.pl chip_timesim.vhd report.txt</pre>		<pre>use English; \$string = "Novice to Expert in a 3 day Perl course.\n"; if(\$string =~ /\w+/) { # \w+ matches one or more # alphanumeric characters in a row print "Matched: \$MATCH\n"; } Matched:Novice</pre>	
Pattern matching (String) <pre>\$string = "Novice to Expert in a 3 day Perl course.\n"; print \$string; if(\$string =~ m/Expert/) { # A successful match returns 1 so this statement is executed print "This string contains the substring 'Expert'\n"; } # m stands for match # Forward slashes are used to /delimit/ regular expressions. # =~ tells the m operator which string to search. # The m is optional when // are used.</pre>		<pre>Quick intro contd.: • quoted strings: → "\$xyz and other stuff \t \n": like C printf, variables are substituted → '\$xyz and other stuff \t \n': literal printing • operators: numeric vs string → numbers: ==, >, <, >=, <=, !=, <=>; returns -1, 0, or 1 → strings: eq, gt, lt, ge, le, ne • lexicographic comparison: 300 <= 40 is false, 300 le 40 is true • lists and arrays → list syntax: ("abc", "def", "etc"); qw(abc def etc);</pre>	
Grouping: <pre>open(VHDL_FILE, 'chip_timesim.vhd');</pre>		<pre>use English; \$string = "Novice to Expert in a 3 day Perl course.\n"; if(\$string =~ /Perl\s+\w+/) { # first part: matches Perl # second part: matches one or more white # space characters (including space, tab and # newline) # 3rd part: matches one or more alphanumeric # chars. print "Matched: \$MATCH\n"; # Matched: Perl course } # \w? Zero or one letter, digit or # underscore # \w One letter, digit or underscore # \w* Zero or more letters, digits or # underscores # \w+ One or more letters, digits or # underscores # \W One character but not a letter, # digit or underscore # \s White space character, space, tab or # newline # \S One char but not a space, tab or # newline</pre>	

Latex						
$\documentclass{article}$ $\begin{document}$ A minimal \LaTeX document. $\end{document}$	Windows – proTEXt! LATEX processor – TeXnicCenter! editor Landscape: $\special{landscape}$	α ψ ω	$xy \pm$ \times \approx \int_0^∞ \int	x^y x_y x_y^z	$\frac{\partial}{\partial x}$ \begin{figure} \includegraphics{graph} \end{figure}	$\begin{document}$ $\end{document}$
$\$ \alpha = \frac{\beta}{\gamma}$	$\# \$ \% \& \sim _ \wedge \backslash \{ \}$					
$\title{Social network in search}$ $\author{Meisam Hejazinia \ \ Brian Ratchford \ \ Ernan Haruvey}$ $\date{5 December 2012}$ \maketitle	There has to be at least two new line to separate the paragraphs. To double space: $\renewcommand{\baselinestretch}{2}$ \maketitle	$\begin{tabular}{ l r c }$ \hline Person & Money Owing & Silly Com \hline Mr. C & \$1943.12 & pay him agn, Sam \hline Mr. P & \$55.55 & what robbery? \hline Mr. Sc & \$666.00 & the golden rule \hline Mr. Ca & \$300.51 & bad accountants \hline $\end{tabular}$ \begin{figure}				Document Wide Stuff: $\documentstyle{options}[style]$ style: article report book slides options: 11pt 12pt twoside twocolumn titlepage leqno openbib fleqn \pagestyle{style}
Title Page Stuff: \maketitle $\begin{titlepage}$... $\end{titlepage}$ $\begin{abstract}$... $\end{abstract}$	Bibliography and Citation There is (well, it'll be here RSN) some more information here in this document that gives an example of the					

<p>Cross reference: \label{key} assign current counter value to key \ref{key} print value assigned to key</p>	<p>bibtex stuff. \bibliography{...} \begin{thebibliography}{label} ... \end{...} make bibliography; lable is the widest entry label \bibitem[label]{key} begin bibliography entry for citation key with label as its label \cite[not]{keys} cite reference(s) keys with added note</p>	<p>Math: \$... \$ or \ (... \) :Intext formulas \[... \] : displayed formulas \begin{equation} ... \end{equation} a numbered equation \begin{eqnarray} ... \end{eqnarray} numberedequation, like 3 column array environm. \nonumber omits one equation number, eqnarray* omits all _ { ... } subscript. NB: don't need the braces for one character ^{ ... } superscript. NB: don't need the braces for one character ' :prime \frac{n}{d} print the numerator over the denominator \sqrt[n]{arg} the nth root of the argument arg ellipsis \ldots ... \cdots ... \vdots ... Greek letters \alpha ... \omega and \Alpha ... \Omega delimiters \left or \right followed by delimiters \overline{expression} print a rule over the expression space thin \. medium \: thick \; negative thin \!</p>	<p>style: plain empty headings myheadings \pagenumbering{style} style: arabic roman alph Roman Alph</p>
