

# Kenyan Demographic and Health Survey 2003: Data Preparation and Visualization

## Preparing data

The dataset *childrenfinal.dta* is obtained from Kenyan Demographic and Health Survey 2003 and contains various variables sampled in 2003 on the Kenyan children of age between 0 and 5 years. We want to “play” with this data. We will work with the dataset, where we will remove unnecessary columns and visualize some relationships between variables.

```
library(tidyverse) # to be able to use data visual.tools
library(foreign) #reading data in diff.formats
library(raster) #to manipulate geographic data
library(viridis) #to make plots easier to read
library(ggmap) #ggmap provides geoms for ggplot2
```

```
childrenfinal <- read.dta("childrenfinal.dta") # read the data
head(childrenfinal,2) #quick look at the dataset with 2 rows
```

```
## hypage deathu5 v001 ruralfacto female tetanusmother breastfeeding wantedchild
## 1 6 0 1 1 0 1 6 no more
## 2 28 0 1 1 0 NA 24 then
## anetalvisits placedelivery caesarian birthweight m37a m37c m37f m37h m37l
## 1 5 govt. hospital no 3500 8 8 8 8 8
## 2 NA govt. hospital no 3600 NA NA NA NA NA
## m37m m37n m37o m37p m37q m37r m37u m37v m37w m37x aidsinfo vaccbcg
## 1 8 8 8 8 8 8 8 8 8 yes vacc. date on card
## 2 NA NA NA NA NA NA NA NA NA <NA> reported by mother
## vaccdpt1 vaccpolio1 vaccdpt2 vaccpolio2
## 1 vacc. date on card vacc. date on card vacc. date on card vacc. date on card
## 2 reported by mother reported by mother reported by mother reported by mother
## vaccdpt3 vaccpolio3 vaccmeasles vaccpolio0
## 1 vacc. date on card vacc. date on card no vacc. date on card
## 2 reported by mother no reported by mother reported by mother
## diarrhea1 diarrhea2 diarrhea3 childage childweight childheight zstunt zweight
## 1 no <NA> <NA> 6 8.3 73 2.23 0.72
## 2 no <NA> <NA> 28 14.4 94 1.67 0.81
## zwastr sdist s820 s821 s823 v824 married v505 v002 v003 v005 interviewdate
## 1 -1.32 mbeere no no no no married 0 20 2 1374352 1241
## 2 0.12 mbeere no no no no married 0 20 2 1374352 1241
## v012 v024 v103 v104 v105 v106 v107 water
## 1 36 eastern countryside 1 countryside secondary 1 piped into dwelling
## 2 36 eastern countryside 1 countryside secondary 1 piped into dwelling
## v115 toilet electricity radio television v122 bicycle motor car floor
## 1 NA flush toilet no yes no no yes no no cement
## 2 NA flush toilet no yes no no yes no no cement
```

```

## walls roof v130 ethnicity
## 1 NA corrugated iron (mabati) protestant/other christian embu
## 2 NA corrugated iron (mabati) protestant/other christian embu
## yearsofedu v134 v135 v136 numberchildrenbelow5 v141
## 1 8 countryside usual resident 6 3 countryside
## 2 8 countryside usual resident 6 3 countryside
## v149 relationtohead sexhh agehh awfactt awfactu awfactr
## 1 incomplete secondary wife 0 40 100 100 100
## 2 incomplete secondary wife 0 40 100 100 100
## awfacte twin birthdate deathdateexact deathdatemonths birthinterval
## 1 100 single birth 1235 <NA> NA 22
## 2 100 single birth 1213 <NA> NA 40
## deadson deaddaughter agefirstbirth numberlivingchild knowledgecontraception
## 1 0 0 21 7 knows modern method
## 2 0 0 21 7 knows modern method
## contraceptionuse v367 v420 v421 v437 v438 v445 v446
## 1 used modern method wanted no more NA NA 741 1649 2725 1653
## 2 used modern method wanted no more NA NA 741 1649 2725 1653
## v704 v716 v739
## 1 other professional & related workers sales workers husband/partner alone
## 2 other professional & related workers sales workers husband/partner alone
## v742 v743a v743b
## 1 less than half respondent alone respondent and husband/partner
## 2 less than half respondent alone respondent and husband/partner
## v743c v743d v743e v753
## 1 respondent alone respondent and husband/partner respondent alone yes
## 2 respondent alone respondent and husband/partner respondent alone yes
## v754cp v754dp
## 1 don't know reduce chance of aids: have 1 sex partnr with no oth partner
## 2 don't know reduce chance of aids: have 1 sex partnr with no oth partner
## v754jp v754wp v756 v774 wealth assetindex motherid deathu1 deathu3 death
## 1 no no yes yes richest 1.04446 1000200200 0 0 0
## 2 no no yes yes richest 1.04446 1000200200 0 0 0
## periodborn periodborn3 birthage birthorder childorder ff modhypage edumother
## 1 0 0 35 7 1 1 6 3
## 2 0 0 34 6 2 1 28 3
## yearsofedu2 primary secondary birthorder2 childorder2 ruraljure birthage2
## 1 64 1 0 49 1 1 1225
## 2 64 1 0 36 4 1 1156
## deadchildren dtwin dbreast BMI motherunderweight severeunderweight Rohrer
## 1 0 NA 1 27.25 0 0 16.53
## 2 0 NA 1 27.25 0 0 16.53
## ai_toiletqual waterquality1 waterquality2 contraknowledge numbvac vacindex
## 1 2 1 0 1 8 1
## 2 2 1 0 1 8 1
## circumcision health1 health2 health3 cluster adm2 identifier distance
## 1 0 12393 848 1.5 1 NITHI 34 330.2652
## 2 0 12393 848 1.5 1 NITHI 34 330.2652
## ddistance1 ddistance2 hivclust hivnumb hivline hiv03 hiv05 hiv
## 1 0 0 1 20 2 <NA> 0 NA
## 2 0 0 1 20 2 <NA> 0 NA

