

ORACLE

Session 6: Oracle Machine Learning for R Statistics Engine

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Oracle R Enterprise Statistics Engine

Example Features

Special Functions

- Gamma function
- Natural logarithm of the Gamma function
- Digamma function
- Trigamma function
- Error function
- Complementary error function

Tests

- Chi-square, McNemar, Bowker
- Simple and weighted kappas
- Cochran-Mantel-Haenzel correlation
- Cramer's V
- Binomial, KS, t, F, Wilcox

Base SAS equivalents

- Freq, Summary, Sort
- Rank, Corr, Univariate

Density, Probability, and Quantile Functions

- Beta distribution
- Binomial distribution
- Cauchy distribution
- Chi-square distribution
- Exponential distribution
- F-distribution
- Gamma distribution
- Geometric distribution
- Log Normal distribution
- Logistic distribution

- Negative Binomial distribution
- Normal distribution
- Poisson distribution
- Sign Rank distribution
- Student's t distribution
- Uniform distribution
- Weibull distribution
- Density Function
- Probability Function
- Quantile

In-Database SQL-based Statistical Functions

Supporting OML4R scalability through in-database execution

[Native Oracle Database Statistical Functions](#)

[SQL for Analysis and Reporting](#)

[DMBS_STATS_FUNCS Reference](#)

[Descriptive Statistics](#)

[Hypothesis Testing - Parametric Tests](#)

[Crosstab Statistics](#)

[Hypothesis Testing - Non-Parametric Tests](#)

[Non-Parametric Correlation](#)

Statistical Tests – Examples

```
ore.create(iris,table="IRIS_TABLE")
IRIS_TABLE$PETALBINS=ifelse(IRIS_TABLE$Petal.Length < 2, 1, 2)
# Binomial Test
binom.test(IRIS_TABLE$PETALBINS)

# Chi Square Test
chisq.test(IRIS_TABLE$PETALBINS)

# One sample K-S Test for given probabilities
ks.test(IRIS_TABLE$Petal.Length, "pexp", rate=4)

# Two sample K S Test
ks.test(IRIS_TABLE$Petal.Length, IRIS_TABLE$Sepal.Length)

# T-test with different alternate hypothesis possibilities */
t.test(IRIS_TABLE$Petal.Length, alternative="two.sided", mu=0, conf.level=0.9)

# F test to compare variances
var.test(IRIS_TABLE$Petal.Length, IRIS_TABLE$Sepal.Length, ratio=0.75,
         alternative="two.sided", conf.level=0.9)
```

Statistical Tests – Results

```
> IRIS_TABLE$PETALBINS=ifelse(IRIS_TABLE$Petal.Length < 2, 1, 2)
> # Binomial Test
> binom.test(IRIS_TABLE$PETALBINS)

Exact binomial test

data: x
number of successes = 50, number of trials = 150, p-value =
5.448e-05
alternative hypothesis: true probability of success is not equal to 0.5
95 percent confidence interval:
0.2585564 0.4148430
sample estimates:
probability of success
0.3333333

>
> # Chi Square Test
> chisq.test(IRIS_TABLE$PETALBINS)

Chi-squared test for given probabilities

data: x
X-squared = 16.6667, df = 1, p-value = 4.456e-05

>
> # One sample K-S Test for given probabilities
> ks.test(IRIS_TABLE$Petal.Length, "pexp", rate=4)

One-sample Kolmogorov-Smirnov test

data: IRIS_TABLE$Petal.Length
D = 0.975, p-value < 2.2e-16
alternative hypothesis: two-sided
```

```
> # Two sample K S Test
> ks.test(IRIS_TABLE$Petal.Length, IRIS_TABLE$Sepal.Length)

Two-sample Kolmogorov-Smirnov test

data: IRIS_TABLE$Petal.Length and IRIS_TABLE$Sepal.Length
D = 0.56, p-value < 2.2e-16
alternative hypothesis: two-sided

>
> # T-test with different alternate hypothesis possibilities */
> t.test(IRIS_TABLE$Petal.Length, alternative="two.sided", mu=0,
conf.level=0.9)

One Sample t-test

data: IRIS_TABLE$Petal.Length
t = 26.0726, df = 149, p-value < 2.2e-16
alternative hypothesis: true mean is not equal to 0
90 percent confidence interval:
3.519434 3.996566
sample estimates:
mean of x
3.758

>
> # F test to compare variances
> var.test(IRIS_TABLE$Petal.Length, IRIS_TABLE$Sepal.Length, ratio=0.75,
+           alternative="two.sided", conf.level=0.9)

F test to compare two variances

data: IRIS_TABLE$Petal.Length and IRIS_TABLE$Sepal.Length
F = 6.0596, num df = 149, denom df = 149, p-value < 2.2e-16
alternative hypothesis: true ratio of variances is not equal to 0.75
90 percent confidence interval:
3.468071 5.955583
sample estimates:
ratio of variances
4.54471
```

Oracle R Enterprise “PROCs”

SUMMARY / MEANS

RANK

SORT

CROSSTAB

FREQ

CORR

UNIVARIATE

ore.summary

```
ore.summary(data, var, stats = c("n", "mean", "min", "max"),
            class = NULL, types = NULL, ways = NULL, weight
            = NULL,
            order = NULL, maxid = NULL, minid = NULL, mu =
            0,
            no.type = FALSE, no.freq = FALSE)
```

Provides descriptive statistics and extensive analysis of columns in an ore.frame with flexible row aggregations

Statistics, e.g.,

- n, count/cnt, num miss, mean/avg, min, max, sum, sumwgt
- corrected and uncorrected sum of squares, range of values, stddev, stderr, variance
- t-test for testing hypothesis that the population mean is zero
- kurtosis, skew, Coefficient of Variation
- quantiles: p1,p5,p10,p25,p50,p75,p90,p95,p99,qrange
- 1 and 2 sided Confidence Limits for the mean: clm, rclm, lclm
- extreme value tagging, ...and others

Simple syntax abstracting complex SQL queries

ore.summary – Parameters

data: ore.frame on which to compute descriptive statistics

var: vector of column names *on which to apply statistics functions*

stats: list of statistics functions available to be applied on **var** columns - e.g.,
mean,min,max,cnt,n,nmiss,css,uss,cv,sum,sumwgt,range, stddev, stderr, var, t, probt, kurt, skew, p1, p5, p10, p25, p50, p75, p90, p95, p99, qrange, lclm, rclm, clm, mode

class: vector of column names *to aggregate (i.e., SQL group by)*

types: list of character string vectors specifying the combinations of column names in **class** within which the aggregations will be executed in the returning summary

ways: integer vector with each value indicating number of columns in **class** used to generate types

weight: column name whose numeric values provide a multiplicative factor for **var** columns

order: string vector specifying sorting criteria: ‘freq’ or ‘-freq’, ‘type’ or ‘-type’, ‘class’ or ‘-class’

maxid, minid: for each group optionally list max/min value from other columns in

mu: single number or a vector whose elements correspond to each value in **var** to supply additional numeric parameters for some statistics

Examples

```
IRIS <- ore.push(iris)

ore.summary(IRIS, c("Sepal.Length", "Petal.Length"))

ore.summary(IRIS, c("Sepal.Length", "Petal.Length"), c("mean", "std", "p10"), class="Species")

ore.summary(IRIS, list(c("Sepal.Length", "Petal.Length"), "Sepal.Width"), c(AVG="mean", "std"),
class="Species")

ore.summary(IRIS, c("Sepal.Length", "Petal.Length"), c("mean", "std"), class="Species",
weight="Sepal.Width")

ore.summary(IRIS, c("Sepal.Length", "Petal.Length"), c("mean", "std"),
class=c("Species", "Petal.Width"), types=list("Species", c("Species", "Petal.Width")),
order=c("type", "-freq", "class"))

ore.summary(IRIS, c("Sepal.Length", "Petal.Length"), c("mean", "std"),
class=c("Species", "Petal.Width"), ways=1, order=c("type", "-freq", "class"))

ore.summary(IRIS, c("Sepal.Length", "Petal.Length"), c("mean", "prt"), class="Species", mu=c(5.8, 3.7))

ore.summary(IRIS, c("Sepal.Length", "Petal.Length"), "mean",
class="Species", maxid=c(Sepal.Length="Sepal.Width", Petal.Length="Petal.Width"))

ore.summary(IRIS, c("Sepal.Length", "Sepal.Width"), "max",
maxid=c(Sepal.Length="Species", Sepal.Width="Species"))
```

ore.summary – Results

```
..  
R> IRIS <- ore.push(iris)  
R> ore.summary(IRIS, c("Sepal.Length", "Petal.Length"))  
  FREQ N(Sepal.Length) N(Petal.Length) MEAN(Sepal.Length) MEAN(Petal.Length) MIN(Sepal.Length) MIN(Petal.Length) MAX(Sepal.Length)  
1 150          150          150        5.843333        3.758          4.3           1          7.9  
  MAX(Petal.Length)  
1          6.9  
R> ore.summary(IRIS, c("Sepal.Length", "Petal.Length"), c("mean", "std", "p10"), class="Species")  
  Species FREQ TYPE MEAN(Sepal.Length) MEAN(Petal.Length) STD(Sepal.Length) STD(Petal.Length) P10(Sepal.Length)  
1  setosa   50    0     5.006000      1.462      0.3524897      0.1736640      4.59  
2 versicolor 50    0     5.936000      4.260      0.5161711      0.4699110      5.38  
3 virginica  50    0     6.588000      5.552      0.6358796      0.5518947      5.80  
4    <NA>   150    1     5.843333      3.758      0.8280661      1.7652982      4.80  
  P10(Petal.Length)  
1          1.30  
2          3.59  
3          4.90  
4          1.40  
R> ore.summary(IRIS, list(c("Sepal.Length", "Petal.Length"), "Sepal.Width"), c(AVG="mean", "std"), class="Species")  
  Species FREQ TYPE AVG(Sepal.Length) AVG(Petal.Length)      STD  
1  setosa   50    0     5.006000      1.462  0.3790644  
2 versicolor 50    0     5.936000      4.260  0.3137983  
3 virginica  50    0     6.588000      5.552  0.3224966  
4    <NA>   150    1     5.843333      3.758  0.4358663
```

```