<p>Input from Different Files: \input{file} read the file \include{file} read the file unless not in \includeonly{} \includeonly{filelist} exclude any file not in filelist</p>	<p>\begin{quote} ... \end{quote} short displayed quotation \begin{quotation} ... \end{quotation} long displayed quotation \begin{flushleft} ... \end{flushleft} left flush lines, separated by \\ \begin{center} ... \end{center} centered lines, separated by \\ \begin{flushright} ... \end{flushright} right flush lines, separated by \\ \begin{verse} ... \end{verse} \\ between lines, blank line between stanzas \begin{verbatim} ... \end{verbatim} Fixedlength, typewriter face exactly as formatted use any characters you like!</p>	<p>\begin{em} A long segment of \end{em}</p>	<p>\begin{eqnarray*} % "*" = no line numbering \sum_{n=1}^k \frac{1}{n} & \approx & \ln k + \gamma \\ & = & (\ln 10)(\log_{10} k) + \gamma \\ & \approx & 2.3026 \log_{10} k + 0.57772 \\ \end{eqnarray*}</p>
<p>Lists: \begin{itemize} ... \end{itemize} a 'bulleted' list \begin{enumerate} ... \end{enumerate}: a numbered list \begin{description} ... \end{description}: a list of labeled items</p>			<p>\section{Ordinary Text} % Produces section heading. Lower-level sections % \subsection commands; numbering is automatic!</p>
<p>\begin{thebibliography}{9} % 9 = maximum expected references! \bibitem[Lam]{Lamport, Leslie. \LaTeX: A Document Preparation System. \\ Copyright \copyright 1986, Addison-Wesley Publ.Co.,Inc. \bibitem[Sch]{Schl\oe ff{d{o}nff{t{oo}{ae g}{e}n, \L}{a}{rs. Silly Typography. \\ {em Journal of Linguistic Horseplay 19D} (1977), 23-37. \end{thebibliography} \end{document}</p>			<p>\subsection{Spacing in the source text}</p>
<p>To include images in the document this package should be used. \usepackage{graphicx}</p>			<p>\indent {\bf Bold face type,} \indent {\tt typewriter style type,} \indent {\sf sans-serif type,} \indent {\sl slanted type,} \\ \indent {\sc all caps type.} \\</p>
<p>For probability use: \sim</p>			<p>\begin{equation} a^p + b^p \neq c^p \sim \mbox{for } p>2 \sim \mbox{(see proof in margin)} \label{eq:fermat} \end{equation} $\lim_{n \rightarrow \infty} x_n \neq \pi$ $\forall x \in \mathcal{O} \sim \exists$ $\Delta \sim \mbox{such that} \sim$ $y-x < \Delta \rightarrow y \in \mathcal{O}$ $\text{vspace}{4mm}$ $\Psi' = \frac{d}{d\phi} \left(\begin{array}{c} \phi_2 \\ \phi_1 \end{array} \right)$ $\phi_2 - \phi_1^2 / 2$ $\begin{array}{c} \Theta = \\ \left(\begin{array}{c} ccc \\ 0 \\ -\theta_1 \psi_1 - \psi_2 \end{array} \right) \\ \&0 \& \psi_3 - \phi_1 \end{array}$ $\text{vspace}{4mm}$</p>

<p>#PROGRAM MATCHLINE.TXT TO SEARCH FOR THE PHRASE THAT MOST CLOSELY MATCHES A PHRASE.</p> <p>It finds the record most similar to the input phrase "Credibility of the CAS", using the cosine measure, though other similarity measures could be used instead</p> <ol style="list-style-type: none"> 1. Read the database. 2. Create a hash of all words on each record. 3. Create a hash of all words in the database. 4. Compute the TF-IDF statistic for each term on each record of the database. 5 Read the search string. 6. Compute the TF-IDF for the search string. 7. Compute the cosine correlation between the TF-IDF of the search string and each record in the database. 8. Determine which record is the closest match and print it out. 	<pre> if (exists \$tf[\$i][\$word]) { \$tf_val = \$tf[\$i][\$word]; } else { \$tf_val = 0; } #print OUTP "Word ". \$word. " " . \$tf_val . " df: " . \$df{\$word}. "\n"; \$weight[\$i][\$word] = \$tf_val * log(\$n / \$df{\$word}) / log(2); #print "Weight ". \$weight[\$i][\$word]. " " . "\n"; } # Compute weight of input phrase foreach \$word (sort keys %granddict) { if (exists \$inph{\$word}) { \$tf_val = \$inph{\$word}; } else { \$tf_val = 0; } \$inph_weight{\$word} = \$tf_val * log(\$n / \$df{\$word}) / log(2); } # Step 4 Normalize the column of weights </pre>
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<pre>#!/perl -w # matchline.pl # Usage: matchline.pl <datafile><in phrase file><outputfile> # datafile must be present and a cmd line arg # create a dictionary of all words in a file and alphabetize them open(MYDATA, \$ARGV[0]) or die("Error: cannot open file '\$ARGV[0]'\n"); open(INPH, \$ARGV[1]) or die("Error: cannot open file '\$ARGV[1]'\n"); open(OUTP, ">\$ARGV[2]") or die("Cannot open file '\$ARGV[2]' for writing\n"); print OUTP "#Output results for ".\$ARGV[0]."\n"; \$npnr = 0; # read in the file, get rid of newline and punctuation chars while(\$line = <MYDATA>){ chomp(\$line); \$line =~ s/[-?"'{}]/ /g; @words = split(/ /, \$line); foreach \$word (@words) { ++\$granddict{lc(\$word)}; # this is the hash assignment lc is lowercase ++\$tf[\$npnr]{lc(\$word)}; } \$npnr++; # Read in the input phrase from file \$linecnt = 0; while(\$line = <INPH>){ print OUTP "Input Phrase: " . \$line . "\n"; chomp(\$line); \$line =~ s/[-?"'