project 2

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Project 2

```
# Load library
library(countreg)
## Loading required package: MASS
## Loading required package: Formula
library(ROCR)
library(ggplot2)
library(foreign)
library(statmod)
## Attaching package: 'statmod'
## The following object is masked from 'package:countreg':
##
##
       qresiduals
library(dplyr)
##
## Attaching package: 'dplyr'
## The following object is masked from 'package:MASS':
##
##
       select
## The following objects are masked from 'package:stats':
##
##
       filter, lag
## The following objects are masked from 'package:base':
##
       intersect, setdiff, setequal, union
##
```

library(caret)

Loading required package: lattice

```
library(xlsx)
suppressMessages(library(car))
```

```
# Load data
athletes <- read.xlsx("/home/wud18/Rstudio/STA712/project2/athletes_data.xlsx", sheetIndex = 1)
# Remove NA
athletes <- na.omit(athletes)
# Races under 10 seconds minus 1
athletes$Races.under.10.seconds <- athletes$Races.under.10.seconds-1
athletes</pre>
```

##		Date.of.birth	Name	РВ	ID	Height.cm.	X.Weight.kg.
##	1	1946-09-10	Jim Hines 9	.95	20720	183	81
##	2	1955-09-20	Silvio Leonard 9	.98	40973	173	64
##	3	1961-07-01	Carl Lewis 9	.86	6473	188	90
##	4	1961-01-08	Calvin Smith 9	.93	6367	178	69
##	5	1959-08-10	Mel Lattany 9	.96	30160	175	78
##	7	1960-04-02	Linford Christie 9	.87	26791	189	94
##	8	1965-03-18	Raymond Stewart 9	.96	36637	178	73
##	9	1967-02-21	Leroy Burrell 9	.85	26399	183	82
##	10	1966-02-20	Dennis Mitchell 9	.91	10489	174	69
##	11	1967-10-02	Frankie Fredericks 9	.86	13845	180	73
##	12	1969-01-20	Andre Cason 9	.92	2654	170	70
##	13	1969-08-19	Olapade Adeniken 9	.95	33434	186	78
##	14	1967-08-04	Michael Marsh 9	.93	30428	178	75
##	15	1971-11-22	Davidson Ezinwa 9		10167	184	82
##	16	1972-06-17	Daniel Effiong 9	.98	9052	187	80
##	17	1968-09-09	Jon Drummond 9		21961	175	75
##	18	1967-12-16	Donovan Bailey 9		11422	183	82
##	19	1967-07-12	Bruny Surin 9	.84	6164	180	81
##	20	1973-12-30	Ato Boldon 9	.86	4186	176	75
##	21	1974-07-23	Maurice Greene 9		29969	175	75
##	22	1973-03-30	Kareem Streete-Thompson 9	.96	23495	183	84
##	23	1975-01-28	Tim Montgomery 9		43767	178	73
##	24	1975-02-24	Percival Spencer 9		35013	182	68
##	25	1977-12-28	Seun Ogunkoya 9		40352	180	86
##	26	1972-10-20	Vincent Henderson 9		45952	175	66
##	27	1976-03-30	Obadele Thompson 9		33349	175	67
##	28	1973-05-09	Leonard Myles-Mills 9	.98	26307	172	66
##	29	1978-04-05	Dwain Chambers 9		11723	180	83
##	30	1975-09-17	Jason Gardener 9		19545	178	70
##	31	1974-01-27	Tim Harden 9		43738	178	79
##	32	1976-10-19	Coby Miller 9		8252	168	68
##	33	1978-01-19	Bernard Williams 9		4930	181	73
##	34	1978-11-22	Francis Obikwelu 9	.86	13713	195	80
##	35	1978-01-14	Shawn Crawford 9	.88	40509	181	80
##	36	1976-05-10	Joshua J. Johnson 9		20907	191	91
##	37	1974-12-05	Brian Lewis 9	.99	5946	170	72

