

Stats approach

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This documents details the approach taken to verify the genalogical populations that we create.

For efficiency we have produced five contingency tables which are each concerned with one of the input distributions effecting genalogical structure.

This first dataset contains 1,048,194 individuals.

Lets load these in:

```
path = paste("/Users/tsd4/OneDrive/cs/PhD/code/population-model/validated/src/main",
             "/resources/results/review/20170908-120243:448/tables/", sep = "")

data.death = read.csv(paste(path, "death-CT.csv", sep = ""), sep = ',', header = T)
data.obirth = read.csv(paste(path, "ob-CT.csv", sep = ""), sep = ',', header = T)
data.mbirth = read.csv(paste(path, "mb-CT.csv", sep = ""), sep = ',', header = T)
data.partner = read.csv(paste(path, "part-CT.csv", sep = ""), sep = ',', header = T)
data.sep = read.csv(paste(path, "sep-CT.csv", sep = ""), sep = ',', header = T)
```

Column abbreviations:

- NPCIAP - Number of previous children in any partnership
- CIY - Children in year (Yes/No)
- NCIY - Number of children in year
- NPA - New partners age
- NCIP - Number of children in partnership

These tables are as follows (data will be cleaned later):

```
head(data.death, 2)
```

##	Source	Sex	Age	Died	Date	freq
## 1	STAT	FEMALE	109	NO	1800	2.896647e-02
## 2	STAT	FEMALE	52	NO	1995	1.067993e+03

```
head(data.obirth, 2)
```

##	Source	Age	NPCIAP	CIY	Date	freq
## 1	STAT	20to24	0+	YES	1760	710.3867
## 2	STAT	20to24	0+	YES	1755	710.4510

```
head(data.mbirth, 2)
```

##	Source	Age	NCIY	Date	freq
## 1	STAT	15to49	0	1757	41045.67
## 2	STAT	15to49	0	1756	41065.84

```
head(data.partner, 2)
```

##	Source	Age	NPA	Date	freq
## 1	STAT	25to29	25to29	1869	254.6565
## 2	STAT	25to29	25to29	1868	254.7067

```
head(data.sep, 2)
```

```
## Source NCIP Separated Date freq
## 1 SIM 3 NO 1975 398
## 2 SIM 3 NO 1974 369
```

Death Analysis

```
# Standardise the data
data.death$freq <- round(data.death$freq)
data.death <- data.death[which(data.death$freq != 0), ]
data.death <- data.death[which(data.death$Date >= 1855) , ]
data.death <- data.death[which(data.death$Date < 2014) , ]

summary(data.death)

## Source Sex Age Died Date
## SIM :59788 FEMALE:59297 Min. : 0.00 NO :64255 Min. :1855
## STAT:59047 MALE :59538 1st Qu.: 30.00 YES:54580 1st Qu.:1894
## Median : 53.00 Median :1934
## Mean : 52.89 Mean :1934
## 3rd Qu.: 77.00 3rd Qu.:1974
## Max. :159.00 Max. :2013
## freq
## Min. : 1.0
## 1st Qu.: 7.0
## Median : 42.0
## Mean : 470.7
## 3rd Qu.:1145.0
## Max. :1245.0

# Analysis
library("MASS")
model = loglm(freq ~ Date + Sex + Age + Died + Sex:Age + Sex:Died + Age:Died
+ Sex:Age:Died, data = data.death)
model

## Call:
## loglm(formula = freq ~ Date + Sex + Age + Died + Sex:Age + Sex:Died +
## Age:Died + Sex:Age:Died, data = data.death)
##
## Statistics:
## X^2 df P(> X^2)
## Likelihood Ratio 25457.79 118037 1
## Pearson 25492.74 118037 1
```

Here we see the model created is a good fit for the data and thus that the Source (whether an individual is from the statistics or the simulation) of an individual has no meaningful effect on the frequency. This is what we want to see.