{}]/ /g; @words = split(/ /, \$line); \$linecnt++; # FIXME if (\$linecnt != 1) die("Input phrase file must contain only 1 line"); print "inph linecount ". \$linecnt . "\n"; foreach \$word (@words){ ++\$inph{lc(\$word)}; } #print %tf[0]; # compute document frequencies foreach \$word (sort (keys(%granddict))) { \$sum = 0; for \$i (0 .. \$npnr) { #print \$word . "\n"; if (exists \$tf[\$i]{ \$word }) { ++\$sum; } } \$df{\$word} = \$sum; } # Step 3 Compute tf-idf weights \$n = \$npnr + 1; foreach \$word (sort keys %granddict) { for \$i (0 .. \$npnr) {</pre>	<pre>for \$i (0 .. \$npnr - 1){ \$len2 = 0; foreach \$word (sort keys %granddict){ \$len2 += \$weight[\$i]{ \$word }**2; } #print \$word . " len2 " . \$len2 . "\n"; \$len = sqrt(\$len2); foreach \$word (sort keys %granddict){ \$unit[\$i]{ \$word } = \$weight[\$i]{ \$word }/\$len; } ## Normalize input weight so it can be compared with the others foreach \$word (sort keys %granddict){ \$len2 += \$inph_weight{ \$word }; } #print "inph " . \$word . " len2 " . \$len2 . "\n"; \$len = sqrt(\$len2); foreach \$word (sort keys %granddict){ \$inph_unit{ \$word } = \$inph_weight{ \$word }/\$len; } ## Step 5 Compute cosine similarities between input phrase and other phrases \$best = 0; \$best_idx = 0; for \$i (0 .. \$npnr-1) { \$sum = 0; foreach \$word (sort keys %granddict) { \$sum += \$unit[\$i]{ \$word } * \$inph_unit{ \$word }; } \$inph_cosine[\$i] = \$sum; printf "INPH %d %.5f", \$i, \$inph_cosine[\$i]; printf OUTP "INPH %d %.5f\n", \$i, \$inph_cosine[\$i]; if (\$inph_cosine[\$i] > \$best) { \$best = \$inph_cosine[\$i]; \$best_idx = \$i; } printf "\nBest Match %.5f, %d\n", \$best, \$best_idx; printf OUTP "\nBest Match %.5f, %d\n", \$best, \$best_idx; # reopen the data file to get the best line since we didn't store it to save memory open(MYDATA, \$ARGV[0]) or die("Error: cannot open file '\$ARGV[0]'\n"); \$linecnt = 0; while(\$line = <MYDATA>){ print \$line . " linecount: ". \$linecnt . "\n"; if(\$linecnt == \$best_idx){ #print \$line; print OUTP "Best Line: " . \$line . "\n"; \$linecnt++; last; } else { \$linecnt++; } } ## close the files close MYDATA; close INPH;</pre>
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<pre>Libname disc "C:\Users\MHE\Desktop\ActiveCourses\MKT.Eng"; /*the place the files would be put*/ run; /*always create library before importing*/ Proc contents data= disc.Discrim position; Run; /*check the data that is imported*/ /*analysis of description of the data*/ proc means data=disc.Discrim nmeanstdminmax; var outdoor social conservative; run; proc means data=disc.Discrim nmeanstd; class job; var outdoor social conservative; run; proc corr data=disc.Discrim; var outdoor social conservative; run; proc freq data=disc.Discrim; tables job; run; /*discriminant analysis for specific dat"candisc" or proc discrim*/ proc candisc data=disc.Discrim out=discrim_out ; class job; var outdoor social conservative; run; /*figure of discriminant*/ data fakedata; do outdoor = 0to30by1; do social = 5to40by1; do conservative = 0to25by1; output; end; end; end; run; proc discrim data=disc.Discrim testdata=fakedata testout=fake_out out=discrim_out canonical; class job; var outdoor social conservative; run; data plotclass;</pre>	
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<pre>merge fake_out discrim_out; run; proctemplate; define statgraph classify; begingraph; layout overlay; contourplotparm x=Can1 y=Can2 z=_into_ / contourtype=fill nhint = 30 gridded = false; scatterplot x=Can1 y=Can2 / group=job includemissinggroup=false markercharactergroup = job; endlayout; endgraph; end; run; procsgrenderdata = plotclass template = classify; run;</pre>	
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Text mining code in R (Re text mining &Data mining hand book)	Pang 2008 [My General impression: really fruitful but untouched since people in marketing do not have skill, are reluctant to work under burden, are less empirical worker, so it is complete niche to be expert in and publish interesting papers]: predication it will become hot after skill availability for social network within 4 years. (Complexity of programming: recall compiler, not simple as SAS: mental efoort and intelligence), low spillover of computer experts(really nerd ones) to marketing
<pre># preprocessing of the document library(tm) firefox.csv<-read.csv("c:/CDBookSurvey/Comments.csv") firefox <- Corpus(DataframeSource(firefox.csv)) # create corpus for analysis firefoxcopy <- firefox # keep a copy of corpus to use later as a dictionary for stem completion firefox <-tm_map(firefox, tolower) # convert to lower case firefox <- tm_map(firefox, removeNumbers) # remove numbers for (j in 1:length(firefox)) firefox[[j]] <- gsub(""," ",firefox[[j]])# to remove special punctuation but not connect firefox <- tm_map(firefox, removePunctuation)# remove punctuations firefox <- tm_map(firefox, removeWords, stopwords("english")) #remove stop words newstopwords <- c("and", "for", "the", "to", "in", "when", "then", "he", "she", "than", "a", "for", "it", "of", "on", "to", "im") firefox <- tm_map(firefox, removeWords, newstopwords) firefox <- tm_map(firefox, stemDocument)# stem words inspect(firefox[1:10]) firefox <- tm_map(firefox, stemCompletion, dictionary=firefoxcopy) #stem completion inspect(firefoxcopy[1:10]) summary(firefox) myTdm <- TermDocumentMatrix(firefox, control = list(wordLengths=c(1,Inf))) myTdm # printing dtm summery idx <- which(dimnames(myTdm)\$Terms == "alexa") inspect(myTdm[idx+(0:5),1:10]) # look at 5 keywords after the keyword alexa over 10 documents that used for dtm inspect(myTdm[0:20,1:10]) # check items of dtm rownames(myTdm) # write all the keywords you have used findFreqTerms(myTdm, lowfreq=3) #find frequent terms # plot of more frequent words termFrequency <- rowSums(as.matrix(myTdm)) # go over matrix and filtering for drawing a plot termFrequency <- subset(termFrequency, termFrequency>=3) # go for terms that are in text more than 3 times library(ggplot2) # use graphic package to draw plots qplot(names(termFrequency), termFrequency, geom="bar")+ coord_flip() # draw horizontal bar plot barplot(termFrequency, las=2) # draw vertical bar plot findAssocs(myTdm, "love", 0.25)# find words with highest asocation library(wordcloud) # used for importance of the word check m <- as.matrix(myTdm) # convert document term matrix to normal matrix # calculate the frequency of words and sort it descendingly by frequency wordFreq <- sort(rowSums(m), decreasing=TRUE) # word cloud set.seed(375) # to make it reproducibile grayLevels <- gray((wordFreq+10) / (max(wordFreq)+10)) # frequency below 1 is not plotted in the following # random.order=F: frequent words plotted first in the center of the cloud # set colour to: grayLevels or raingbow() to colorful or gray map wordcloud(words=names(wordFreq), freq=wordFreq, min.freq=2, random.order=F, colors=grayLevels) # clustering # remove sparse terms # you can remove sparce terms to avoid being flooded with words myTdm2 <- removeSparseTerms(myTdm, sparse=0.95) m2 <- as.matrix(myTdm2)</pre>	<p>Opinion oriented information seeking opinion mining and sentiment analysis treatment of opinion, sentiment, and subjectivity in text, summarization of evaluative text how product or services perceived classification of comments response of firm after monitoring by modifying their marketingmessages, brand positioning, product development query classification</p> <p>Synonyms: opinion, view, belief, conviction, persuasion, sentiment mean a judgment one holds as true. • opinion implies a conclusion thought out yet open to dispute (each expert seemed to have a different opinion.) • view suggests a subjective opinion (very assertive in stating his views) • belief implies often deliberate acceptance and intellectual assent (a firm belief in her party's platform.) • conviction applies to a firmly and seriously held belief (the conviction that animal life is as sacred as human) • persuasion suggests a belief grounded on assurance (as by evidence) of its truth (was of the persuasion thateverything changes) • sentiment suggests a settled opinion reflective of one's feelings (her feminist sentiments are well-knowing)</p> <p>Dave et al. 2003: "process a set of search results for a given item, generating a list of productattributes (<i>quality, features</i>, etc.) and aggregating <i>opinions</i> about each of them (poor, mixed, good)" Classifying reviews as to their polarity (either positive or negative).</p> <p>Importance of opinion of others while decision making Internet role that allows not acquitants and non professional critic, not colligue and not friend, people we never heard of opinion Bias of rating of users and need correction</p> <p>!! Question answering is another useful area !! Context of the text (other advertising and things in the page that make people recall something) Detection of "flames" (overly-heated or antagonistic language) Summarization for accounting for multiple view point View bettered when includes more information Why someone else is cited in the review (for literally reputation or supporting evidence? computational treatment of affect</p> <p>Subjective judgment of intangible qualities explanation of lack of purchase — e.g., "the design is tacky" or "customer service was condescending" — or even misperceptions —e.g., "updated device drivers aren't available"</p> <p>1. creates condensed versions of individual reviews or adigest of overall consensus points 2. Idea about new product development: Market research from sources such as: Web — newsgroups, individual blogs, and aggregation sites such as Epinions 3. Besides reputation management and public relations, by tracking public viewpoints, one could perform trend prediction in sales or other relevant data 4. focus on what consumer are thinking 5. the issue of how ideas and innovations diffuse involves the question of who is positively or negatively disposed towards whom, and hence who would be more or less receptive to new information transmission from a given source 6. polarity of "ties" between people [54] and how this relates to group cohesion</p> <p>Fundamental technology: Classification and extraction encompasses regression and ranking Examples of problems this method used for: 1. making a decision for a particular phrase or document ("how positive is it?"), 2. ordering a set of texts ("rank these reviews by how positive they are"), 3. giving a single label to an entire document collection ("where on the scale between liberal and conservative do the writings of this author lie?"), and categorizing 4. The relationship between two entities based on textual evidence ("does A</p>