##	38	1976-04-05	Kim Collins	9.93	24788	175	64
##	39	1972-09-26	Patrick Johnson		34303	177	73
##	40	1975-11-22	Deji Aliu	9.95	10325	187	75
##	41	1978-10-27	John Capel	9.95	21377	180	82
##	42	1982-02-10	Justin Gatlin	9.74	23308	185	79
##	43	1976-10-10	Mickey Grimes	9.99	30705	185	84
##	44	1976-09-17	Uchenna Emedolu	9.97	45192	183	79
##	45	1982-11-23	Asafa Powell	9.72	4109	190	88
##	46	1976-09-02	Aziz Zakari	9.99	4377	177	73
##	47	1983-01-07	Marc Burns	9.96	28099	185	88
##	48	1984-10-11	Darrel Brown	9.99	9427	184	79
##	49	1982-11-16	Ronald Pognon	9.99	38218	184	70
##	50	1980-01-19	Leonard Scott	9.91	26327	181	84
##	51	1984-07-09	Olusoji Fasuba	9.85	51282	170	78
##	52	1982-08-09	Tyson Gay	9.69	45173	180	73
##	53	1978-04-24	Marcus Brunson	9.99	28377	185	77
##	54	1984-01-05	Derrick Atkins	9.91	10613	185	88
##	55	1986-01-31	Walter Dix	9.88	52892	178	84
##	56	1987-03-27	Samuel Francis	9.99	61273	190	80
##	57	1984-12-24	Wallace Spearmon	9.96	54727	190	80
##	58	1986-08-21	Usain Bolt		45032	196	88
##	59	1986-12-13	Travis Padgett	9.89	56297	174	80
##	60	1985-06-07	Richard Thompson	9.82	61016	187	79
##	61	1982-12-22	Rodney Martin	9.95	37847	175	79
##	62	1984-04-10	Mark Jelks	9.99	50988	170	66
##	63	1977-12-04	Darvis Patton	9.89	9546	183	75
##	64	1985-05-02	Ivory Williams	9.94	18479	173	77
##	65	1985-10-11	Nesta Carter	9.78	59574	178	70
##	66	1984-07-03	Churandy Martina		40902	178	75
##	67	1982-10-06	Michael Frater	9.88	30332	170	67
##	68	1986-09-09	Daniel Bailey		55393	173	70
	69	1985-04-24	Mike Rodgers		58527	178	73
##		1989-12-26	Yohan Blake		69837	181	79
	71	1981-06-12	Lerone Clarke		26397	174	66
##		1990-06-11	Christophe Lemaitre		84220	189	74
##	73	1985-07-13	Trell Kimmons		58466	178	77
##		1989-04-13	Ryan Bailey		88407	193	98
##		1985-10-30	Mario Forsythe		59594	173	68
##		1982-11-29	Steve Mullings		41985	173	68
##		1987-03-11	Ngonidzashe Makusha		83912	178	73
##		1990-04-07	Nickel Ashmeade		79234	184	87
##		1988-03-08	Keston Bledman		66064	183	75
##		1990-04-05	Mookie Salaam			180	73
##		1984-01-01	Jaysuma Saidy Ndure		51645	192	72
##		1989-11-27	Harry Adams		72677	182	81
##		1989-10-11	Kemar Hyman		70904	178	74
##		1992-01-10	Kemar Bailey-Cole			195	86
##		1990-01-05	Isiah Young			183	75
##		1989-12-12	Dentarius Locke			170	68
##		1988-02-23	Gabriel Mvumvure		83908	172	75
##		1991-07-04	Charles Silmon		93272	175	72
##		1987-09-05	James Dasaolu		82420	180	75
##		1992-02-27	Jimmy Vicaut			188	83
##	91	1985-05-25	Simon Magakwe	9.98	90520	177	73

##	00	1002-07-20	Vomenler Press	0 02	100621	190	70
		1992-07-20	Kemarley Brown			180	72
##		1994-03-05	Chijindu Ujah			180	75
##		1995-07-10	Trayvon Bromell			175	71
##		1991-05-15	Femi Ogunode			183	79
##		1992-05-15	Clayton Vaughn			173	77
##	97	1994-11-10	Andre De Grasse			180	73
##	98	1993-11-13	Bryce Robinson			178	75
##	99	1993-12-15	Marvin Bracy	9.85	123692	175	74
##	100	1989-08-29	Su Bingtian	9.83	84128	172	64
##	101	1993-10-06	Adam Gemili	9.97	125508	178	73
##	102	1992-07-13	Diondre Batson	9.94	117751	188	75
##	103	1993-03-05	Beejay Lee	9.99	150807	168	72
##	104	1992-09-18	Quentin Butler	9.96	129036	175	70
##	105	1993-09-21	Akani Simbine	9.84	135967	174	67
##	106	1993-07-16	Henricho Bruintjies	9.97	139896	178	70
##	107	1991-12-15	Andrew Fisher	9.94	115854	168	64
##	108	1992-07-15	Wayde Van Niekerk	9.94	122672	183	73
##	109	1994-04-25	Omar McLeod	9.99	127853	180	73
##	110	1991-03-19	Ameer Webb	9.94	117296	175	75
##	111	1986-11-11	Ben Youssef Meïté	9.96	67421	179	70
##	112	1993-12-30	Senoj-Jay Givans		136582	178	73
##	113	1992-05-27	Aaron Brown			185	79
	114	1989-05-04	Jak Ali Harvey			182	73
	115	1986-01-21	Rondel Sorrillo		80875	178	62
	116	1996-03-06	Christian Coleman			175	73
	117	1988-10-11	Joel Fearon		95593	175	77
	118	1995-09-26	Thando Roto			173	68
	119	1993-10-15	Ronnie Baker			178	73
	120	1994-08-28	Odean Skeen			181	75
	121		Nethaneel Mitchell-Blake			186	75
	122	1994-09-11	Cameron Burrell			173	68
	123	1994-01-29				175	75
	123	1994-01-29	Christopher Belcher Julian Forte				73
	125	1990-05-29				186	73
			Ramil Guliyev			187	
	126	1995-12-15	Yoshihide Kiryū			175	69
	127	1995-09-23	Kendal Williams			180	73
	128	1996-08-05	Jaylen Bacon			183	75 70
	130	1995-07-13	Zharnel Hughes			190	79
	131	1997-07-18	Noah Lyles			180	70
	133	1993-08-17	Xie Zhenye			185	80
	134	1998-06-15	Filippo Tortu			182	68
	135	1988-06-15	Barakat Al Harthi			172	64
	136	1993-06-10	Tyquendo Tracey			179	75
	137	1996-02-29	Reece Prescod			184	68
	138	1991-11-12	Roberto Skyers			187	83
	139	1996-10-07	Divine Oduduru			175	70
	140	1999-03-06	Abdul Hakim Sani Brown			187	73
##	141	1996-07-31	Cravon Gillespie	9.93	162131	175	66
##	142	1997-03-18	Mario Burke	9.98	138213	175	64
##	143	1995-05-13	Yuki Koike	9.98	133238	178	72
##	145	1997-12-03	Michael Norman	9.86	163895	183	73
##	146	1998-01-14	Benjamin Azamati-Kwaku	9.90	330232	170	60
##	147	1994-07-09	Kyree King	9.97	137530	181	68
##	148	1999-07-15	Jo'Vaughn Martin			178	74