```
R> ore.summary(iris, c("Sepal.Length", "Petal.Length"), c("mean", "std"), class="Species", weight="Sepal.Width")
 Species FREQ TYPE MEAN(Sepal.Length) MEAN(Petal.Length) STD(Sepal.Length) STD(Petal.Length)
1 setosa 50 0 5.034384 1.465344 0.3517111 0.1734107
2 versicolor 50 0 5.966137 4.289242 0.5095037 0.4548644
3 virginica 50 0 6.618897 5.575521 0.6276732 0.5454365
4 <NA> 150 1 5.829546 3.650894 0.8343801 1.8126149
R> ore.summary(iris, c("Sepal.Length", "Petal.Length"), c("mean", "std"), class=c("Species", "Petal.Width"), types=list("Species", c("Species", "Petal.Width")), order=c("type", "-freq", "class"))
 Species Petal.Width FREQ TYPE MEAN(Sepal.Length) MEAN(Petal.Length) STD(Sepal.Length) STD(Petal.Length)
1 setosa 0.2 29 0 4.972414 1.444828 0.3544579 0.17234974
2 versicolor 1.3 13 0 5.884615 4.176923 0.3804518 0.25869495
3 virginica 1.8 11 0 6.445455 5.381818 0.4568668 0.51926522
4 versicolor 1.5 10 0 6.190000 4.580000 0.4724640 0.20976177
5 virginica 2.3 8 0 6.912500 5.700000 0.5436320 0.59761430
6 setosa 0.3 7 0 4.971429 1.428571 0.3988077 0.13801311
7 setosa 0.4 7 0 5.300000 1.571429 0.2449490 0.18898224
8 versicolor 1.0 7 0 5.414286 3.628571 0.4450789 0.31997024
9 versicolor 1.4 7 0 6.357143 4.500000 0.6133437 0.30550505
10 virginica 2.0 6 0 6.650000 5.550000 0.9710819 0.78676553
11 virginica 2.1 6 0 6.916667 5.783333 0.4070217 0.43550737
12 setosa 0.1 5 0 4.820000 1.380000 0.3271085 0.16431677
13 versicolor 1.2 5 0 5.780000 4.240000 0.2167948 0.32093613
14 virginica 1.9 5 0 6.340000 5.320000 0.6542171 0.44944410
15 versicolor 1.1 3 0 5.400000 3.566667 0.2645751 0.49328829
16 versicolor 1.6 3 0 6.100000 4.766667 0.1732051 0.30550505
17 virginica 2.2 3 0 6.866667 6.033333 0.7234178 0.58594653
18 virginica 2.4 3 0 6.266667 5.433333 0.4509250 0.28867513
19 virginica 2.5 3 0 6.733333 5.933333 0.4509250 0.20816660
20 virginica 1.5 2 0 6.150000 5.050000 0.2121320 0.07071068
21 setosa 0.5 1 0 5.100000 1.700000 NA NA
22 setosa 0.6 1 0 5.000000 1.600000 NA NA
23 versicolor 1.7 1 0 6.700000 5.000000 NA NA
24 versicolor 1.8 1 0 5.900000 4.800000 NA NA
25 virginica 1.4 1 0 6.100000 5.600000 NA NA
26 virginica 1.6 1 0 7.200000 5.800000 NA NA
27 virginica 1.7 1 0 4.900000 4.500000 NA NA
28 setosa NA 50 1 5.006000 1.462000 0.3524897 0.17366400
29 versicolor NA 50 1 5.936000 4.260000 0.5161711 0.46991098
30 virginica NA 50 1 6.588000 5.552000 0.6358796 0.55189470
31 <NA> 150 3 5.843333 3.758000 0.8280661 1.76529823
```

```

```
R> ore.summary(iris, c("Sepal.Length", "Petal.Length"), c("mean", "std"), class=c("Species", "Petal.Width"), ways=1, order=c("type", "-freq", "class"))
   Species Petal.Width FREQ TYPE MEAN(Sepal.Length) MEAN(Petal.Length) STD(Sepal.Length) STD(Petal.Length)
1 setosa      NA    50    1      5.006000      1.462000  0.3524897  0.1736640
2 versicolor  NA    50    1      5.936000      4.260000  0.5161711  0.4699110
3 virginica   NA    50    1      6.588000      5.552000  0.6358796  0.5518947
4 <NA>        0.2    29    2      4.972414      1.444828  0.3544579  0.1723497
5 <NA>        1.3    13    2      5.884615      4.176923  0.3804518  0.2586949
6 <NA>        1.5    12    2      6.183333      4.658333  0.4323999  0.2644319
7 <NA>        1.8    12    2      6.400000      5.333333  0.4631905  0.5228129
8 <NA>        1.4    8     2      6.325000      4.637500  0.5750776  0.4808846
9 <NA>        2.3    8     2      6.912500      5.700000  0.5436320  0.5976143
10 <NA>       0.3    7     2      4.971429      1.428571  0.3988077  0.1380131
11 <NA>       0.4    7     2      5.300000      1.571429  0.2449490  0.1889822
12 <NA>       1.0    7     2      5.414286      3.628571  0.4450789  0.3199702
13 <NA>       2.0    6     2      6.650000      5.550000  0.9710819  0.7867655
14 <NA>       2.1    6     2      6.916667      5.783333  0.4070217  0.4355074
15 <NA>       0.1    5     2      4.820000      1.380000  0.3271085  0.1643168
16 <NA>       1.2    5     2      5.780000      4.240000  0.2167948  0.3209361
17 <NA>       1.9    5     2      6.340000      5.320000  0.6542171  0.4494441
18 <NA>       1.6    4     2      6.375000      5.025000  0.5678908  0.5737305
19 <NA>       1.1    3     2      5.400000      3.566667  0.2645751  0.4932883
20 <NA>       2.2    3     2      6.866667      6.033333  0.7234178  0.5859465
21 <NA>       2.4    3     2      6.266667      5.433333  0.4509250  0.2886751
22 <NA>       2.5    3     2      6.733333      5.933333  0.4509250  0.2081666
23 <NA>       1.7    2     2      5.800000      4.750000  1.2727922  0.3535534
24 <NA>       0.5    1     2      5.100000      1.700000      NA          NA
25 <NA>       0.6    1     2      5.000000      1.600000      NA          NA
26 <NA>      NA    150    3      5.843333      3.758000  0.8280661  1.7652982

R> ore.summary(iris, c("Sepal.Length", "Petal.Length"), c("mean", "prt"), class="Species", mu=c(5.8, 3.7))
   Species FREQ TYPE MEAN(Sepal.Length) MEAN(Petal.Length) PRT(Sepal.Length) PRT(Petal.Length)
1 setosa    50    0      5.006000      1.462  5.204621e-21  2.403948e-56
2 versicolor 50    0      5.936000      4.260  6.845138e-02  4.284727e-11
3 virginica  50    0      6.588000      5.552  1.334983e-11  1.605835e-28
4 <NA>      150   1      5.843333      3.758  5.225603e-01  6.879680e-01

R> ore.summary(iris, c("Sepal.Length", "Petal.Length"), "mean", class="Species", maxid=c(Sepal.Length="Sepal.Width", Petal.Length="Petal.Width"))
   Species FREQ TYPE MEAN(Sepal.Length) MEAN(Petal.Length) MAXID(Sepal.Length->Sepal.Width) MAXID(Petal.Length->Petal.Width)
1 setosa    50    0      5.006000      1.462           4.0            0.2
2 versicolor 50    0      5.936000      4.260           3.2            1.6
3 virginica  50    0      6.588000      5.552           3.8            2.3
4 <NA>      150   1      5.843333      3.758           3.8            2.3
```

ore.summary - Examples

Compute the mean, standard deviation, and count for arrival delay for each destination airport

```
res <- ore.summary(data=ONTIME_S,  
                     var='ARRDELAY',  
                     class='DEST',  
                     stats=c('mean', 'std', 'n'))  
  
head(res)
```

```
R> res <- ore.summary(data=ONTIME_S,  
+                         var='ARRDELAY',  
+                         class='DEST',  
+                         stats=c('mean', 'std', 'n'))  
  
R> res  
   DEST    FREQ TYPE      MEAN      STD     N  
1  ABE    237   0  5.68141593 26.678086 226  
2  ABI     34   0 21.38235294 61.406255  34  
3  ABQ   1357   0  6.99331352 24.111867 1346  
4  ABY     10   0 18.80000000 44.263855  10  
5  ACK      3   0 11.66666667 16.072751   3  
6  ACT     33   0  1.63636364 16.607262  33  
7  ACV     41   0  8.07317073 17.757238  41  
8  ACY     18   0  3.27777778 16.640862  18  
9  ADK      2   0 -0.50000000 23.334524   2  
10 ADQ    20   0 12.78947368 30.817706  19  
11 AEX     24   0 13.70833333 41.577041  24  
12 AGS     81   0 14.33750000 33.809947  80
```