<pre> # cluster of terms/words (come together e.g. couple of tweets on text mining analysis, and couple of tweets on job vacancies in PhD in different clusters) distMatrix <- dist(scale(m2)) # calculate distance between terms after scaling fit <- hclust(distMatrix, method="ward") # clustering agglomeration method is set to ward: icreased variance when two clusters are merged; other options are: single linkage, complete linkage, average linkage, median and centroid plot(fit) # cut tree into 10 clusters rect.hclust(fit, k=10) # cut into 10 clusters (groups <- cutree(fit, k=10)) # clustering using k-min of documents # transpose the matrix to cluster documents (tweets) m3 <- t(m2) # take value of matrix as numeric & transpose to document term # set a fixed random seed set.seed(122) # to produce the clustering result # k-means clustering of tweets k <- 3 # 8 clusters kmeansResult <- kmeans(m3, k) # cluster centers round(kmeansResult\$centers, digits=3) # popular words in cluster and center # check k mean cluster by top 3 words for (i in 1:k) { cat(paste("cluster ", i, ": ", sep="")) s <- sort(kmeansResult\$centers[i,], decreasing=T) cat(names(s)[1:3], "\n") # print the tweets of every cluster # print(rdmTweets[which(kmeansResult\$cluster==i)]) } library(Rgraphviz)# to use for cluster assowciation matrix plot(myTdm, terms = findFreqTerms(myTdm, lowfreq = 1)[1:20], corThreshold = 0) library(fpc)#draw cluster based on matrix plotcluster(m3, kmeansResult\$cluster) library(fpc) # clustering with Partitioning Around Medoids (PAM): (representative objects) more robust to noise and outliers than k-means clustering # partitioning around medoids with estimation of number of clusters pamResult <- pamk(m3, metric = "manhattan") # estimate number of optimal clusters # number of clusters identified (k <- pamResult\$nc) pamResult <- pamResult\$pamobject # print cluster medoids for (i in 1:k) { cat(paste("cluster", i, ": ")) cat(colnames(pamResult\$medoids)[which(pamResult\$medoids[i,]==1)], "\n") #print tweets in cluster i # print(rdmTweets[pamResult\$clustering==i]) } # plot clustering result layout(matrix(c(1,2),2,1)) # set to two graphs per page plot(pamResult, color=F, labels=4, lines=0, cex=.8, col.clus=1, col.p=pamResult\$clustering) layout(matrix(1)) # change back to one graph per page #create social network of terms termDocMatrix<-m2 termDocMatrix[1:5,1:5] # check Tdm # change it to a Boolean matrix termDocMatrix[termDocMatrix>=1] <- 1 # transform into a term-term adjacency matrix termMatrix <- termDocMatrix %*% t(termDocMatrix) # %*% product of two matrices # inspect terms numbered 5 to 7 termMatrix[5:7,5:7] library(igraph) # build a graph from the above matrix g <- graph.adjacency(termMatrix, weighted=T, mode = "undirected") # remove loops g <- simplify(g) # set labels and degrees of vertices V(g)\$label <- V(g)\$name V(g)\$degree <- degree(g) # set seed to make the layout reproducible set.seed(3952) layout1 <- layout.fruchterman.reingold(g) plot(g, layout=layout1) #dynamically rearranged layout get detail by running ?igraph::layout plot(g, layout=layout.kamada.kawai) tkplot(g, layout=layout.kamada.kawai)#extremely interesting graph creation </pre>	<pre> approve of B's actions?"). 1. extraction problems (e.g., retrieving opinions on various features of a laptop) are often solved by casting many sub-problems as classification problems (e.g., given a text span, determine whether it expresses any opinion at all). 