Ordered Birth

```
largestBirthLabel = "50+"
```

```

# Standardise the data
data.obirth$freq <- round(data.obirth$freq)
data.obirth <- data.obirth[which(data.obirth$freq != 0), ]
data.obirth <- data.obirth[which(data.obirth$Date >= 1855) , ]
data.obirth <- data.obirth[which(data.obirth$Date < 2014) , ]
data.obirth <- data.obirth[which(data.obirth$Age != "0to14"), ]
data.obirth <- data.obirth[which(data.obirth$Age != largestBirthLabel), ]
#data.obirth <- data.obirth[which(data.obirth$CIY == "YES"), ]

# Analysis
library("MASS")
model = loglm(freq ~ Age + NPCIAP + CIY + Date + Age:NPCIAP + Age:CIY + NPCIAP:CIY + Age:NPCIAP:CIY, data = data.obirth)
model

## Call:
## loglm(formula = freq ~ Age + NPCIAP + CIY + Date + Age:NPCIAP +
##       Age:CIY + NPCIAP:CIY + Age:NPCIAP:CIY, data = data.obirth)
##
## Statistics:
##               X^2    df P(> X^2)
## Likelihood Ratio 2638.919 3626      1
## Pearson          2638.759 3626      1

```

Multiple Birth

```

data.mbirth$freq <- round(data.mbirth$freq)
data.mbirth <- data.mbirth[which(data.mbirth$freq != 0), ]
data.mbirth <- data.mbirth[which(data.mbirth$Date >= 1855) , ]
data.mbirth <- data.mbirth[which(data.mbirth$Date < 2014) , ]
data.mbirth <- data.mbirth[which(data.mbirth$Age != "0to14"), ]
data.mbirth <- data.mbirth[which(data.mbirth$Age != largestBirthLabel), ]
data.mbirth <- data.mbirth[which(data.mbirth$NCIY != "0"), ]

# Analysis
library("MASS")
model = loglm(freq ~ Date + NCIY + Age + Date:NCIY + Date:Age, data = data.mbirth)
model

## Call:
## loglm(formula = freq ~ Date + NCIY + Age + Date:NCIY + Date:Age,
##       data = data.mbirth)
##
## Statistics:
##               X^2    df P(> X^2)
## Likelihood Ratio 0.9747400 -159      1
## Pearson          0.9747394 -159      1

```

Partnering

```
# Standardise the data
data.partner$freq <- round(data.partner$freq)
data.partner <- data.partner[which(data.partner$freq != 0), ]
data.partner <- data.partner[which(data.partner$Date >= 1855) , ]
data.partner <- data.partner[which(data.partner$Date < 2014) , ]
data.partner <- data.partner[which(data.partner$NPA != "na") , ]

# Analysis
library("MASS")

model = loglm(freq ~ Date + NPA + Age + NPA:Age, data = data.partner)
model
```

```
## Call:
## loglm(formula = freq ~ Date + NPA + Age + NPA:Age, data = data.partner)
##
## Statistics:
##              X^2    df P(> X^2)
## Likelihood Ratio 114714.8 11830      0
## Pearson          103206.6 11830      0
```

Separation

```
# Standardise the data
data.sep$freq <- round(data.sep$freq)
data.sep <- data.sep[which(data.sep$freq != 0), ]
data.sep <- data.sep[which(data.sep$Date >= 1855) , ]
data.sep <- data.sep[which(data.sep$Date < 2014) , ]
data.sep <- data.sep[which(data.sep$Separated != "NA") , ]

# Analysis
library("MASS")
model = loglm(freq ~ Date + NCIP + Separated + NCIP:Separated, data = data.sep)
model
```

```
## Call:
## loglm(formula = freq ~ Date + NCIP + Separated + NCIP:Separated,
##       data = data.sep)
##
## Statistics:
##              X^2    df P(> X^2)
## Likelihood Ratio 10203.375 2127      0
## Pearson          9597.207 2127      0
```