- Compute the maximum arrival and departure delay for each airline and the corresponding destination airport

```
ore.summary(ONTIME_S, class="UNIQUECARRIER",  
            var = list("ARRDELAY", "DEPDELAY"),  
            stats=c(MAXARRDELAY="max", MAXDEPDELAY="max"),  
            maxid=c(ARRDELAY="DEST"),  
            minid=c(DEPDELAY="DEST"))
```

Results next slide...

ore.summary

Results

	UNIQUECARRIER	FREQ	TYPE	MAXARRDELAY	MAXDEPDELAY	MAXID	MINID
1	9E	747	0	247	249	MSP	BDL
2	AA	27655	0	1216	1200	DFW	BOS
3	AQ	234	0	123	114	HNL	HNL
4	AS	5174	0	333	364	SEA	ANC
5	B6	1187	0	322	347	LAS	JFK
6	CO	15299	0	929	947	IAH	MOT
7	DH	1045	0	348	339	IAD	CVG
8	DL	30305	0	530	1438	LAX	RIC
9	EA	2326	0	378	403	ATL	LGA
10	EV	2338	0	683	687	ATL	AMA
11	F9	481	0	191	197	DEN	DEN
12	FL	1760	0	407	415	ATL	HPN
13	HA	353	0	162	158	ITO	HNL
14	HP	6820	0	462	476	IND	SUX
15	ML(1)	134	0	167	60	MDW	BOS
16	MQ	5943	0	1025	1024	LAX	BOS
17	NW	18656	0	1423	1432	DTW	MSO
18	OH	2100	0	415	350	CVG	CVG
19	OO	4294	0	338	316	PDX	LAX
20	PA(1)	734	0	341	390	LAX	JFK
21	PI	2320	0	340	333	CLT	SAV
22	PS	416	0	190	188	YKM	SFO
23	TW	7477	0	412	420	PDX	PUB
24	TZ	300	0	910	132	MDW	MDW
25	UA	24307	0	484	493	ORD	MIA
26	US	25745	0	427	433	GSO	CLT
27	WN	27323	0	455	455	HOU	MSY
28	XE	3300	0	395	329	IAH	IAH
29	YV	1159	0	385	366	IAD	ORD
30	<NA>	219932	1	1423	1438	DTW	RIC

ore.rank

Enables investigation of distribution of values in numeric columns of an ore.frame

Highlights

- Ranking within groups
- Partition rows into groups based on rank tiles
- Cumulative percentages and percentiles
- Treatment of ties
- Calculation of normal scores from ranks

Simple syntax abstracting complex SQL queries

ore.rank – Parameters

Returns an ore.frame as output in all cases

data : ore.frame of the data to compute rankings on

var : numeric columns in **data** to rank

desc : ranks in descending (asc is default) order if TRUE

groups : partition rows into #groups based on ranks. For percentiles, #groups=100, For deciles #groups=10, For quartiles #groups=4.

group.by : rank each group identified by group.by columns separately

ties : specification of tie treatment. Assign largest of/smallest of/mean of corresponding ranks of tied values

fraction : rank of a column value \div # non missing column values

nplus1 : rank of a column value \div # non missing column values + 1

- Fraction and nplus1 options can be used to estimate cumulative distribution function

percent : (rank of a column value \div # non missing column values) * 100

Scoring Methods

- To compute Exponential scores from ranks use **savage**
- To compute normal scores – Use one of **blom**, **tukey** or **vw** (Van Der Waerden)

ore.rank – Examples

```
# Rank 2 columns and report them as derived columns
x <- ore.rank(data=NARROW, var='AGE=RankOfAge, CLASS=RankOfClass')
class(x)

y <- ore.sort(data=x, by='RankOfAge')

# Handling of ties
x <- ore.rank(data=NARROW, var='AGE=RankOfAge, CLASS=RankOfClass', ties='low')
head(x,10)

# Rank within groups
x <- ore.rank(data=NARROW, var='AGE=RankOfAge, CLASS=RankOfClass', group.by='COUNTRY')
head(x,10)
```

ore.rank – Examples

```
# Partition rows into groups e.g. Deciles
```

```
x <- ore.rank(data=NARROW, var='AGE=RankOfAge', CLASS=RankOfClass',groups=10)
head(x)
```

```
# Partition rows into groups e.g. Quartiles
```

```
x <- ore.rank(data=NARROW, var='AGE=RankOfAge', CLASS=RankOfClass',groups=4)
head(x)
```

```
# Estimating cumulative distribution function
```

```
x <- ore.rank(data=NARROW, var='AGE=RankOfAge', CLASS=RankOfClass',nplus1=TRUE)
head(x)
```

```
# Scores calculation
```

```
x <- ore.rank(data=NARROW, var='AGE=RankOfAge',
               CLASS=RankOfClass',score='savage', groups=100, group.by='COUNTRY')
head(x)
```

```
x <- ore.rank(data=NARROW, var='AGE=RankOfAge', CLASS=RankOfClass',score='blom')
```

```
head(x)
```

ore.sort

Enables flexible sorting of columns in a data frame

Can be used with other data pre-processing functions

- Sorting happens in the database
- (Top k) results of sorting can be provided as input to R visualization

Supports database nls.sort option

ore.sort - Parameters

Returns an ore.frame as output in all cases

data : ore.frame of the data to be sorted

by : columns in **data** to sort

nls.sort: A character string specifying Oracle Database NLS_SORT options

reverse: Allows optional reversal of collation order for character variables (TRUE/FALSE)

stable : Allows relative order to be maintained within sorted groups (TRUE/FALSE)

unique.keys : Allows optional removal of rows with duplicate values in the column(s) being sorted from appearing in the result (TRUE/FALSE)

unique.data: Allows optional removal of duplicate rows from appearing in the result (TRUE/FALSE)

ore.sort – Examples

```
# Sort all specified columns in desc order
x <- ore.sort(data=NARROW,by='AGE,GENDER' , reverse=TRUE)
head(x)

# Sort AGE in desc order but GENDER in ascending order
x <- ore.sort(data=NARROW,by=' -AGE,GENDER' )
head(x)

# Keep just 1 row per unique value of AGE
x <- ore.sort(data=NARROW,by='AGE' , unique.key=TRUE)
head(x)

# Remove duplicate rows
x <- ore.sort(data=NARROW,by='AGE' , unique.data=TRUE)
head(x)

# Remove duplicate rows as well as rows with duplicate values for AGE
x <- ore.sort(data=NARROW,by='AGE' , unique.data=TRUE , unique.key = TRUE)
head(x)

# Maintain relative order within sorted output
x <- ore.sort(data=NARROW,by='AGE' , stable=TRUE)
head(x)
```

ore.sort – Examples

Sort the ONTIME_S data by airline descending and departure delay ascending

```
sortedOntime <- ore.sort(data=ONTIME_S, by='-UNIQUECARRIER,DEPDELAY')
head(sortedOntime[,c(11,18)], 20)
```

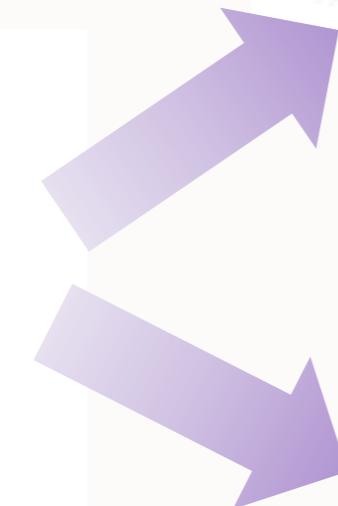
Sort ONTIME_S by airline and departure delay, but select one of each combination, e.g., unique key

```
sortedOntime2 <-
ore.sort(ONTIME_S,by='UNIQUECARRIER,DEPDELAY',unique.key=TRUE)
head(sortedOntime2[,c(11,18)], 20)
```

ore.crosstab - Basics

Input data set

	gender	age
Jill	female	12
Tom	male	11
Bob	male	12
Sally	female	13
Leigha	female	11
Phil	male	11



One way tables

gender frequency	male	female
	3	3

age frequency	11	12	13
	3	2	1

2 way table

	11	12	13
female	1	1	1
male	2	0	1

A general KxL table

	Column 1	Column 2	...	Column j	...	Column L	Total
Row 1	x_{11}	x_{12}	...	x_{1j}	...	x_{1L}	$x_{1\cdot}$
Row 2	x_{21}	x_{22}	...	x_{2j}	...	x_{2L}	$x_{2\cdot}$
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
Row i	x_{i1}	x_{i2}	...	x_{ij}	...	x_{iL}	$x_{i\cdot}$
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
Row K	x_{K1}	x_{K2}	...	x_{Kj}	...	x_{KL}	$x_{K\cdot}$
Total	$x_{\cdot 1}$	$x_{\cdot 2}$...	$x_{\cdot j}$...	$x_{\cdot L}$	n

ore.crosstab

Enables cross column frequency analysis of an ore.frame

A sophisticated variant of R's function **table ()** for two variables

Builds tables of frequency counts across columns of a data frame

Required as a pre-cursor to frequency analysis using ore.freq

R translated to 100% SQL

ore.crosstab – Parameters

Returns an ore.frame as output in all cases, except when multiple tables are created – in this case an ore.list is returned

expr: cross tabulation definition in the form

```
[COLUMN_SPEC] ~ COLUMN_SPEC [ * <WEIGHTING COLUMN>]  
[ / <GROUPING COLUMN>]  
[ ^ <STRATIFICATION COLUMN>]  
[ | ORDER_SPECIFICATION]
```