	149	1995-05-07	Fred Kerley				191		93
	150	1994-09-26	Marcell Jacobs				184		73
	151		lotliso Leotlela				180		70
	152	1992-06-10	Ryota Yamagata				176		70
	154	2001-11-12	Micah Williams				173		80
	155	2000-03-03	Enoch Adegoke				170		71
	156		Ferdinand Omurwa				175		83
	159		oseph Paul Amoah				180		68
	160	2003-06-07	Letsile Tebogo				184		72
	164	1994-08-22	Elijah Hall				174		68
	165	1991-04-15	Emmanuel Matadi				183		85
	168	1996-11-21	Reynier Mena				174		79
	169		léba-Mickaël Zeze				175		65 70
	173		Udodi Onwuzurike				178		70
	177		Shaun Maswanganyi				185		77
	180	1993-08-17	Emmanuel Eseme				184	NA.	83
## ##	1	Races.under.10.seconds		.race: 20		ice.oi.	17055	NA. 1	
##		0		66			20352	2	
##		14		160			22463	3	
##		3		138			22289	4	
	5	0		99			21772	5	
	7	8		194			22008	7	
##		1		144			23819	8	
##		8		11:			24524	9	
	10	11		259			24158	10	
	11	26		220			24747	11	
	12	4		93			25223	12	
	13	2		122			25434	13	
##	14	2		94			24688	14	
##	15	3	}	108	3		26259	15	
##	16	1		66	3		26467	16	
##	17	9)	18	5		25090	17	
##	18	14	:	149	9		24822	18	
##	19	6	;	189	9		24665	19	
##	20	27	•	150)		27028	20	
##	21	51		160)		27233	21	
##	22	0		92	2		26753	22	
##	23	5		160)		27422	23	
##	24	0		34			27449	24	
##		3		69			28487	25	
##		0		74			26592	26	
##		2		140			27849	27	
##		1		103			26793	28	
##		3		27			28585	29	
##		0		14:			27654	30	
##		0		99			27056	31	
##		2		134			28052	32	
##		4		196			28509	33	
##		7		308			28816	34	
##		8		153			28504	35 36	
##		0		8:			27890	36	
##		0		175			27368	37	
##	3 8	9	1	358	5		27855	38	

				22522	
##		0	230	26568	39
##		1	248	27720	40
##		1	100	28790	41
##		63	216	29992	42
##		0	143	28043	43
##		0	185	28020	44
##	45	96	257	30278	45
##	46	1	225	28005	46
##	47	2	219	30323	47
##	48	0	137	30966	48
##	49	0	207	30271	49
##	50	2	98	29239	50
##	51	1	131	30872	51
##	52	37	162	30172	52
##	53	0	104	28604	53
##	54	2	116	30686	54
##	55	9	143	31443	55
##	56	0	116	31863	56
##	57	0	65	31040	57
##	58	52	86	31645	58
##	59	2	118	31759	59
##	60	12	187	31205	60
##	61	0	72	30307	61
##	62	0	106	30782	62
##	63	7	138	28463	63
##	64	2	104	31169	64
##	65	28	228	31331	65
##	66	5	279	30866	66
##	67	6	292	30230	67
##	68	6	154	31664	68
##	69	45	403	31161	69
##	70	45	214	32868	70
##	71	1	168	29749	71
##	72	6	151	33035	72
##	73	0	151	31241	73
##	74	8	82	32611	74
##	75	1	120	31350	75
##	76	3	113	30284	76[7]
##	77	1	20	31847	77
##	78	10	109	32970	78
##	79	4	240	32210	79
##	80	1	76	32968	80
##	81	0	172	30682	81
##		0	55	32839	82
##	83	0	191	32792	83
##		10	130	33613	84
##		9	159	32878	85
##		1	118	32854	86
##		0	68	32196	87
##		0	112	33423	88
##		1	120	32025	89
##		21	169	33661	90
##		0	174	31192	91
##		0	41	33805	92
		O .	-11	55500	72