2. extraction is often a means to the further goal of providing effective summaries of the extracted information to users (combine information mined from multiple subjective text segments into a suitable summary) Problem formulation and key concepts: 1. Sentiment polarity and degrees of positivity (locate its position on the continuum between these two polarities): sentiment-related classification/regression/ranking binary categorization, multi-class categorization, regression, and/or ranking 2. Related categories: extract info on why reviewer liked or disliked the product “pros & cons” 3. Rating inference (ordinal regression): multi-class text categorization problem Predicting degree of positivity provides more fine-grained rating information; ordinal regression; mediocre& neutral that is not strong feeling of good or bad (different from “lack of opinion”): reduce retaliation of seller, yet is perceived neg 4. Agreement detection: two text shall receive same or differing sentiment-related labels based on relationship b/w pairs? • identification of subjectivity versus objectivity (effects of adjective orientation and gradability on sentence subjectivity: wiebe et. al): roots in studies in genre classification • joint topic sentiment analysis: whether the document topic is related to subject of interest • view point and perspective: more about attitude, n-ary classification • various affect types six “universal” emotions: anger, disgust, fear, happiness, sadness, and surprise • style analysis of the text • feature vector or other representation that makes its most salient and important features 1.binary versus frequency based 2. Position at the beginning or end of document? (trigram and hierarchy) 3. Part of the speech: e.g. adjective 4. Syntax: e.g. modeling valence shifters such as negation, intensifiers, and diminishers 5. Negation “not, don’t”; “I don’t like deadlines”, the token “like” is converted into the new token “like-NOT”. Controversy when No does not negate: “No wonder this is considered one of the best”; problem of negation is more salient in sarcasm. E.g. “avoid” 6. Topic oriented features: PARTY will win”, “go PARTY again”, and “OTHER will win” Approaches(Machine learning methods, mostly data mining methods (genetic, clustering, regression, ..., but predictive usage), all are statistic methods; Not pure mechanical human should input into process): • mapping a given piece of text, such as a document, paragraph, or sentence, to a label drawn from a pre-specified finite set or to a real number • The impact of labeled data: Maximize entropy method • Domain consideration: “unpredictable” is a positive description for a movie plot but a negative description for a car’s steering abilities; 1. Use domain specific classifier • Topic and subtopic: 1. on-topic text in the description or off topic , 2. Multiple topics Unsupervised approaches: • Classification using clustering technique • Frequency of occurrence, prior polarity, coocurance in the certain context • Bootstrapping: use the output of an available initial classifier to create labeled data, to which a supervised learning algorithm may be applied • Classification based on relationship between documents: e.g. relationship between subdocuments or sentences. Degree of continuity (story telling), graph based techniques. “Respond to”: when people respond to each other and that relationship which mostly has been antagonistic [addressing other person] • Relationship between classes in contrast to multi-class categorization: 5-star is much similar to 4-star than 2-star. • Discourse structure (overriding previous 4 line by simple sentence): e.g. [they] act wacky as hell...the ninth floor of hell...a cheap [beep] movie...The plotis such a mess that it’s terrible. But I loved it.[incorporating location of information is very important] • Identifying opinion holder, a person who does comparison and reveals her preference Language models: topic relevancy, sentiment relevancy: • difference in perspective upon the Kullback-Leibler(KL) divergence between posterior distributions induced from document collection pairs, and discover that the KL divergence between different aspects is an order of magnitude smaller than that between different topics. • Probabilistic latent semantic analysis (PLSA) or latent Dirichlet allocation(LDA) can also be cast as language-modeling work The basic idea is to infer language models that correspond to unobserved “factors” in the data, with the hope that the factors that are learned represent topics or sentiment categories. Oder of information algorithm: Another way of capturing discourse structure information in documents is to model the global sentiment of a document as a trajectory of local sentiments. Using sentiment flow as a sequential model to represent an opinionated document. More specifically, each sentence in the document receives a local sentiment score from an isotonic-conditional-random-field-based sentence level predictor. The sentiment flow is defined as a function $h: [0, 1) \rightarrow O$(the ordinal set), where the interval $[(t - 1)/n, t/n)$ is mapped to the label of the t-th sentence in a document with n sentences. </pre>