COLUMN_SPEC is <column-name>[+COLUMN_SET][+COLUMN_RANGE]

COLUMN_SET is <column_name>[+COLUMN_SET]

COLUMN_RANGE is <FROM COLUMN>-<TO COLUMN>

ORDER_SPEC is one of [-]NAME, [-]DATA, [-]FREQ, or INTERNAL

data: ore.frame of data to cross tabulate

group.by: as many cross tabulations as unique values in grouping columns

order: optional sorting of table data

[-] NAME: Sort by tabulation column names, [-]FREQ: Sort by frequency counts in the table

weights : Column of data that indicates frequency of occurrence of the corresponding row

where: An optional character vector specifying arbitrary partitions of argument data

strata : Column name used to cluster, or group, the data in combination

ore.crosstab – Examples

```
# For comparison, look at R's table function on 2 columns
table(NARROW$MARITAL_STATUS, NARROW$GENDER)

# Corresponding ore.crosstab(), extensible to more than 2 columns
ore.crosstab(MARITAL_STATUS ~ GENDER, data=NARROW)

# MARITAL_STATUS x GENDER and MARITAL_STATUS x CLASS
x=ore.crosstab(MARITAL_STATUS ~ GENDER+CLASS, data=NARROW)

# One way table
ore.crosstab(~AGE, data=NARROW)

# Weight values in AGE and GENDER using values in CLASS
x=ore.crosstab(AGE~GENDER*CLASS, data=NARROW)

# Order rows of cross tab by frequency counts
x=ore.crosstab(AGE~GENDER|FREQ, data=NARROW)
```

ore.crosstab – Results

```
> table(NARROW$MARITAL_STATUS, NARROW$GENDER)
```

	F	M
Divorc.	85	56
Mabsent	6	7
Mar-AF	1	0
Married	55	471
NeverM	147	201
Separ.	23	18
Widowed	27	4

```
> ore.crosstab(MARITAL_STATUS ~ GENDER, data=NARROW)
```

	MARITAL_STATUS	GENDER	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	Divorc.	F	85	1	1
2	Divorc.	M	56	1	1
3	Mabsent	F	6	1	1
4	Mabsent	M	7	1	1
5	Mar-AF	F	1	1	1
6	Married	F	55	1	1
7	Married	M	471	1	1
8	NeverM	F	147	1	1
9	NeverM	M	201	1	1
10	Separ.	F	23	1	1
11	Separ.	M	18	1	1
12	Widowed	F	27	1	1
13	Widowed	M	4	1	1

```
> x=ore.crosstab(MARITAL_STATUS ~ GENDER+CLASS, data=NARROW)
```

```
>
```

```
> x
```

```
$`MARITAL_STATUS~GENDER`
```

	MARITAL_STATUS	GENDER	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	Divorc.	F	85	1	1
2	Divorc.	M	56	1	1
3	Mabsent	F	6	1	1
4	Mabsent	M	7	1	1
5	Mar-AF	F	1	1	1
6	Married	F	55	1	1
7	Married	M	471	1	1
8	NeverM	F	147	1	1
9	NeverM	M	201	1	1
10	Separ.	F	23	1	1
11	Separ.	M	18	1	1
12	Widowed	F	27	1	1
13	Widowed	M	4	1	1

```
$`MARITAL_STATUS~CLASS`
```

	MARITAL_STATUS	CLASS	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	Divorc.	0	150	1	1
2	Divorc.	1	13	1	1
3	Mabsent	0	17	1	1
4	Mar-AF	0	1	1	1
5	Married	0	336	1	1
6	Married	1	284	1	1
7	NeverM	0	395	1	1
8	NeverM	1	24	1	1
9	Separ.	0	42	1	1
10	Separ.	1	4	1	1
11	Widowed	0	33	1	1
12	Widowed	1	1	1	1

ore.crosstab – Examples

```
# 4, 2 way cross tabs (GENDER,AGE,MARITAL_STATUS,COUNTRY) ~CLASS  
x=ore.crosstab(GENDER~COUNTRY~CLASS, data=NARROW)  
length(x)
```

```
# 1 way table with as many rows as unique values of COUNTRY for each unique value  
of AGE  
x = ore.crosstab(~AGE/COUNTRY, data=NARROW)
```

```
# Same as above, but 2 way  
x = ore.crosstab(AGE~GENDER/COUNTRY, data=NARROW)
```

```
# Post-process output - 2 2-way tables AGExGENDER and AGExCLASS  
x=ore.crosstab(AGE ~ GENDER+CLASS, data=NARROW)  
class(x)  
class(x[[1]])  
names(x[[1]])  
z <- x[[1]][,c(1,2,3)]
```

ore.crosstab – Results

```
> x=ore.crosstab(GENDER~COUNTRY~CLASS, data=NARROW) > x = ore.crosstab(~AGE/COUNTRY, data=NARROW) > x = ore.crosstab(AGE~GENDER/COUNTRY, data=NARROW)
> length(x)
[1] 4
>
> head(x)
$`GENDER~CLASS`
  GENDER CLASS ORE$FREQ ORE$STRATA ORE$GROUP
1   F     0      378       1       1
2   F     1       43       1       1
3   M     0      598       1       1
4   M     1      282       1       1
$`AGE~CLASS`
  AGE CLASS ORE$FREQ ORE$STRATA ORE$GROUP
1  17    0      14       1       1
2  18    0      16       1       1
3  19    0      30       1       1
4  20    0      23       1       1
5  21    0      22       1       1
6  22    0      39       1       1
7  23    0      29       1       1
8  24    0      24       1       1
9  25    0      34       1       1
10 25    1      2       1       1
11 26    0      17       1       1
```

↓

	AGE	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	17	1	1	1
2	17	1	1	3
3	17	12	1	19
4	18	16	1	19
5	19	30	1	19
6	20	23	1	19
7	21	1	1	1
8	21	1	1	9
9	21	20	1	19
10	22	2	1	1
11	22	1	1	8
12	22	36	1	19
13	23	1	1	3
14	23	28	1	19
15	24	3	1	1
16	24	2	1	9
17	24	1	1	15
18	24	1	1	18
19	23		F	9
20	23		M	18
21	23		M	1
22	24		F	1
23	24		F	1
24	24		F	1
25	24		F	6
26	24		F	1

↓

	AGE	GENDER	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	17	F	4	1	19
2	17	F	1	1	3
3	17	M	8	1	19
4	17	M	1	1	1
5	18	F	6	1	19
6	18	M	7	1	19
7	19	F	15	1	19
8	19	M	13	1	19
9	20	F	9	1	19
10	20	M	13	1	19
11	21	F	10	1	19
12	21	F	1	1	1
13	21	F	1	1	9
14	21	M	6	1	19
15	22	F	1	1	8
16	22	F	15	1	19
17	22	M	19	1	19
18	22	M	1	1	1
19	23	F	9	1	19
20	23	M	18	1	19
21	23	M	1	1	3
22	24	F	1	1	9
23	24	F	1	1	1
24	24	F	1	1	18
25	24	F	6	1	19
26	24	F	1	1	15

ore.crosstab – Results

```
> x=ore.crosstab(AGE ~ GENDER+CLASS, data=NARROW)
> class(x)
[1] "list"
> class(x[[1]])
[1] "ore.frame"
attr(,"package")
[1] "OREbase"
> names(x[[1]])
[1] "AGE"          "GENDER"        "ORE$FREQ"      "ORE$STRATA"    "ORE$GROUP"
> z <- x[[1]][,c(1,2,3)]
>
> z
   AGE GENDER ORE$FREQ
1  17     F      5
2  17     M      9
3  18     F      6
4  18     M      7
5  19     F     15
6  19     M     13
7  20     F      9
8  20     M     13
9  21     F     12
10 21     M      6
```

> x					
\$`AGE~GENDER`					
	AGE	GENDER	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	17	F	5	1	1
2	17	M	9	1	1
3	18	F	6	1	1
4	18	M	7	1	1
5	19	F	15	1	1
6	19	M	13	1	1
7	20	F	9	1	1
8	20	M	13	1	1
9	21	F	12	1	1
10	21	M	6	1	1
11	22	F	16	1	1
12	22	M	20	1	1
13	23	F	9	1	1
14	23	M	19	1	1
15	24	F	10	1	1
16	24	M	13	1	1
\$`AGE~CLASS`					
	AGE	CLASS	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	17	0	14	1	1
2	18	0	16	1	1
3	19	0	30	1	1
4	20	0	23	1	1
5	21	0	22	1	1
6	22	0	39	1	1
7	23	0	29	1	1
8	24	0	24	1	1
9	25	0	34	1	1
10	25	1	2	1	1
11	26	0	17	1	1
12	26	1	5	1	1
13	27	0	42	1	1

ore.crosstab – Examples

```
# Stratification - As many 2 way tables as unique values of CLASS
x <- ore.crosstab(AGE~GENDER^CLASS, data=NARROW)

# Custom binning and subsequent cross tabulation
NARROW$AGEBINS=ifelse(NARROW$AGE<20, 1,
                      ifelse(NARROW$AGE<30, 2,
                            ifelse(NARROW$AGE<40, 3, 4)))
ore.crosstab(GENDER~AGEBINS, NARROW)
```

ore.crosstab – Results

```
> x <- ore.crosstab(AGE~GENDER^CLASS, data=NARROW)
>
> x
```