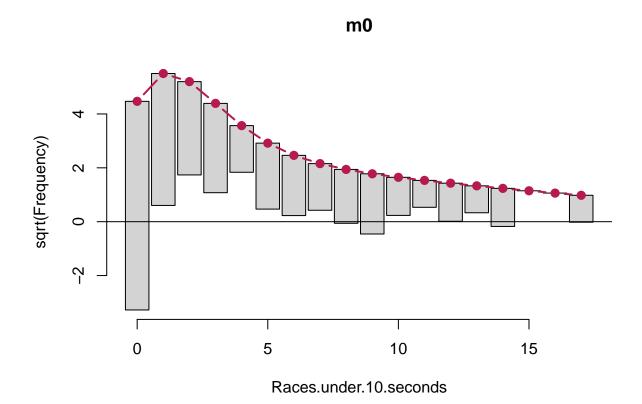
##		3	148	34398	93
##	94	29	108	34890	94[8]
##	95	6	89	33373	95
##	96	0	58	33739	96
##	97	18	107	34648	97[9]
##	98	0	67	34286	98
##	99	21	116	34318	99
##	100	9	151	32749	100
##	101	0	101	34248	101
##	102	0	68	33798	102
##	103	0	75	34033	103
##	104	0	59	33865	104
##	105	38	220	34233	105
##	106	0	147	34166	106
##	107	0	114	33587	107
##	108	1	21	33800	108
##	109	0	5	34449	109
##	110	0	82	33316	110
##	111	7	110	31727	111
##	112	0	65	34333	112
##	113	1	153	33751	113
##	114	1	186	32632	114
##	115	0	86	31433	115
##	116	30	84	35130	116
##	117	0	94	32427	117
##	118	0	112	34968	118
##	119	21	89	34257	119
##	120	0	47	34574	120
##	121	0	46	34426	121
##	122	1	92	34588	122
##	123	0	80	34363	123
##	124	3	104	33976	124
##	125	0	123	33022	125
##	126	0	153	35048	126
##	127	1	133	34965	127
##	128	0	66	35282	128
##	130	13	84	34893	130
##	131	17	77	35629	131
##	133	1	118	34198	133
##	134	0	63	35961	134
##	135	0	158	32309	135
##	136	1	145	34130	136
##	137	5	58	35124	137
##	138	0	92	33554	138
##	139	2	61	35345	139
##	140	3	90	36225	140
##	141	3	77	35277	141
##	142	0	122	35507	142
##	143	0	135	34832	143
##	145	2	9	35767	145
##	146	3	39	35809	146
##	147	3	108	34524	147
##	148	0	27	36356	148
##	149	25	40	34826	149

```
## 150
                            5
                                                  75
                                                                 34603
                                                                          150
## 151
                            0
                                                  77
                                                                 35927
                                                                          151
## 152
                                                                 33765
                            0
                                                 156
                                                                          152
                            5
                                                  47
## 154
                                                                 37207
                                                                          154
                                                                          155
## 155
                            0
                                                  94
                                                                 36588
## 156
                           12
                                                  92
                                                                 35066
                                                                         156
## 159
                            0
                                                  50
                                                                 35442
                                                                         159
                            5
## 160
                                                  29
                                                                 37779
                                                                          160
## 164
                            1
                                                  39
                                                                 34568
                                                                          164
## 165
                            2
                                                 118
                                                                 33343
                                                                          165
## 168
                            0
                                                 166
                                                                 35390
                                                                         168
                            0
## 169
                                                 109
                                                                 34473
                                                                          169
## 173
                            1
                                                  38
                                                                 37650
                                                                          173
## 177
                            1
                                                  67
                                                                 36923
                                                                          177
## 180
                                                  59
                                                                 34198
                                                                          180