```

There are 4686 observations on 177 variables, most of the variable names are self-explanatory.

Now we remove all variables that start with “s”, “v” and “m”. First of all we look how much are variables in

the dataset, which have names starting with these characters:

```
# names() allows us to show the column names from our dataset
# substring() allows us to get first letters from each column name
s <- substring(names(childrenfinal),1,1) # show dataset columns with only 1st letters
s
```

```
## [1] "h" "d" "v" "r" "f" "t" "b" "w" "a" "p" "c" "b" "m" "m" "m" "m" "m" "m"
## [19] "m" "m" "m" "m" "m" "m" "m" "m" "m" "a" "v" "v" "v" "v" "v" "v" "v" "v"
## [37] "v" "d" "d" "d" "c" "c" "c" "z" "z" "z" "s" "s" "s" "s" "v" "m" "v" "v"
## [55] "v" "v" "i" "v" "v" "v" "v" "v" "v" "w" "v" "t" "e" "r" "t" "v" "b"
## [73] "m" "c" "f" "w" "r" "v" "e" "y" "v" "v" "v" "n" "v" "v" "r" "s" "a" "a"
## [91] "a" "a" "a" "t" "b" "d" "d" "b" "d" "d" "a" "n" "k" "c" "v" "v" "v" "v"
## [109] "v" "v" "v" "v" "v" "v" "v" "v" "v" "v" "v" "v" "v" "v" "v" "v" "v"
## [127] "v" "w" "a" "m" "d" "d" "d" "p" "p" "b" "b" "c" "f" "m" "e" "y" "p" "s"
## [145] "b" "c" "r" "b" "d" "d" "d" "B" "m" "s" "R" "a" "w" "w" "c" "n" "v" "c"
## [163] "h" "h" "h" "c" "a" "i" "d" "d" "d" "h" "h" "h" "h" "h" "h" "h"
```

```
# define a list with letters, where columns from dataset must be dropped
dropped <- c("s","v","m")
```

```
#we modify our df without columns
#which names start with s,v and m
childrenfinal <- childrenfinal[, !(s %in% dropped)]
# The function above from the right side allows us
# To select us columns without s,v and m in the beginnig of names
```

```
head(childrenfinal)#quick look at the modifed df
```

```
## hypage deathu5 ruralfacto female tetanusmother breastfeeding wantedchild
## 1 6 0 1 0 1 6 no more
## 2 28 0 1 0 NA 24 then
## 3 20 0 1 0 2 20 then
## 4 47 0 1 1 NA 24 then
## 5 14 0 1 1 3 14 then
## 6 15 0 0 0 2 15 then
## anetalvisits placedelivery caesarian birthweight aidsinfo
## 1 5 govt. hospital no 3500 yes
## 2 NA govt. hospital no 3600 <NA>
## 3 4 private hosp/clinic yes 2500 yes
## 4 NA respondents home no 2500 <NA>
## 5 4 govt. hospital no 2900 no
## 6 4 govt. health center no 2800 yes
## diarrhea1 diarrhea2 diarrhea3 childage childweight
## 1 no <NA> <NA> 6 8.3
## 2 no <NA> <NA> 28 14.4
## 3 no <NA> <NA> 20 12.4
## 4 no <NA> <NA> 47 12.7
## 5 yes, last two weeks yes: no treatment no 14 8.1
## 6 no <NA> <NA> 15 9.5
## childheight zstunt zweight zwast interviewdate water
## 1 73.0 2.23 0.72 -1.32 1241 piped into dwelling
## 2 94.0 1.67 0.81 0.12 1241 piped into dwelling
```