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<pre>pdf("term-network.pdf") # put terms plot in a pdf file plot(g, layout=layout.fruchterman.reingold) dev.off() # size of plot's term according to the degree: important terms stand out # set the width and transparency of edges based on their weights # vertices and edges are accessed with V() and E() # rgb(red, green, blue,alpha) defines a color with an alpha transparency V(g)\$label.cex <- 2.2 * V(g)\$degree / max(V(g)\$degree)+.2 V(g)\$label.color <- rgb(0, 0, .2, .8) V(g)\$frame.color <- NA egam <- (log(E(g)\$weight)+.4) / max(log(E(g)\$weight)+.4) E(g)\$color <- rgb(.5, .5, 0, egam) E(g)\$width <- egam # plot the graph in layout1 plot(g, layout=layout1) #build network of documents (tweets) first phase # remove "r", "data" and "mining" most used if they make the document crowded # idx <- which(dimnames(termDocMatrix)\$Terms %in% c("r", "data", "mining")) #M <- termDocMatrix[-idx,] # remove terms from matrix M<-termDocMatrix # since I did not wanted to remove anything # build a tweet-tweet adjacency matrix tweetMatrix <- t(M) %%% M library(igraph) g <- graph.adjacency(tweetMatrix, weighted=T, mode = "undirected") V(g)\$degree <- degree(g) g <- simplify(g) # set labels of vertices to tweet IDs V(g)\$label <- V(g)\$name V(g)\$label.cex <- 1 V(g)\$label.color <- rgb(.4, 0, 0, .7) V(g)\$size <- 2 V(g)\$frame.color <- NA barplot(table(V(g)\$degree)) # check degree distribution of vertices #build network of documents (tweets) second phase idx <- V(g)\$degree == 0 V(g)\$label.color[idx] <- rgb(0, 0, .3, .7) # set based on degree # set labels to the IDs and the first 10 characters of tweets # limit to the first 20 character of every tweet # label of each set to tweet ID so that graph would not be overcrowded # set color and width of edges based on their weights #V(g)\$label[idx] <- paste(V(g)\$name[idx], substr(df\$text[idx], 1, 20), sep=" ") egam <- (log(E(g)\$weight)+.2) / max(log(E(g)\$weight)+.2) E(g)\$color <- rgb(.5, .5, 0, egam) E(g)\$width <- egam set.seed(3152) layout2 <- layout.fruchterman.reingold(g) plot(g, layout=layout2) # remove isolated vertices and draw again g2 <- delete.vertices(g, V(g)[degree(g)==0]) plot(g, layout=layout2) # remove edges with low degree and draw again g3 <- delete.edges(g, E(g)[E(g)\$weight <= 1]) g3 <- delete.vertices(g3, V(g3)[degree(g3) == 0]) plot(g3, layout=layout.fruchterman.reingold) # look at specific clique: considerably connected {replacement for dftext inspect(firefox[c(15,16)]) #graph g directly from termDocMatrix # create a graph g <- graph.incidence(termDocMatrix, mode=c("all")) # get index for term vertices and tweet vertices nTerms <- nrow(M) nDocs <- ncol(M) idx.terms <- 1:nTerms idx.docs <- (nTerms+1):(nTerms+nDocs) # set colors and sizes for vertices V(g)\$degree <- degree(g) V(g)\$color[idx.terms] <- rgb(0, 1, 0, .5) V(g)\$size[idx.terms] <- 6</pre>	<p>The flow is then smoothed out through convolution with a smoothing kernel. Finally, the distances between two flows (e.g., Lp distance between the two smoothed, continuous functions) should reflect, to some degree, the distances between global sentiments.</p> <p>Sentiment without action is the ruin of the soul. — Edward Abbey Romance should never begin with sentiment. It should begin with science and end with settlement. — Oscar Wilde, An Ideal Husband</p> <p>Challenges:</p> <ol style="list-style-type: none"> determining which documents/portion are topically relevant to an opinion-oriented query Quotation saying that it is from someone else Summarizing the sentiment: Visualizing: <p>(a) aggregation of “votes” that may be registered on different scales (e.g., one reviewer uses a star system, but another uses letter grades)</p> <p>(b) selective highlighting of some opinions</p> <p>(c) representation of points of disagreement and points of consensus</p> <p>(d) identification of communities of opinion holders</p> <p>(e) accounting for different levels of authority among opinion holders</p> <p>4. sentiment polarity text-classification: positive or negative: the inference and indirect sarcasm sentence may not have negative word but imply negative: my explanation: 1. not apply here since else people will have hard time understanding 2. It is limited domain with limited words 3. 20-80% as far as predicts sales and normal person understands it, it is good. 4. modern international people do not speak complicated (the targeted customer of this product), showing off their literature</p> <p>5. hypocritical people say something like (I don’t want to talk about this), but they actually do</p> <p>6. categorization of fact vs. opinion</p> <p>7. previously loved but now hate (IMO: multiplication of positive and negative sense could work, prior and posterior; title and stars could be helpful in this sense)</p> <p>8. abbreviations</p> <p>9. product reviewer homophily with me in term of language conditional on I care (some people don’t care)</p> <p>10. Sentiment and subjectivity are quite context-sensitive, and, at a coarser granularity, quite domain dependent. even the exact same expression can indicate different sentiment in different domains. (Go read the book in movie means