An Analysis

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1	Descriptive Statistics	
_	etHdata(support) # Use Hmisc/getHdata to get dataset from VU DataSets — subset(support, select=c(age,sex,race,edu,income,hospdead,slos,dzgromeanbp,hrt))	
1 a	atex(describe(d), file='')	

$\begin{array}{cc} & d \\ 1000 & Observations \end{array}$

age:Age										unistratistriidillillillillillillillininissa
n missing unique 1000 0 9 10 18.41 19.	70 1		.05 33.76				.50 .90	$ \begin{array}{c} .75 \\ 74.50 \end{array} $	$ \begin{array}{c} .90 \\ 81.87 \end{array} $.95 86.00
	71 100.13									
sex										
n missing unique 1000 0	1e									
female (438, 44%), male (56	2, 56%)									
race										
n missing unique	e 5									
white black asian Frequency 781 157 9		spanic 36								
<u>%</u> 78 16 1	1	4								
edu: Years of Education	on									
n missing unique		Mean	.05	.10 .25	.50	$.75 \\ 14$.90			
798 $20\overline{2}$ 29 lowest: 0 1 2 3 4, hi	0.97	11.78 21 22 24	6 30	8 10) 12	14	16	18		
income										
n missing unique	e 1									
under \$11k (309, 47%), \$11->\$50k (75, 12%)	\$25k (161	, 25%), \$	25-\$50	k (106,	16%)					

```
hospdead: Death in Hospital
             missing unique
                                        Sum
253
slos: Days from Study Entry to Discharge
                                        Mean .05
                                                      \begin{array}{cc} .10 & .25 \\ 4 & 6 \end{array}
        n missing unique Info
          3 4 5 6 7, highest: 145 164 202 236 241
lowest :
dzgroup
           missing unique
0 8
     1000
          ARF/MOSF w/Sepsis COPD CHF Cirrhosis Coma Colon Cancer Lung Cancer 391 116 143 55 60 49 100 39 12 14 6 6 6 5 10
Frequency
meanbp: Mean Arterial Blood Pressure Day 3
                                                                                                      \begin{array}{ccc} \text{missing} & \text{unique} & \text{Info} \\ 0 & 122 & 1 \end{array}
                                        Mean
84.98
                                                05 \\ 47.00
                                                         55.00
                                                                  64.75
                                                                                   107.00
                                                                           78.00
                                                                                            120.00 128.05
          0 20 27 30 32, highest: 155 158 161 162 180
lowest :
hrt: Heart Rate Day 3
                                                                                                n missing unique Info 0 0 124 1
                                        Mean
97.87
                                               05 \\ 54.0
          0 11 30 35 36, highest: 189 193 199 232 300
    Race is reduced to three levels (white, black, OTHER) because of low frequencies in other levels (minimum
relative frequency set to 0.05).
```

d	\leftarrow	upData(d,	٦
		<pre>race = combine.levels(race, minlev = 0.05))</pre>	

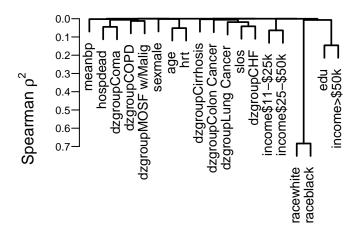
Input object size:	107336 bytes;	10 variables	1000 observations
Modified variable	race		
New object size:	107216 bytes;	10 variables	1000 observations

2 Redundancy Analysis and Variable Interrelationships

```
Redundancy Analysis
redun(formula = \simage + sex + race + edu + income + dzgroup +
    meanbp + hrt, data = d)
n: 617 p: 8
                nk: 3
Number of NAs:
                 383
Frequencies of Missing Values Due to Each Variable
            sex
                  race
                            edu
                                 income dzgroup meanbp
                                                              hrt
    age
      0
              0
                             202
                                     349
Transformation of target variables forced to be linear
R^2 cutoff: 0.9 Type: ordinary
R^2 with which each variable can be predicted from all other variables:
```

```
income dzgroup
                                edu
    age
              sex
                      race
                                                         meanbp
                                                                      hrt.
                                       0.339
                                                          0.067
  0.196
           0.088
                     0.120
                              0.284
                                                 0.253
                                                                   0.163
No redundant variables
```

```
# Alternative: redun(\sim., data=subset(d, select=-c(hospdead,slos)))
```



Note that the clustering of black with white is not interesting; this just means that these are mutually exclusive higher frequency categories, causing them to be negatively correlated.