## 3	85.0	0.24	0.44	0.45	1241	piped into dwelling			
## 4	91.0	-2.44	-1.82	-0.40	1241	piped into compound/plot			
## 5	73.5	-0.84	-1.65	-1.38	1241	piped into compound/plot			
## 6	76.9	-0.68	-1.16	-0.91	1242	public tap			
##	toilet electricity radio television bicycle car floor walls								
## 1	flush toilet		no	yes	no	yes	no	cement	NA
## 2	flush toilet		no	yes	no	yes	no	cement	NA
## 3	traditional pit toilet		no	yes	yes	yes	yes	cement	NA
## 4	flush toilet		no	yes	yes	yes	no	cement	NA
## 5	flush toilet		no	yes	yes	yes	no	cement	NA
## 6	traditional pit toilet		no	yes	no	no	no	cement	NA
##	roof ethnicity yearsofedu numberchildrenbelow5								
## 1	corrugated iron (mabati)		embu		8				3
## 2	corrugated iron (mabati)		embu		8				3
## 3	corrugated iron (mabati)		meru		15				1
## 4	corrugated iron (mabati)		embu		8				2
## 5	corrugated iron (mabati)		embu		8				2
## 6	corrugated iron (mabati)	taita/tavate			8				1
##	relationtohead	agehh	awfactt	awfactu	awfactr	awfacte	twin birthdate		
## 1	wife	40	100	100	100	100	single birth		1235
## 2	wife	40	100	100	100	100	single birth		1213
## 3	wife	43	100	100	100	100	single birth		1221
## 4	wife	30	100	100	100	100	single birth		1194
## 5	wife	30	100	100	100	100	single birth		1227
## 6	wife	30	100	100	100	100	single birth		1227
##	deathdateexact	deathdatemonths	birthinterval		deadson	deaddaughter			
## 1	<NA>		NA		22	0	0		
## 2	<NA>		NA		40	0	0		
## 3	<NA>		NA		24	0	0		
## 4	<NA>		NA		NA	0	0		
## 5	<NA>		NA		33	0	0		
## 6	<NA>		NA		24	0	0		
##	agefirstbirth	numberlivingchild	knowledgecontraception		contraceptionuse				
## 1	21		7		knows modern method		used modern method		
## 2	21		7		knows modern method		used modern method		
## 3	24		1		knows modern method		used modern method		
## 4	18		2		knows modern method		used modern method		
## 5	18		2		knows modern method		used modern method		
## 6	18		1		knows modern method		used modern method		
##	wealth	assetindex	deathu1	deathu3	death	periodborn	periodborn3	birthage	
## 1	richest	1.04446	0	0	0	0	0	35	
## 2	richest	1.04446	0	0	0	0	0	34	
## 3	richest	1.05364	0	0	0	0	0	24	
## 4	richest	0.98059	0	0	0	0	1	18	
## 5	richest	0.98059	0	0	0	0	0	20	
## 6	richest	0.93924	0	0	0	0	0	18	
##	birthorder	childorder	ff	edumother	yearsofedu2	primary	birthorder2		
## 1	7	1	1	3	64	1	49		
## 2	6	2	1	3	64	1	36		
## 3	1	1	1	5	225	1	1		
## 4	1	2	1	2	64	1	1		
## 5	2	1	1	2	64	1	4		
## 6	1	1	1	2	64	1	1		
##	childorder2	ruraljure	birthage2	deadchildren	dtwin	dbreast	BMI Rohrer		

```
## 1      1      1      1225      0      NA      1 27.25 16.53
## 2      4      1      1156      0      NA      1 27.25 16.53
## 3      1      1      576      0      NA      1 23.00 14.35
## 4      4      1      324      0      NA      1 28.01 18.58
## 5      1      1      400      0      NA      1 28.01 18.58
## 6      1      0      324      0      NA      1 21.14 13.41
##      ai_toiletqual waterquality1 waterquality2 contraknowledge numbvac
## 1      2      1      0      1      8
## 2      2      1      0      1      8
## 3      1      1      0      1      0
## 4      2      1      2      1      9
## 5      2      1      2      1      9
## 6      1      0      0      1      9
##      circumcision health1 health2 health3 cluster adm2 identifier distance
## 1      0 12393      848 1.500000      1 NITHI      34 330.2652
## 2      0 12393      848 1.500000      1 NITHI      34 330.2652
## 3      0 12393      848 1.500000      1 NITHI      34 330.2652
## 4      0 12393      848 1.500000      1 NITHI      34 330.2652
## 5      0 12393      848 1.500000      1 NITHI      34 330.2652
## 6      0 14022      819 2.433333      2 KILIFI      14 655.0334
##      ddistance1 ddistance2 hivclust hivnumb hivline hiv03 hiv05 hiv
## 1      0      0      1      20      2 <NA>      0 NA
## 2      0      0      1      20      2 <NA>      0 NA
## 3      0      0      1      30      2 <NA>      0 NA
## 4      0      0      NA      NA      NA <NA>      NA NA
## 5      0      0      NA      NA      NA <NA>      NA NA
## 6      0      0      NA      NA      NA <NA>      NA NA
```

```
substring(names(childrenfinal),1, 1)#to be sure that we dropped necessary columns
```

```
## [1] "h" "d" "r" "f" "t" "b" "w" "a" "p" "c" "b" "a" "d" "d" "d" "c" "c" "c" "z"
## [20] "z" "z" "i" "w" "t" "e" "r" "t" "b" "c" "f" "w" "r" "e" "y" "n" "r" "a" "a"
## [39] "a" "a" "a" "t" "b" "d" "d" "b" "d" "d" "a" "n" "k" "c" "w" "a" "d" "d" "d"
## [58] "p" "p" "b" "b" "c" "f" "e" "y" "p" "b" "c" "r" "b" "d" "d" "d" "B" "R" "a"
## [77] "w" "w" "c" "n" "c" "h" "h" "h" "c" "a" "i" "d" "d" "d" "h" "h" "h" "h" "h"
## [96] "h"
```