negative sentiment but in book means good): IMO: complementary product and substitute product mentioning</p> <p>11. the order in which different opinions are presented can result in a completely opposite overall sentiment polarity (in contrast to discourse analysis)</p> <p>12. Course changing words such as “However”, “But”: my idea it does not change course completely but adds a second vector of negativity (like hygiene parameter of working and incentive): and consider the asymmetric answer of humans to negative and positive information</p> <p>13. Order dependence of comparisons: Comparison words finding & order analysis. Two category of words [+ vs. -] and then greater than or equal to: substitute products, complementary products.</p> <p>14. certainty vs. uncertainty (words : maybe, vs. must, will)</p> <p>15. Past, present, future tense of the word (may not be really precise)</p> <p>16. context that may make the difference for example stock price rise is a good news or bad?</p> <p>17. objective information such as “long battery life”² is often used to help determine the overall sentiment& whether this objective information is good or bad</p> <p>18. The effect of specific words such as only: “the battery lasts 2 hours” vs “the battery only lasts 2 hours”& proximity of the meaning in the context: e.g. (“This laptop only costs \$399”: how people judge attributes</p> <p>19. determining degree of positivity: “The new model is more expensive than the old one” or “I prefer the new model to the old model”</p> <p>20. identification of subjectivity versus objectivity (effects of adjective orientation and gradability on sentence subjectivity: wiebe et. al): roots in studies in genre classification</p> <p>21. various affect types six “universal” emotions: anger, disgust, fear, happiness, sadness, and surprise</p> <p>22. Style analysis of the text and characteristics of a person</p> <p>23. feature vector or other representation that makes its most salient and important features 1. binary versus frequency based 2. Position at the beginning or end of document? (trigram and hierarchy) 3. Part of the speech: e.g. adjective 4. Syntax: e.g. modeling valence shifters such as negation, intensifiers, and diminishers 5. Negation “not, don’t”; “I don’t like deadlines”, the token “like” is converted into the new token “like-NOT”. Controversy when No does not negate: “No wonder this is considered one of the best”.; problem of negation is more salient in sarcasm. E.g. “avoid” 6. Topic oriented features: PARTY will win”, “go PARTY again”, and “OTHER will win”</p>
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<pre> V(g)\$color[idx.docs] <- rgb(1, 0, 0, .4) V(g)\$size[idx.docs] <- 4 V(g)\$frame.color <- NA # set vertex labels and their colors and sizes V(g)\$label <- V(g)\$name V(g)\$label.color <- rgb(0, 0, 0, 0.5) V(g)\$label.cex <- 1.4*V(g)\$degree/max(V(g)\$degree) + 1 # set edge width and color E(g)\$width <- .3 E(g)\$color <- rgb(.5, .5, 0, .3) set.seed(958)#5365, 227 plot(g, layout=layout.fruchterman.reingold) # returns all vertices of "love" # if node does not exist returns "invalid vertex name" V(g)[nei("love")] V(g)[neighborhood(g, order=1, "love")[[1]]]# alternative way of geting vertices #check which vertices include all three elements "thank", "perfect", "love" (rdmVertices <- V(g)[nei("love") & nei("perfect") & nei("thank")]) inspect(firefox[as.numeric(rdmVertices\$label)])# check content of the doc that includes these three terms # remove three words to see the relationship with doc with other words idx <- which(V(g)\$name %in% c("love", "perfect", "thank")) g2 <- delete.vertices(g, V(g)[idx-1]) g2 <- delete.vertices(g2, V(g2)[degree(g2)==0]) set.seed(209) plot(g2, layout=layout.fruchterman.reingold) </pre>	<p>My research</p> <ol style="list-style-type: none"> 1 Time varying effect of comments on sales: comment window 2. product category and attributes 3. substitute and complimentary product mentioning 4. use Google keyword for relevant keywords and analysis 5. Use other available data sources as well to simplify this process of wording. There are many websites that you can capture content and take the intersection of sets or do the weighting. 5. product attribute such as (log in, theme, version, working, perceived risk) 6. Thesaurus through term-term matrix 7. Download source 8. Politeness “using f words, Damn) 9. Forward looking or myopic (hedonistic: emotion or utilitarian: attributes, profit,logic) 10. Whether the person asked the question, provides fact, or opinion 10. When model built run on other products as well and check the result [Cross category analysis: competing structure <p>Part that I skipped was summarization/ broader implication chapter, since I thought it did not really provide more information.</p>
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