	AGE	GENDER	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	17	F	5	1	1
2	17	M	9	1	1
3	18	F	6	1	1
4	18	M	7	1	1
5	19	F	15	1	1
6	19	M	13	1	1
114	77	M	1	1	1
115	80	M	2	1	1
116	82	M	1	1	1
117	90	F	1	1	1
118	90	M	2	1	1
119	25	F	1	2	1
120	25	M	1	2	1
121	26	F	2	2	1
122	26	M	2	2	1
123	27	M	3	2	1

```
> NARROW$AGEBINS=ifelse(NARROW$AGE<20, 1,
+                         ifelse(NARROW$AGE<30, 2,
+                               ifelse(NARROW$AGE<40, 3, 4)))
> ore.crosstab(GENDER~AGEBINS, NARROW)
```

	GENDER	AGEBINS	ORE\$FREQ	ORE\$STRATA	ORE\$GROUP
1	F	1	26	1	1
2	F	2	108	1	1
3	F	3	86	1	1
4	F	4	164	1	1
5	M	1	29	1	1
6	M	2	177	1	1
7	M	3	230	1	1
8	M	4	381	1	1

ore.freq

Operates on output of ore.crosstab() and automatically determines the techniques relevant to the nature of the result

1-way cross tables

- Goodness of fit tests for equal proportions or specified null proportions, confidence limits, and tests for equivalence

2-way cross tables

- Various statistics that describe relationships between columns in the cross tabulation
- Chi-square tests, cochrane-mantel-haenzsel statistics, measures of association, strength of association, risk differences, odds ratio and relative risk for 2x2 tables, tests for trend

N-way cross tables

- N 2-way cross tables
- Statistics across and within strata

Leverages database SQL functions when available

ore.freq - Parameters

Returns an OML4R data frame as output in all cases

x, crosstab: ore.frame output from ore.crosstab()

stats: List of statistics required

- ChiSquare: AJCHI, LRCHI, MHCHI, PCHISQ
- Lambda:LAMCR, LAMRC, LAMDAS
- Stuart's Tau, Somer's D|C:STUTC,SMDCR,SMDRC
- Odds Ratio:OR, MHOR, LGOR
- Others: MCNEM, PHI, CRAMV, CONTGY, TSYM, TREND, GAMMA,
- Kappa:KAPPA, WTKAP,
- Correlation:KENTB,PCORR, SCORR
- Fisher's, Cochran's Q:FISHER, COCHQ
- Relative Risk:RR,MHRR,ALRR

params: Control parameters to the specific statistical function

- SCORE: TABLE|RANK|RIDIT|MODRIDIT
- ALPHA: <number>
- WEIGHTS: <number>

skip.failed: (TRUE/FALSE) if a statistical test required fails on the cross table because it is found to be in-applicable to the table then return immediately

skip.missing: (TRUE/FALSE) skip cells with missing values in the cross table

ore.freq – Examples

Compute cross tabulation for number of diverted flights for each airline.
Compute the Pearson CHISQ for the results.

```
ct <- ore.crosstab(UNIQUECARRIER~DIVERTED, data=ONTIME_S)
ct
freq <- ore.freq(ct)
freq
```

For each airline, compute cross tabulation for number of diverted flights and day of week. Compute the Pearson CHISQ for each result.

```
ct <-
ore.crosstab(UNIQUECARRIER~DIVERTED+DAYOFWEEK, data=ONTIME_S)
ct
freq <- ore.freq(ct)
freq
```

```
R> ct <- ore.crosstab(UNIQUECARRIER~DIVERTED, data=ONTIME_S)
R> ct
  UNIQUECARRIER DIVERTED ORE$FREQ ORE$STRATA ORE$GROUP
0      9E          0     741        1        1
1      9E          1       6        1        1
2      AA          0   27564        1        1
3      AA          1     91        1        1
4      AQ          0    234        1        1
5      AS          0   5159        1        1
```

```
R> freq <- ore.freq(ct)
R> freq
  METHOD      FREQ DF      PVALUE      DESCRIPTOR GROUP
0 PCHISQ 54.58397 28 0.001907040 Pearson Chi-Square    1
```

```
R> ct <- ore.crosstab(UNIQUECARRIER~DIVERTED+DAYOFWEEK,data=ONTIME_S)
R> ct
$`UNIQUECARRIER~DIVERTED`
  UNIQUECARRIER DIVERTED ORE$FREQ ORE$STRATA ORE$GROUP
0      9E          0     741        1        1
1      9E          1       6        1        1
2      AA          0   27564        1        1
3      AA          1     91        1        1
```

```
$`UNIQUECARRIER~DAYOFWEEK`
  UNIQUECARRIER DAYOFWEEK ORE$FREQ ORE$STRATA ORE$GROUP
0      9E          1     117        1        1
1      9E          2     115        1        1
2      9E          3     113        1        1
3      9E          4     120        1        1
4      9E          5      99        1        1
```

```
R> freq <- ore.freq(ct)
R> freq
$`UNIQUECARRIER~DIVERTED`
  METHOD      FREQ DF      PVALUE      DESCRIPTOR GROUP
0 PCHISQ 54.58397 28 0.001907040 Pearson Chi-Square    1
```

```
$`UNIQUECARRIER~DAYOFWEEK`
  METHOD      FREQ DF      PVALUE      DESCRIPTOR GROUP
0 PCHISQ 260.6275 168 6.014678e-06 Pearson Chi-Square    1
```

ore.corr

Correlation analysis across numeric columns in an ore.frame
Supports partial correlations with a control column
Enables aggregations prior to correlations
Allows post-processing of results and integration into R code flow
Output can be made to conform to output of R's cor() function and so can
be post-processed by any CRAN function or graphics

ore.corr – Parameters

Returns an ore.frame as output in all cases except when group.by is used in which case an ore.list object is returned

data: ore.frame of the data to compute correlation coefficients

var: numeric column(s) of the **data** for which to build correlation matrix

stats: pearson (default), spearman, kendall

group.by: as many correlation matrices as unique values in group.by columns

freq: character string specifying a numeric column within argument **data** to use as a frequency count

with: character vector specifying the numeric columns in argument **data** to pair with columns specified in argument **var**

weight: column of **data** whose numeric values provide a multiplicative factor for **var** columns

partial: columns of **data** to use as control variables for partial correlation

Use OREeda:::ore.corr.as.matrix() to convert into R's cor() compatible output format

ore.corr – Examples

```
# R's cor -project out non-numeric columns first
names(NARROW)
N <- ore.pull(NARROW[,c(3,8,9)])
cor(N, use="complete.obs")
cor(N, method='spearman' , use="complete.obs")
cor(N, method='kendall' , use="complete.obs")

# Corresponding ore.corr
x1 <- ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS')
x2 <- ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS' , stats='spearman')
x3 <- ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS' , stats='kendall')
cor_compatible_matrix = OREeda:::ore.corr.as.matrix(x3)
class(cor_compatible_matrix)
```

ore.corr - Results

```
> names(NARROW)
[1] "ID"          "GENDER"       "AGE"          "MARITAL_STATUS" "COUNTRY"
[6] "EDUCATION"   "OCCUPATION"  "YRS_RESIDENCE" "CLASS"
> N <- ore.pull(NARROW[,c(3,8,9)])
> cor(N, use="complete.obs")
      AGE YRS_RESIDENCE CLASS
AGE    1.0000000  0.6568534 0.2206276
YRS_RESIDENCE 0.6568534  1.0000000 0.3362991
CLASS   0.2206276  0.3362991 1.0000000
> cor(N, method='spearman' , use="complete.obs")
      AGE YRS_RESIDENCE CLASS
AGE    1.0000000  0.7477093 0.2632748
YRS_RESIDENCE 0.7477093  1.0000000 0.3646760
CLASS   0.2632748  0.3646760 1.0000000
> cor(N, method='kendall' , use="complete.obs")
      AGE YRS_RESIDENCE CLASS
AGE    1.0000000  0.6332196 0.2173413
YRS_RESIDENCE 0.6332196  1.0000000 0.3200559
CLASS   0.2173413  0.3200559 1.0000000
> x1 <- ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS')
> x1
      ROW           COL PEARSON_T PEARSON_P PEARSON_DF
1     AGE YRS_RESIDENCE 0.6568534 0e+00    1098
2     AGE             CLASS 0.2200960 1e-15    1298
3     YRS_RESIDENCE    CLASS 0.3561869 0e+00    1298
> x2 <- ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS', stats='spearman')
> x2
      ROW           COL SPEARMAN_T SPEARMAN_P SPEARMAN_DF
1     AGE YRS_RESIDENCE 0.7462684 0e+00    1098
2     AGE             CLASS 0.2601221 1e-15    1298
3     YRS_RESIDENCE    CLASS 0.3835252 0e+00    1298
> x3 <- ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS', stats='kendall')
> x3
      ROW           COL KENDALL_T KENDALL_P KENDALL_DF
1     AGE YRS_RESIDENCE 0.6332196 0.000000e+00 <NA>
2     AGE             CLASS 0.2147107 4.285594e-31 <NA>
3     YRS_RESIDENCE    CLASS 0.3362078 1.094478e-73 <NA>
> cor_compatible_matrix = OREeda:::ore.corr.as.matrix(x3)
> class(cor_compatible_matrix)
[1] "matrix"
```