Now we have 96 columns.

Now we remove all but the variables *hypage*, *ruralfacto*, *breastfeeding*, *birthweight*, *yearsofedu*, *female*, *zstunt*, *zweight*, *zwast*, *adm2*.

```
#define a list with names, which want to have in our df
nec.vars <- c("hypage","ruralfacto","breastfeeding", "birthweight",
  "yearsofedu", "female", "zstunt", "zweight", "zwast", "adm2")

#we make same procedure as above
#but without ! in the right side
#it means that we leave only necessary columns in the df
childrenfinal <- childrenfinal[, (names(childrenfinal) %in% nec.vars)]
head(childrenfinal,2)
```

```
##      hypage ruralfacto female breastfeeding birthweight zstunt zweight zwast
```

```
## 1      6      1      0      6      3500      2.23      0.72 -1.32
## 2     28      1      0     24      3600      1.67      0.81  0.12
##   yearsofedu adm2
## 1          8 NITHI
## 2          8 NITHI
```

So, now our dataset has 10 columns, what will allow us to work further. Although, we have to be sure, that all remaining variables have reasonable variable type:

```
str(childrenfinal) #quick review of variable types in df
```

```
## 'data.frame':  4686 obs. of  10 variables:
## $ hypage      : num  6 28 20 47 14 15 50 14 10 18 ...
## $ ruralfacto  : num  1 1 1 1 1 0 0 0 0 0 ...
## $ female      : num  0 0 0 1 1 0 1 0 0 1 ...
## $ breastfeeding: int  6 24 20 24 14 15 17 14 10 18 ...
## $ birthweight : int  3500 3600 2500 2500 2900 2800 3000 4000 3000 3500 ...
## $ zstunt      : num  2.23 1.67 0.24 -2.44 -0.84 ...
## $ zweight     : num  0.72 0.81 0.44 -1.82 -1.65 ...
## $ zwast       : num  -1.32 0.12 0.45 -0.4 -1.38 ...
## $ yearsofedu  : int  8 8 15 8 8 8 0 0 8 7 ...
## $ adm2        : chr  "NITHI" "NITHI" "NITHI" "NITHI" ...
```