# Load necessary library
library(dplyr)
# Calculate detailed summary statistics for all variables
detailed_stats <- athletes %>%
  summarise(
   Mean_PB = mean(PB, na.rm = TRUE),
   Median PB = median(PB, na.rm = TRUE),
   SD_PB = sd(PB, na.rm = TRUE),
   Min_PB = min(PB, na.rm = TRUE),
   Max_PB = max(PB, na.rm = TRUE),
   IQR_PB = IQR(PB, na.rm = TRUE),
   Mean Height = mean(Height.cm., na.rm = TRUE),
   Median_Height = median(Height.cm., na.rm = TRUE),
   SD_Height = sd(Height.cm., na.rm = TRUE),
   Min_Height = min(Height.cm., na.rm = TRUE),
   Max_Height = max(Height.cm., na.rm = TRUE),
    IQR Height = IQR(Height.cm., na.rm = TRUE),
   Mean Weight = mean(X.Weight.kg., na.rm = TRUE),
   Median_Weight = median(X.Weight.kg., na.rm = TRUE),
   SD_Weight = sd(X.Weight.kg., na.rm = TRUE),
   Min_Weight = min(X.Weight.kg., na.rm = TRUE),
   Max_Weight = max(X.Weight.kg., na.rm = TRUE),
   IQR_Weight = IQR(X.Weight.kg., na.rm = TRUE),
   Mean_RacesUnder10 = mean(Races.under.10.seconds, na.rm = TRUE),
   Median_RacesUnder10 = median(Races.under.10.seconds, na.rm = TRUE),
   SD_RacesUnder10 = sd(Races.under.10.seconds, na.rm = TRUE),
   Min_RacesUnder10 = min(Races.under.10.seconds, na.rm = TRUE),
   Max_RacesUnder10 = max(Races.under.10.seconds, na.rm = TRUE),
    IQR_RacesUnder10 = IQR(Races.under.10.seconds, na.rm = TRUE),
   Mean_TotalRaces = mean(Total.number.of.races, na.rm = TRUE),
   Median_TotalRaces = median(Total.number.of.races, na.rm = TRUE),
   SD_TotalRaces = sd(Total.number.of.races, na.rm = TRUE),
   Min TotalRaces = min(Total.number.of.races, na.rm = TRUE),
```

```
Max_TotalRaces = max(Total.number.of.races, na.rm = TRUE),
   IQR_TotalRaces = IQR(Total.number.of.races, na.rm = TRUE),
   Mean_DOB = mean(no..Date.of.birth, na.rm = TRUE),
   Median_DOB = median(no..Date.of.birth, na.rm = TRUE),
   SD_DOB = sd(no..Date.of.birth, na.rm = TRUE),
   Min_DOB = min(no..Date.of.birth, na.rm = TRUE),
   Max_DOB = max(no..Date.of.birth, na.rm = TRUE),
   IQR_DOB = IQR(no..Date.of.birth, na.rm = TRUE)
# Print the detailed statistics
print(detailed_stats)
                           SD_PB Min_PB Max_PB IQR_PB Mean_Height Median_Height
##
    Mean_PB Median_PB
## 1 9.918
                 9.94 0.07126706 9.58 9.99 0.0825
                                                        179.6562
   SD_Height Min_Height Max_Height IQR_Height Mean_Weight Median_Weight
## 1 5.979048
                     168
                                196
                                            9
                                                  74.94375
   SD_Weight Min_Weight Max_Weight IQR_Weight Mean_RacesUnder10
## 1 6.761198
                      60
                                 98
                                             9
    Median_RacesUnder10 SD_RacesUnder10 Min_RacesUnder10 Max_RacesUnder10
##
## 1
                      1
                               13.18708
                                                       0
##
    IQR_RacesUnder10 Mean_TotalRaces Median_TotalRaces SD_TotalRaces
                   6
                            123.8688
                                                   112
##
   Min_TotalRaces Max_TotalRaces IQR_TotalRaces Mean_DOB Median_DOB
                                                                       SD_DOB
## 1
                 5
                              403
                                          77.5 31339.29
                                                            32259.5 3886.477
## Min DOB Max DOB IQR DOB
## 1 17055
             37779
# Fit a poisson model
m0 <- glm(Races.under.10.seconds ~ no..Date.of.birth + Total.number.of.races + X.Weight.kg. + Height.cm
summary(m0)
##
## Call:
## glm(formula = Races.under.10.seconds ~ no..Date.of.birth + Total.number.of.races +
      X.Weight.kg. + Height.cm. + PB, family = poisson, data = athletes)
##
## Coefficients:
                          Estimate Std. Error z value Pr(>|z|)
##
## (Intercept)
                         1.167e+02 3.958e+00 29.489 < 2e-16 ***
## no..Date.of.birth
                       7.542e-06 1.090e-05 0.692 0.489003
## Total.number.of.races 6.630e-03 3.870e-04 17.133 < 2e-16 ***
                         2.318e-02 7.014e-03
                                              3.305 0.000948 ***
## X.Weight.kg.
## Height.cm.
                        -2.958e-02 7.367e-03 -4.016 5.93e-05 ***
## PB
                        -1.140e+01 3.328e-01 -34.251 < 2e-16 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## (Dispersion parameter for poisson family taken to be 1)
##
##
      Null deviance: 2475.71 on 159 degrees of freedom
```

```
## Residual deviance: 656.96 on 154 degrees of freedom
## AIC: 1013.5
##
## Number of Fisher Scoring iterations: 5
```