ore.corr – Examples

```
#Partial correlation
ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS', stats='spearman', partial='GENDER')

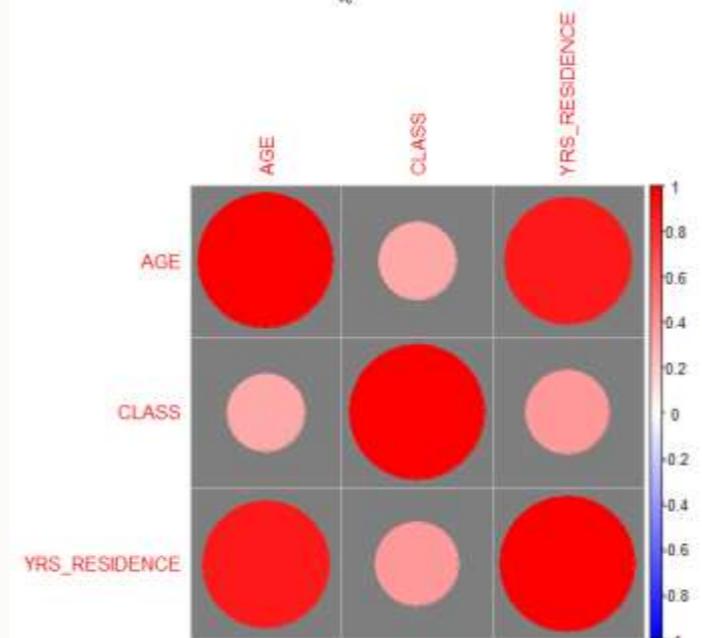
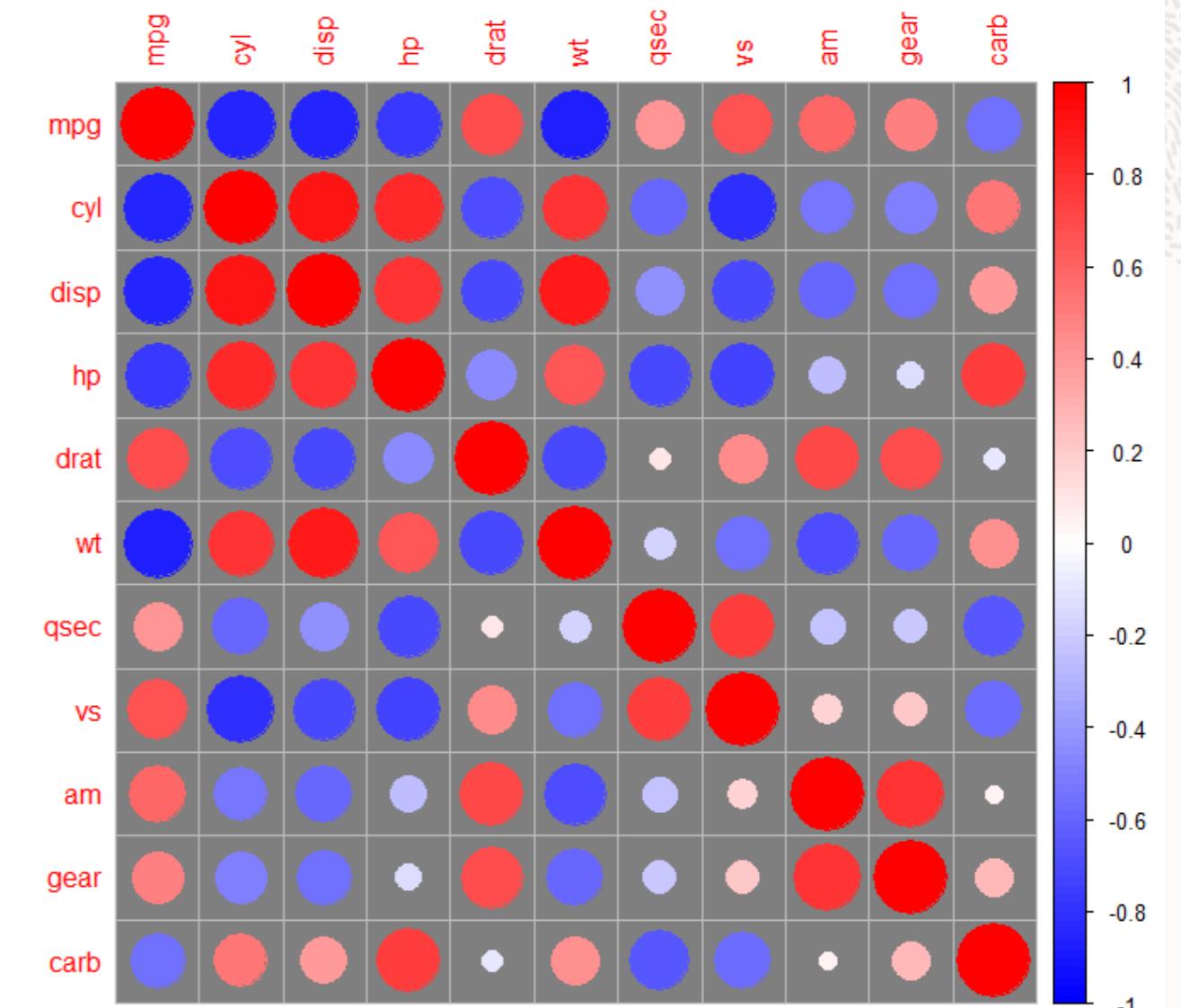
#Creating a number of correlation matrices
x <- ore.corr(NARROW,var='AGE,YRS_RESIDENCE,CLASS',
                stats='spearman', partial='GENDER', group.by='COUNTRY')
class(x)
cor_compatible_matrix <- OREeda:::ore.corr.as.matrix(x[[1]])
```

ore.corr

Post-processing matrix using CRAN visualization

```
library(corrplot)  
  
corrplot( cor(mtcars) , order = "original" , bg =  
"gray50" ,  
          col = colorRampPalette(c("blue","white","red"))(100) )  
  
corrplot(cor_compatible_matrix ,  
          order = "original" , bg = "gray50" ,  
          col = colorRampPalette(c("blue","white","red"))(100) )
```

<http://addictedtor.free.fr/graphiques/RGraphGallery.php?graph=152>



ore.univariate

Enables distribution analysis of numeric variables in an ore.frame

Statistics

- All statistics reported by ore.summary()
- Signed rank test, Student's t-test
- Extreme values reporting

Graphics

- QQ plots
- Scatterplots

ore.univariate – Parameters

Returns an ore.frame as output in all cases except when group.by is used in which case an R list object is returned

data: ore.frame of the data whose columns are to be analyzed

var: numeric column(s) of the **data** for which to compute statistics

weight: A column of the **data** whose numeric values provide a multiplicative factor for **var** columns

stats: optional specification of a subset of statistics to be printed

momemts – n,sumwgt,mean,sum,stddev,var,skew,kurt.,uss.css.cv.stderr

measures – mean,stddev,median,var,mode,range,iqr

quantiles – p100,p99,p95,p90,p75,p50,p25,p10,p5,p1,p0

location – studentt,studentp,signt,signp,srankt,srankp

normality

loccount – loc<,loc>,loc!