It seems that it would be better to format **female**, **ruralfacto** and **adm2** into factor type to easily categorize the data. We can do it with function “as.factor()”:

```
#one approach is to index with the $ sign and the as.factor function
```

```
#convert gender column into factor
```

```
childrenfinal$female <- as.factor(childrenfinal$female)
```

```
#convert territory column into factor
```

```
childrenfinal$ruralfacto <- as.factor(childrenfinal$ruralfacto)
```

```
#convert provinces column into factor
```

```
childrenfinal$adm2 <- as.factor(childrenfinal$adm2)
```

## Plots with Z-score

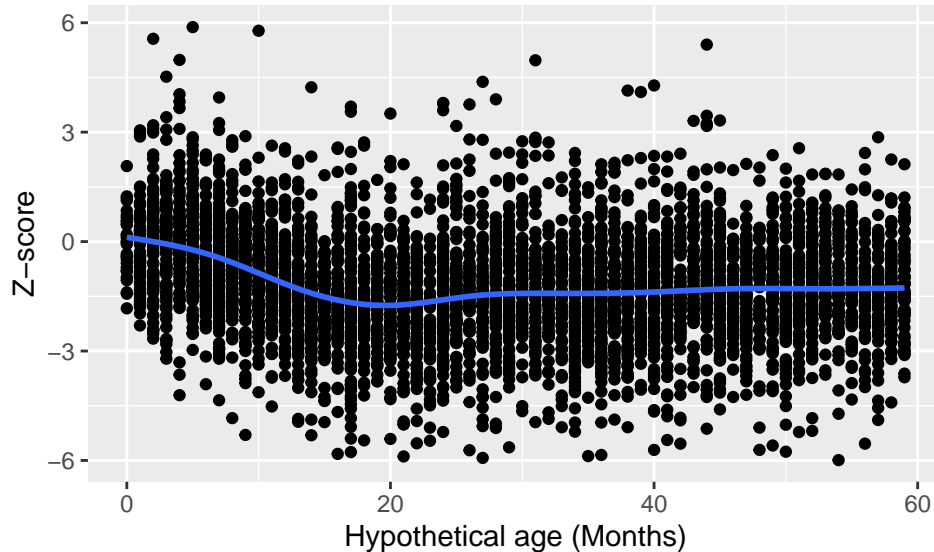
Variable **zstunt** is the so-called *Z-score* for stunting and is defined as the height of a child standardised with the median and standard deviation of heights of children at the same age from a healthy population. Children with Z-score less than  $-2$  are defined to be stunted. We will make a scatter plot of **zstunt** against **hypage** with a smooth line to the plot, without confidence bands.

Lets plot the data. First, we initiate a ggplot2-object:

```
plt <- ggplot(data = childrenfinal)
```

Now, we can plot the relationship between **zstunt** and **hypage** using this object:

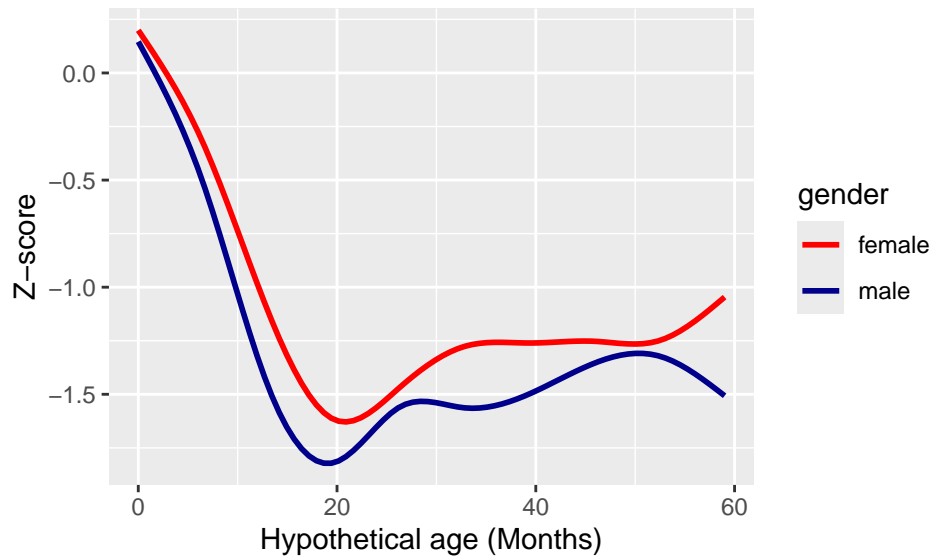
```
plt + geom_point(aes(x = hypage, y = zstunt))+
  geom_smooth(aes(x = hypage, y = zstunt ),se = F)+
  labs(x = "Hypothetical age (Months)",y = "Z-score")#add labels
```