rootogram(m0)



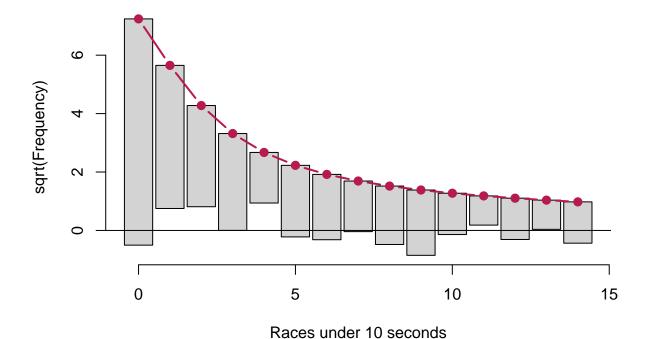
Fit a negative binomial model
m1 <- glm.nb(Races.under.10.seconds ~ no..Date.of.birth + Total.number.of.races + X.Weight.kg. + Height
summary(m1)</pre>

```
##
## Call:
## glm.nb(formula = Races.under.10.seconds ~ no..Date.of.birth +
       Total.number.of.races + X.Weight.kg. + Height.cm. + PB, data = athletes,
##
       init.theta = 1.481785106, link = log)
##
##
## Coefficients:
                           Estimate Std. Error z value Pr(>|z|)
##
## (Intercept)
                          1.773e+02 1.219e+01
                                               14.539 < 2e-16 ***
## no..Date.of.birth
                         -2.109e-06 2.564e-05
                                               -0.082
                                                          0.934
## Total.number.of.races 5.918e-03
                                     1.314e-03
                                                 4.503
                                                        6.7e-06 ***
## X.Weight.kg.
                         -3.480e-03 1.741e-02
                                               -0.200
                                                          0.842
## Height.cm.
                         2.170e-02 1.901e-02
                                                 1.142
                                                          0.254
                         -1.822e+01 1.153e+00 -15.807 < 2e-16 ***
## PB
```

```
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
## (Dispersion parameter for Negative Binomial(1.4818) family taken to be 1)
##
##
      Null deviance: 520.88 on 159 degrees of freedom
## Residual deviance: 163.61 on 154 degrees of freedom
## AIC: 685.25
##
  Number of Fisher Scoring iterations: 1
##
##
##
                 Theta:
                       1.482
                        0.277
##
            Std. Err.:
##
   2 x log-likelihood: -671.246
```

rootogram(m1, main = "Rootogram of Negative Binomial Model", xlab = "Races under 10 seconds")

Rootogram of Negative Binomial Model



```
# 95% Confidence Interval
conf_int <- confint(m1, level = 0.95)</pre>
```