extremes

ore.univariate - Examples

```
# Default univariate statistics
ore.univariate(NARROW, var="AGE,YRS_RESIDENCE,CLASS")

# Compute location statistics on YRS_RESIDENCE
# (Student's t-test and two-tailed p-value for student's t-test)
ore.univariate(NARROW, var="YRS_RESIDENCE",stats="location")

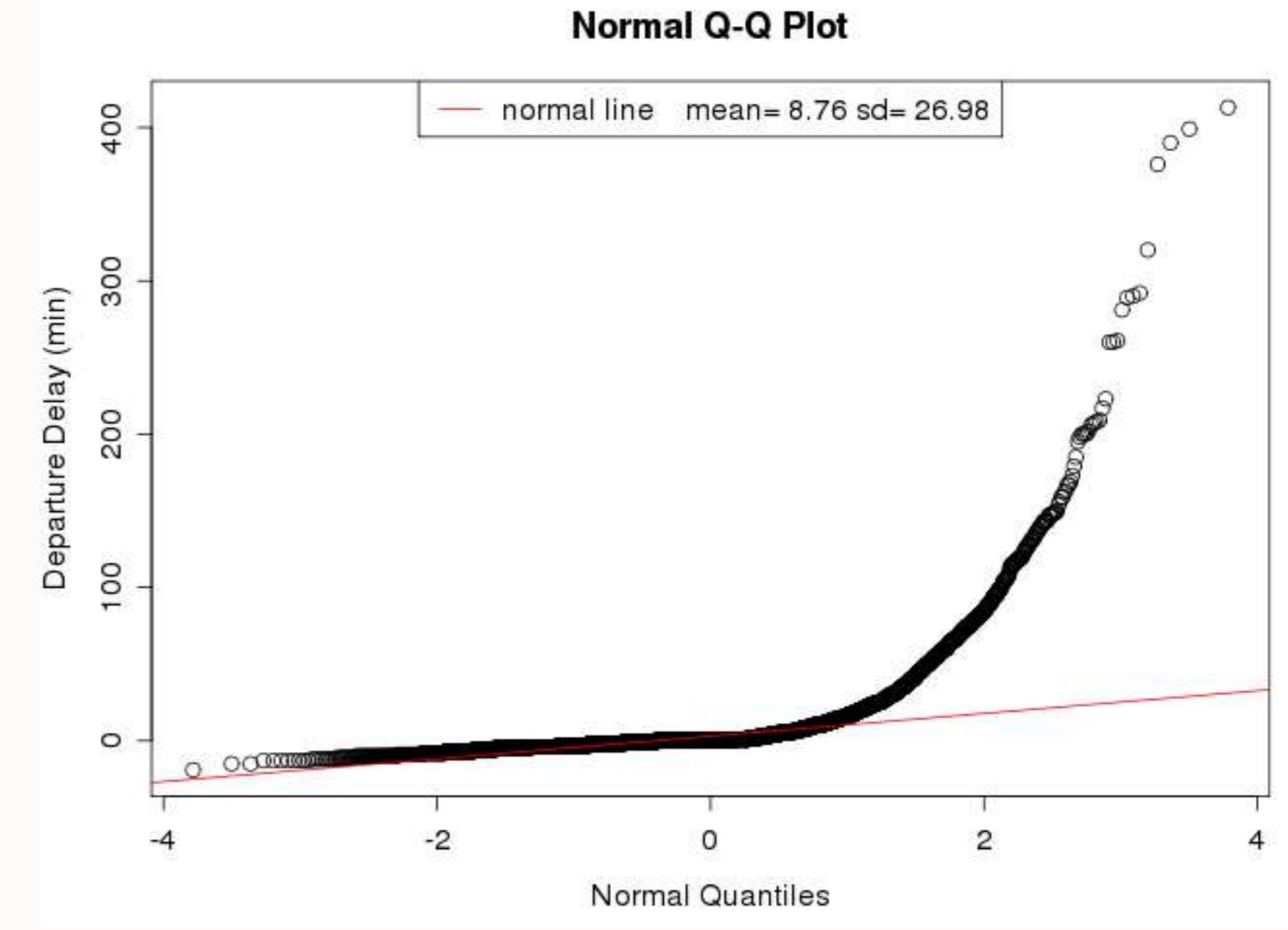
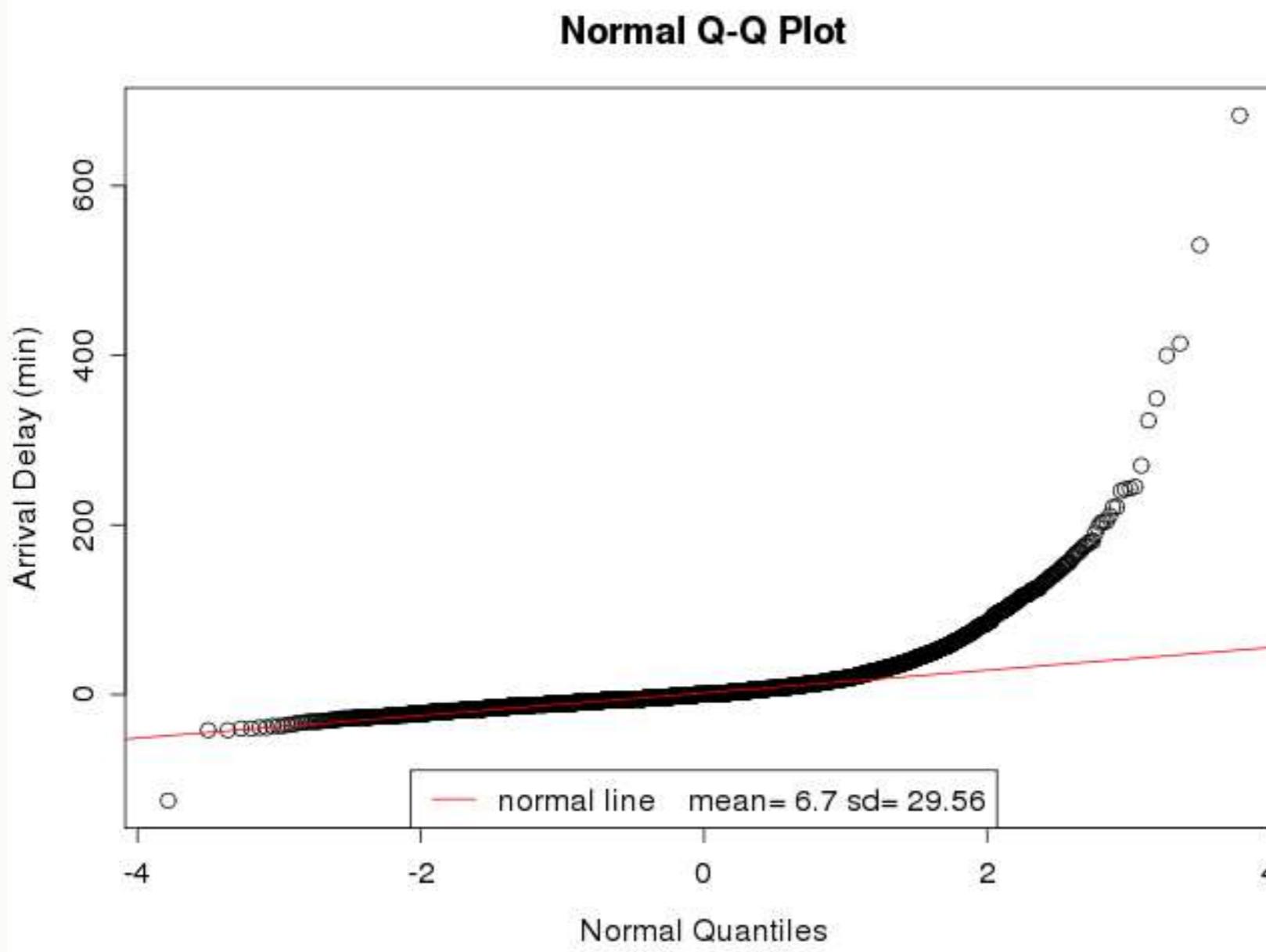
# Compute quantiles statistics on AGE and YRS_RESIDENCE
ore.univariate(NARROW, var="AGE,YRS_RESIDENCE",stats="quantiles")
```

ore.univariate – Results

```
R> # Default univariate statistics
R> ore.univariate(NARROW, var="AGE,YRS_RESIDENCE,CLASS")
   N(AGE) N(YRS_RESIDENCE) N(CLASS) SUMWGT(AGE) SUMWGT(YRS_RESIDENCE) SUMWGT(CLASS) MEAN(AGE) MEAN(YRS_RESIDENCE) MEAN(CLASS)
1 1300      1300     1500      1300      1300     1500 39.12769      4.081538  0.2533333
   SUM(AGE) SUM(YRS_RESIDENCE) SUM(CLASS) STDDEV(AGE) STDDEV(YRS_RESIDENCE) STDDEV(CLASS) VAR(AGE) VAR(YRS_RESIDENCE) VAR(CLASS)
1 50866      5306      380    13.69127      1.932932  0.4350652 187.451      3.736225  0.1892817
   USS(AGE) USS(YRS_RESIDENCE) USS(CLASS) CSS(AGE) CSS(YRS_RESIDENCE) CSS(CLASS) CV(AGE) CV(YRS_RESIDENCE) CV(CLASS) STDERR(AGE)
1 2233768     26510      380 243498.8      4853.357  283.7333 34.99126      47.35792 171.7363  0.3797276
   STDERR(YRS_RESIDENCE) STDERR(CLASS) SKEW(AGE) SKEW(YRS_RESIDENCE) SKEW(CLASS) KURT(AGE) KURT(YRS_RESIDENCE) KURT(CLASS)
1      0.05360988  0.01123334 0.6001366      0.821637  1.135444 0.01924574      1.739197 -0.7117187
R>
R> # Compute location statistics on YRS_RESIDENCE
R> ore.univariate(NARROW, var="YRS_RESIDENCE",stats="location")
   T PRT
1 76.13407  0
R>
R> # Compute quantiles statistics on AGE and YRS_RESIDENCE
R> ore.univariate(NARROW, var="AGE,YRS_RESIDENCE",stats="quantiles")
   P100(AGE) P100(YRS_RESIDENCE) P99(AGE) P99(YRS_RESIDENCE) P95(AGE) P95(YRS_RESIDENCE) P90(AGE) P90(YRS_RESIDENCE) P75(AGE)
1      90          14        75          10       64          7      58.1          7       48
   P75(YRS_RESIDENCE) P50(AGE) P50(YRS_RESIDENCE) P25(AGE) P25(YRS_RESIDENCE) P10(AGE) P10(YRS_RESIDENCE) P5(AGE) P5(YRS_RESIDENCE)
1        5         38          4        28          3       22          2       20          1
   P1(AGE) P1(YRS_RESIDENCE) P0(AGE) P0(YRS_RESIDENCE)
1      17            0        17            0
```