3 Logistic Regression Model

Here we fit a tentative binary logistic regression model. The coefficients are not very useful so they are not printed. Note: the symbolic section reference below was created by the following R comment:

see Section (*\ref{descStats}*) for descriptive statistics

The label was defined in an earlier section using

\section{Descriptive Statistics}\label{descStats}

```
require(rms)
```

Logistic Regression Model

```
lrm(formula = hospdead ~ rcs(age, 4) + sex + race + dzgroup +
rcs(meanbp, 5), data = d)
```

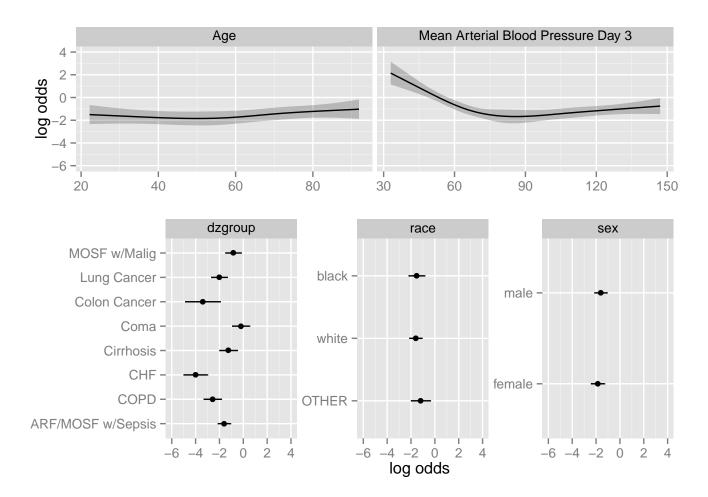
Frequencies of Missing Values Due to Each Variable

			Model Likelihood		Discrimination		Rank Discrim.	
			Ratio Test		Indexes		Indexes	
	Obs	995	LR χ^2	245.83	R^2	0.323	C	0.800
	0			17		1.605		0.601
	1	251	$\Pr(>\chi^2)$	< 0.0001	g_r	4.980	γ	0.602
	$\max \left \frac{\partial \log L}{\partial \beta} \right $	$1\!\times\!10^{-9}$			g_p	0.228	$ au_a$	0.227
	1				Brier	0.144		
latex(anov	va(f), whe	re='h',	file='	') # 0	can al	so try	where	= ' h t b p '

Table 1: Wald Statistics for hospdead

	χ^2	d.f.	P
age	7.12	3	0.0683
Nonlinear	2.91	2	0.2338
sex	2.16	1	0.1413
race	1.38	2	0.5005
dzgroup	78.77	7	< 0.0001
meanbp	65.62	4	< 0.0001
Nonlinear	48.11	3	< 0.0001
TOTAL NONLINEAR	50.15	5	< 0.0001
TOTAL	151.71	17	< 0.0001

ggplot(Predict(f), sepdiscrete='vertical')



4 Test Calculations

$$x \leftarrow 3; y \leftarrow 2$$
 if $(x \leq y)$ 'this' else 'that'

[1] "that"

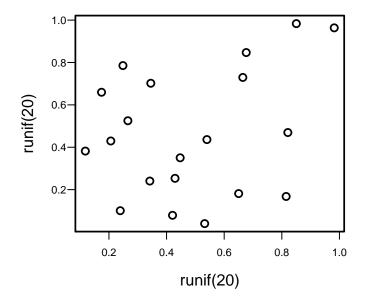
if(y > x) 'that' else 'this'

[1] "this"

 $x^{\wedge}y$

[1] 9

plot(runif(20),runif(20))



5 Computing Environment

These analyses were done using the following versions of R³, the operating system, and add-on packages Hmisc¹, rms², and others:

toLatex(sessionInfo(), locale=FALSE)

- R version 3.2.2 (2015-08-14), x86_64-pc-linux-gnu
- Base packages: base, datasets, graphics, grDevices, grid, methods, stats, utils
- Other packages: Formula 1.2-1, ggplot2 1.0.1, Hmisc 3.17-0, knitr 1.11, lattice 0.20-33, rms 4.4-1, SparseM 1.7, survival 2.38-3
- Loaded via a namespace (and not attached): acepack 1.3-3.3, cluster 2.0.3, codetools 0.2-14, colorspace 1.2-6, digest 0.6.8, evaluate 0.8, foreign 0.8-66, formatR 1.2.1, gridExtra 2.0.0, gtable 0.1.2, labeling 0.3, latticeExtra 0.6-26, magrittr 1.5, MASS 7.3-44, Matrix 1.2-2, MatrixModels 0.4-1, multcomp 1.4-1, munsell 0.4.2, mvtnorm 1.0-3, nlme 3.1-122, nnet 7.3-11, plyr 1.8.3, polspline 1.1.12, proto 0.3-10, quantreg 5.19, RColorBrewer 1.1-2, Rcpp 0.12.1, reshape2 1.4.1, rpart 4.1-10, sandwich 2.3-4, scales 0.3.0, splines 3.2.2, stringi 0.5-5, stringr 1.0.0, TH.data 1.0-6, tools 3.2.2, zoo 1.7-12

References

- [1] Frank E. Harrell. Hmisc: A package of miscellaneous R functions. 2015. URL: http://biostat.mc.vanderbilt.edu/Hmisc.
- [2] Frank E. Harrell. rms: R functions for biostatistical/epidemiologic modeling, testing, estimation, validation, graphics, prediction, and typesetting by storing enhanced model design attributes in the fit. Implements methods in Regression Modeling Strategies, 2nd edition. 2015. URL: http://biostat.mc.vanderbilt.edu/rms.
- [3] R Development Core Team. R: A Language and Environment for Statistical Computing. R Foundation for Statistical Computing. Vienna, Austria, 2015. URL: http://www.R-project.org.