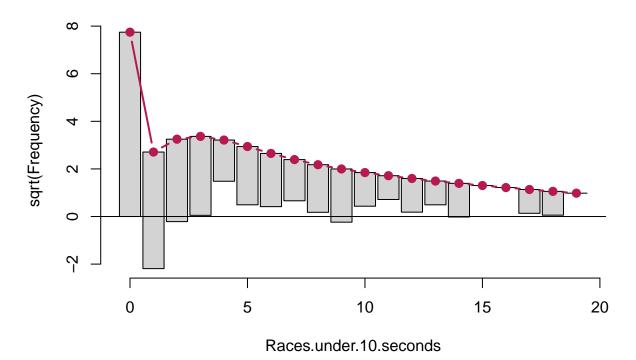
Waiting for profiling to be done...

```
print(conf_int)
##
                                 2.5 %
                                              97.5 %
## (Intercept)
                          1.492405e+02 2.070541e+02
## no..Date.of.birth
                         -5.088661e-05 4.650843e-05
## Total.number.of.races 3.273284e-03 8.754190e-03
                        -3.738078e-02 3.087630e-02
## X.Weight.kg.
## Height.cm.
                         -1.786242e-02 6.152526e-02
## PB
                         -2.118165e+01 -1.544397e+01
library(pscl)
## Registered S3 methods overwritten by 'pscl':
##
    method
                            from
##
     print.zeroinfl
                            countreg
##
    print.summary.zeroinfl countreg
##
     summary.zeroinfl
                            countreg
##
     coef.zeroinfl
                            countreg
##
     vcov.zeroinfl
                            countreg
##
     logLik.zeroinfl
                            countreg
                            countreg
##
    predict.zeroinfl
     residuals.zeroinfl
##
                            countreg
    fitted.zeroinfl
##
                            countreg
##
     terms.zeroinfl
                            countreg
##
    model.matrix.zeroinfl countreg
##
     extractAIC.zeroinfl
                            countreg
##
    print.hurdle
                            countreg
##
    print.summary.hurdle
                            countreg
##
     summary.hurdle
                            countreg
##
     coef.hurdle
                            countreg
##
     vcov.hurdle
                            countreg
##
     logLik.hurdle
                            countreg
##
    predict.hurdle
                            countreg
##
    residuals.hurdle
                            countreg
##
     fitted.hurdle
                            countreg
##
     terms.hurdle
                            countreg
##
    model.matrix.hurdle
                            countreg
##
     extractAIC.hurdle
                            countreg
## Classes and Methods for R developed in the
## Political Science Computational Laboratory
## Department of Political Science
## Stanford University
## Simon Jackman
## hurdle and zeroinfl functions by Achim Zeileis
## Attaching package: 'pscl'
## The following objects are masked from 'package:countreg':
##
       hurdle, hurdle.control, hurdletest, zeroinfl, zeroinfl.control
##
```

```
m2 <- hurdle(Races.under.10.seconds ~ no..Date.of.birth + Total.number.of.races + X.Weight.kg. + Height
## Warning in value[[3L]](cond): system is computationally singular: reciprocal
## condition number = 5.5158e-27FALSE
summary(m2)
## Warning in sqrt(diag(object$vcov)): NaNs produced
##
## Call:
## hurdle(formula = Races.under.10.seconds ~ no..Date.of.birth + Total.number.of.races +
##
       X.Weight.kg. + Height.cm. + PB, data = athletes, dist = "poisson",
##
       zero.dist = "binomial")
##
## Pearson residuals:
                1Q Median
                                3Q
                                       Max
## -3.7330 -0.8296 -0.3776 0.3952 6.9322
##
## Count model coefficients (truncated poisson with log link):
                           Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                          9.919e+01
                                            NA
                                                    NA
                                                             NA
## no..Date.of.birth
                          1.554e-05
                                            NA
                                                    NA
                                                             NA
## Total.number.of.races 5.691e-03
                                            NA
                                                    NA
                                                             NA
## X.Weight.kg.
                          1.926e-02
                                            NA
                                                    NA
                                                             NA
## Height.cm.
                         -2.231e-02
                                            NA
                                                    NA
                                                             NA
## PB
                         -9.711e+00
                                            NA
                                                    NA
                                                             NA
## Zero hurdle model coefficients (binomial with logit link):
##
                          Estimate Std. Error z value Pr(>|z|)
                                                 5.644 1.66e-08 ***
## (Intercept)
                          5.114e+02 9.061e+01
## no..Date.of.birth
                         -1.794e-05
                                           NaN
                                                   NaN
                                                            NaN
## Total.number.of.races 1.335e-02 4.768e-03
                                                 2.799 0.00512 **
## X.Weight.kg.
                         -1.863e-02 4.549e-02 -0.410 0.68213
## Height.cm.
                          6.079e-02 5.075e-02
                                                 1.198 0.23096
## PB
                         -5.243e+01 9.208e+00 -5.694 1.24e-08 ***
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## Number of iterations in BFGS optimization: 15
## Log-likelihood: -430.6 on 12 Df
```

rootogram(m2)