Normal QQ Plot – Univariate Graphics

Arrival and Departure Delay



Normal QQ Plot – Univariate Graphics

Arrival and Departure Delay

```
ontime <- ONTIME_S
ontimeSubset <- ontime[ontime$UNIQUECARRIER == "AA", , drop = TRUE]
ontimeSubset <- ontime[, c("ARRDELAY", "DEPDELAY")]
ontimeSubset <- ore.pull(ontimeSubset)
ontimeSubset <- ontimeSubset[sample(nrow(ontimeSubset), nrow(ontimeSubset) * .03),]

dat = ontimeSubset$ARRDELAY
type = "Arrival"
#dat = ontimeSubset$DEPDELAY
#type = "Departure"

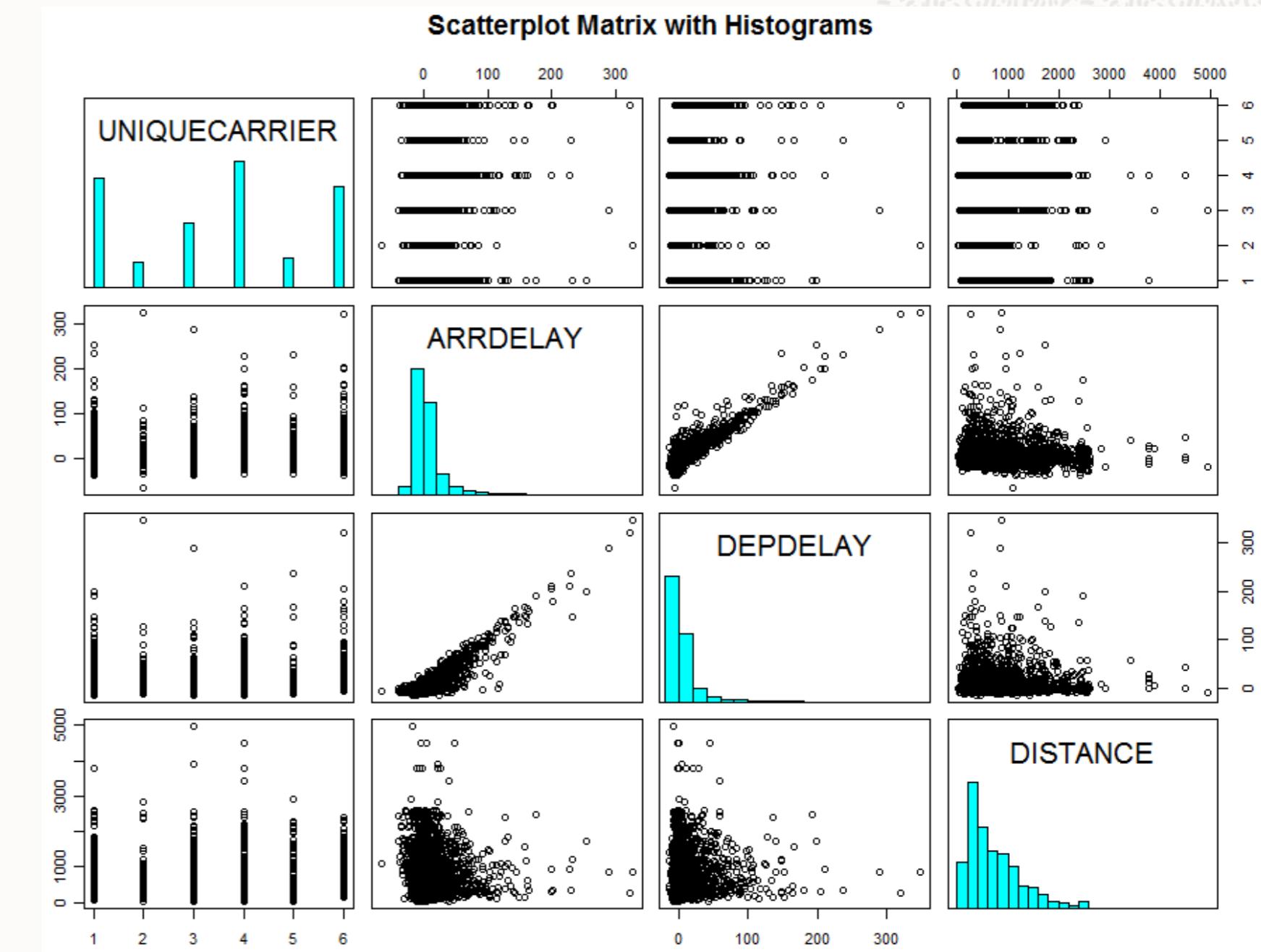
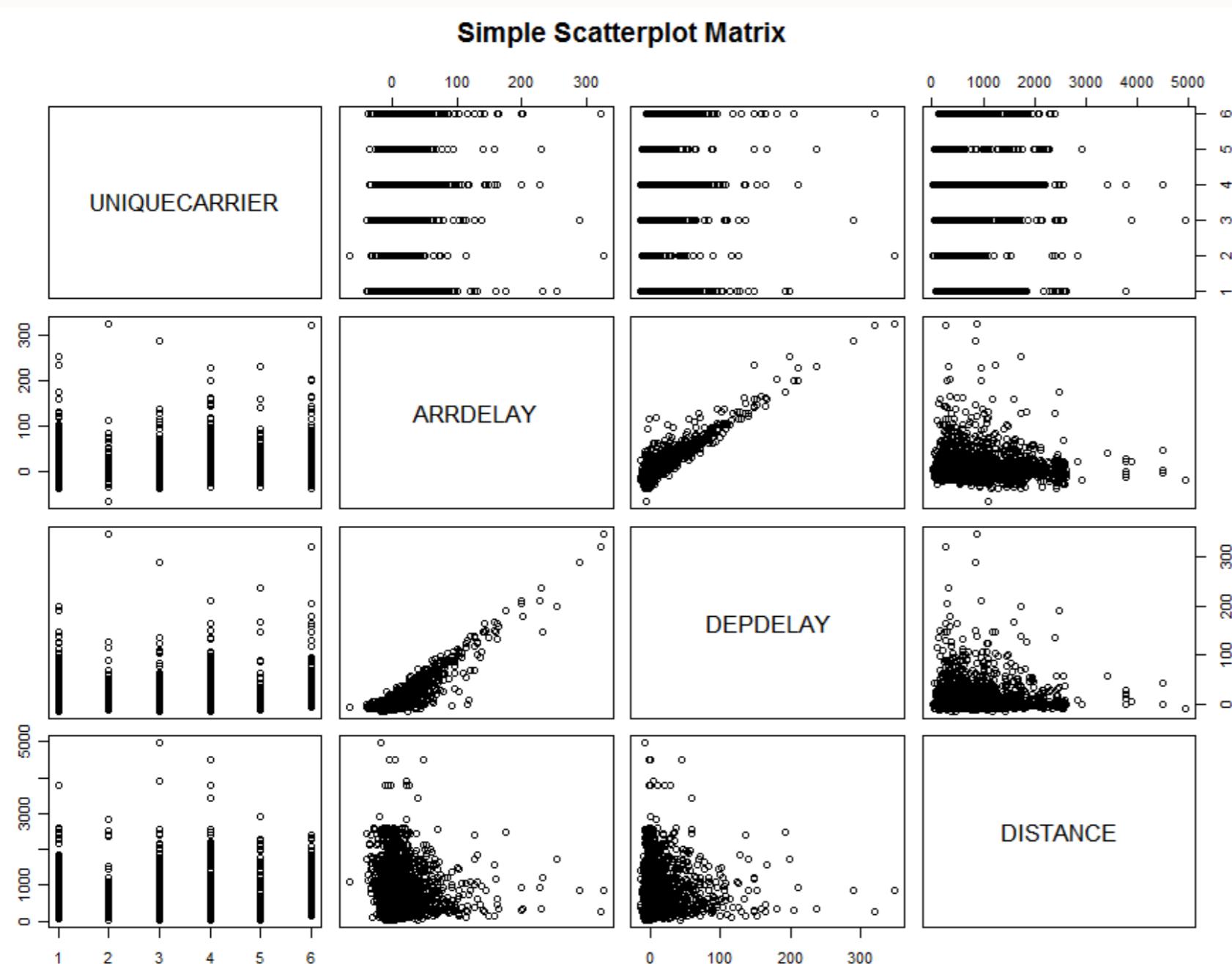
qqnorm(dat, xlab="Normal Quantiles", ylab=paste(type, " Delay (min)", sep=""));
qqline(dat, col = 2)

mean <- mean(dat, na.rm=TRUE)
sd <- sd(dat, na.rm=TRUE)

legend("top", paste("normal line      mean= ", round(mean, 2),
                   " sd= ", round(sd, 2), sep=""), lty=1, col="red")
```

Scatterplot Matrix

Airline, Arrival Delay, Departure Delay, Distance



Scatterplot Matrix

Airline, Arrival Delay, Departure Delay, Distance

```
ontime <- ONTIME_S
ontimeSubset <- ontime[ontime$UNIQUECARRIER %in% c("AA", "AS", "CO", "DL", "HP", "WN"), ,
                           drop = TRUE]
ontimeSubset <- ore.pull(ontimeSubset)
ontimeSubset <- ontimeSubset[sample(nrow(ontimeSubset), nrow(ontimeSubset) * .03), ]

pairs(~UNIQUECARRIER+ARRDELAY+DEPDELAY+DISTANCE, data=ontimeSubset,
      main="Simple Scatterplot Matrix")
```

```
os2 <- with(ontimeSubset, data.frame(UNIQUECARRIER, ARRDELAY, DEPDELAY, DISTANCE))

panel.hist <- function(x, ...) {
  usr <- par("usr"); on.exit(par(usr))
  par(usr = c(usr[1:2], 0, 1.5) )
  h <- hist(x, breaks=20, plot = FALSE)
  breaks <- h$breaks; nB <- length(breaks)
  y <- h$counts; y <- y/max(y)
  rect(breaks[-nB], 0, breaks[-1], y, col="cyan", ...)
}
pairs(os2, diag.panel=panel.hist, main="Scatterplot Matrix with Histograms")
```

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Oracle Machine Learning

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Thank You

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