DISCRETE CORRELATION OF HUMAN RACE MEASUREMENTS

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1. Summary

We introduce discrete correlations of categorical variables that are valid for the first time as a metric in statistics. No measure had existed that is valid for evaluating strength of relationships. We give complete R code for our discrete correlation metric.

Then we present the calculation of the discrete correlation matrix for some data from World Values Survey 7 using n=510 columns. We examine the eigenvalues of the 100×100 correlation matrix and assess the fit of the eigenvalue distribution by a Generalised Hyperbolic Distribution. We consider this distribution to be a universal quantity that allows us to formulate a coherent full scientific theory of Human Nature that is quantitative.

2. Code

```
# Zulf's Proportional Chi-Square
```

```
prop.chisq<-function( data ){</pre>
  m<-dim(data)[1]
  n<-dim(data)[2]
  t<-0
  v0<-data[1,]/sum(data[1,])</pre>
  for (j in 2:m){
    v<-data[j,]/sum(data[j,])</pre>
    t < -t + sum((v-v0)^2/v0^2)
  }
  df<-1
  t0<-qchisq(0.95, df=df)
  pval <- 1 - pchisq( t, df=df)</pre>
  list(tstat=t,pval=pval,crit=t0)
}
g<-function(x){
  #print(x)
  \log(-\log(x)+0.01)
```

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```
zulf.sigma<-function(z){</pre>
  n<-length(z)
  if (n==1){
    return(1)
  }
  #print(n)
  D<-matrix(0, nrow=n, ncol=n)</pre>
  diag(D) < -1
  #print(dim(D))
  s<-0
  for (j in 2:n){
    for (k in 1:j){
      gg = abs(z[j] - z[k])
      D[j,k] \leftarrow gg
      D[k,j] \leftarrow gg
      s<-s + gg
    }
  }
  out < 2*s/(n^2-n)
  out
}
zulf.chisq<-function( data ){</pre>
  data<-as.matrix(data)</pre>
  m<-dim(data)[1]</pre>
  n<-dim(data)[2]
  if ( m<=1 || n<=1){
    return(list(tstat=0,pval=1,t0=1))
  }
  t<-0
  eps<-0.00001
  w0 <- (data[1,]+eps)/sum(data[1,]+eps)</pre>
  #print('this')
  #print(length(w0))
  v0<- log(-log(abs(w0)+eps))
  #print(v0)
  sigma0 <- zulf.sigma( v0 )</pre>
  mu0 \leftarrow mean(v0)
  for (j in 2:m){
    w<-(data[j,]+eps)/sum(data[j,]+eps)</pre>
    #print(w)
    v < -g(w)
    #print('works')
    #print(v)
    #print('---')
    sigma <- zulf.sigma( v )</pre>
    mu <- mean(v)</pre>
```

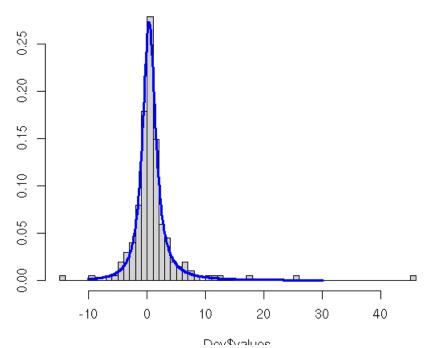
```
z <- v/sigma
    z0 <- v0/sigma0
    t < -t + sum((z - z0)^2)
  df < -(n-1)*(m-1)
  t0 < -qchisq(0.95, df = df)
  pval <- 1 - pchisq( t, df=df)</pre>
  list(tstat=t,pval=pval,crit=t0)
# calculate the discrete correlation
# matrix of WVS7
tablify<-function(a, b){</pre>
  H<-na.omit(data.frame(a,b))</pre>
  T<-matrix(1,nrow=1,ncol=1)
  tryCatch( { T<-table(H) } )</pre>
}
cor.d<-function( data ){</pre>
  n<-dim(data)[2]</pre>
  out<-matrix(0,nrow=n, ncol=n)</pre>
  diag(out)<-1
  if (n==2){
    out[1,1]<-1
    out[2,2] < -1
    a<-data[,1]
    b<-data[,2]
    X<-tablify( a, b )</pre>
    metrics <- zulf.chisq(X)</pre>
    out[1,2] \leftarrow 1 - metrics$pval
    \operatorname{out}[2,1] \leftarrow 1 - \operatorname{metrics}\operatorname{pval}
  }
  if (n>2){
    for (j in 2:n){
       for (k in 1:j){
         print(paste(j,k))
         a<-data[,j]
         b<-data[,k]
         X<-tablify( a, b )</pre>
         metrics <- zulf.chisq(X)</pre>
         print('ok')
         out[j,k] <- 1 - metrics$pval</pre>
         out[k,j] <- 1 - metrics$pval</pre>
      }
    }
```

```
}
out
}
```

3. Distribution of 100 correlation Eigenvalues

First let us take a look at the histogram of eigenvalues and the Generalised Hyperbolic Distribution fit.

Histogram of Dev\$values



The Generalised Hyperbolic Fit is quite gorgeous and accurate. Let's look at the fitted parameters.

Asymmetric Generalized Hyperbolic Distribution:

Parameters:

```
lambda alpha.bar mu sigma
-0.734819351 0.003937837 0.300662297 8.962769977
gamma
2.746209012
```

Call:

fit.ghypuv(data = Dev\$values)

${\tt Optimization} \ {\tt information:}$

log-Likelihood: -480.6028 AIC: 971.2055

Fitted parameters: lambda, alpha.bar, mu, sigma, gamma; (Number: 5)

Number of iterations: 294

Converged:

TRUE

4. Generalised Hyperbolic Distribution Parameters For Human Values Correlation are Scientific Parameters for Human Nature

We have obtained values with 100 Human Nature Variables. It is our view that as the number of variables tend to infinity, there will be a convergence to finite parameters of a GHD, and these parameters are *Human Nature Invariants*. This is one of the most important observations in the entire history of Social Science, and one of the greatest scientific discoveries in all of Human History. This discovery tells us that a smooth parametric distribution describes the empirical distribution of eigenvalues of correlation of every possible human nature measurement at all in the infinite variable limit.

5. What is New Here

I had found that using ordinary correlations, a result of this type exists. Unfortunately ordinary correlations are meaningless for categorical variables. Here we are presenting a result that is far more significant as science, for *discrete correlation* had to be invented and computed for this distribution, and these are valid scientifically for correlation eigenvalues.