There is a bit complicated to say something about linear dependence between variables, it seems that Z-score is negative in average. We can make an assumption that children in Kenya have quite serious problem with stunt.

Now we make smooth plots of **zstunt** against **hypage** for **females** and **males**. Here help us *filter()* function, which allows us elegantly to access specific information within data set, as in our case different plots for male/female. We will drop the scatter plot, that doesn't help to visualize in any way:

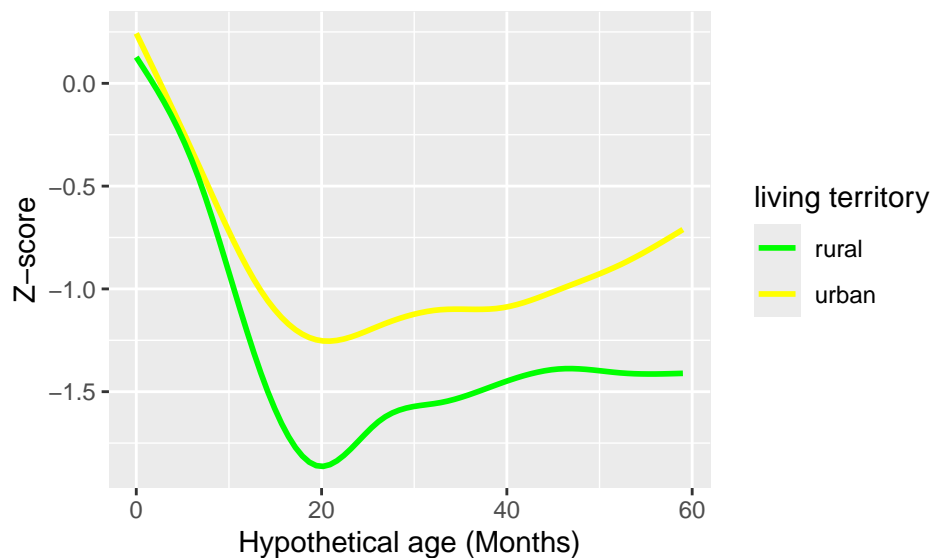
```
#Plot male and female smooth lines
#with filter(), where female == 0 is for male data
ggplot() +
  geom_smooth(data = filter(childrenfinal, female == 0),
    aes(x = hypage, y = zstunt, colour = "male" ),se = F) +
  geom_smooth(data = filter(childrenfinal, female == 1),
    aes(x = hypage, y = zstunt, colour = "female"), se = F)+
  labs(x = "Hypothetical age (Months)",y = "Z-score")+
  scale_colour_manual(name="gender", values=c("red","darkblue"))#set legend
```



We notice that female have smaller Z-score in average comparing to male, it means that in Kenya girls have less problems with stunt than boys.

Similarly, we plot **zstunt** against **hypage** for *urban* and *rural* children. We use the identical code from above:

```
#Plot urban and rural smooth lines
#with filter(), where ruralfacto == 1 is the data for rural children
ggplot() +
  geom_smooth(data = filter(childrenfinal, ruralfacto == 0),
    aes(x = hypage, y = zstunt, colour = "urban" ),se = F) +
  geom_smooth(data = filter(childrenfinal, ruralfacto == 1),
    aes(x = hypage, y = zstunt, colour = "rural"), se = F)+
  labs(x = "Hypothetical age (Months)",y = "Z-score")+
  scale_colour_manual(name="living territory", values=c("green","yellow"))
```





Here we see a big deviation of urban territory from rural. We can conclude that children, who live in urban area, have less problems with stunt comparing to children living in rural area.