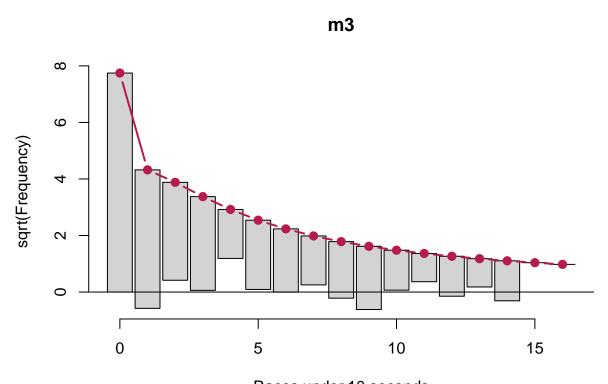


m3 <- hurdle(Races.under.10.seconds ~ no..Date.of.birth + Total.number.of.races + X.Weight.kg. + Height summary(m3)

```
## Warning in sqrt(diag(object$vcov)): NaNs produced
##
## Call:
## hurdle(formula = Races.under.10.seconds ~ no..Date.of.birth + Total.number.of.races +
       X.Weight.kg. + Height.cm. + PB, data = athletes, dist = "negbin",
       zero.dist = "binomial")
##
##
## Pearson residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
## -1.2363 -0.6148 -0.2895 0.3656 3.8141
##
  Count model coefficients (truncated negbin with log link):
##
                           Estimate Std. Error z value Pr(>|z|)
## (Intercept)
                          1.376e+02 1.434e+01
                                                 9.600
                                                       < 2e-16 ***
## no..Date.of.birth
                          6.585e-06
                                           NaN
                                                   NaN
                                                             NaN
## Total.number.of.races 5.216e-03
                                     1.201e-03
                                                  4.342 1.41e-05 ***
## X.Weight.kg.
                         -6.201e-04
                                     1.717e-02
                                                -0.036
                                                           0.971
## Height.cm.
                          9.817e-03
                                     2.082e-02
                                                  0.471
                                                           0.637
## PB
                         -1.402e+01 1.422e+00
                                                -9.860 < 2e-16 ***
## Log(theta)
                          7.026e-01 1.081e-01
                                                 6.497 8.17e-11 ***
## Zero hurdle model coefficients (binomial with logit link):
```

```
Estimate Std. Error z value Pr(>|z|)
##
                         5.114e+02 9.061e+01
## (Intercept)
                                                5.644 1.66e-08 ***
## no..Date.of.birth
                        -1.794e-05
                                          NaN
                                                  NaN
                                                           NaN
## Total.number.of.races 1.335e-02 4.768e-03
                                                2.799
                                                       0.00512 **
## X.Weight.kg.
                        -1.863e-02 4.549e-02
                                               -0.410
                                                       0.68213
## Height.cm.
                         6.079e-02 5.075e-02
                                                1.198 0.23096
## PB
                        -5.243e+01 9.208e+00 -5.694 1.24e-08 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
##
## Theta: count = 2.019
## Number of iterations in BFGS optimization: 32
## Log-likelihood: -322.4 on 13 Df
```

rootogram(m3)

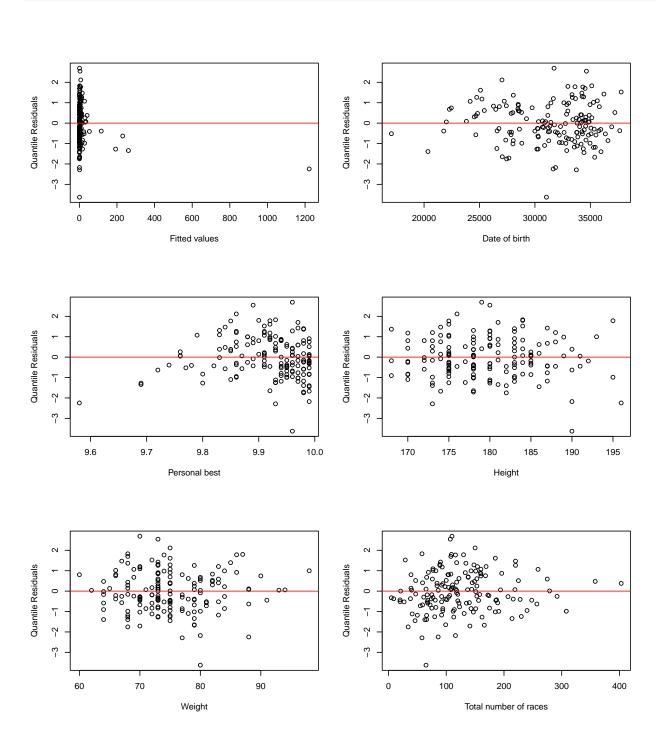


Races.under.10.seconds

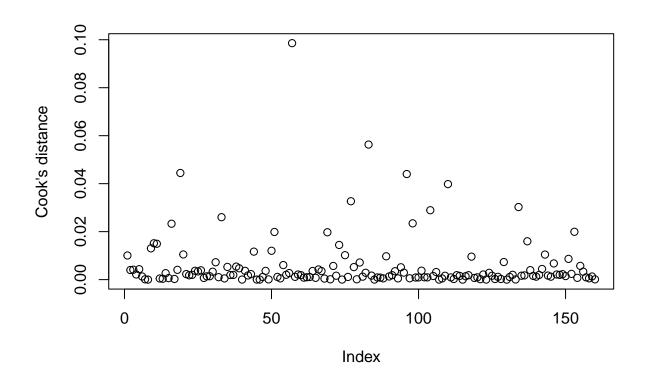
```
# create a new plotting window and set the plotting area into a 2*2 array
par(mfrow=c(3,2))

# Quantile residual plot
qresiduals <- qresid(m1)
plot(m1$fitted.values, qresiduals, xlab="Fitted values", ylab="Quantile Residuals")
abline(h = 0, col = "red")
plot(athletes$no..Date.of.birth, qresiduals, xlab="Date of birth", ylab="Quantile Residuals")
abline(h = 0, col = "red")</pre>
```

```
plot(athletes$PB, qresiduals, xlab="Personal best", ylab="Quantile Residuals")
abline(h = 0, col = "red")
plot(athletes$Height.cm., qresiduals, xlab="Height", ylab="Quantile Residuals")
abline(h = 0, col = "red")
plot(athletes$X.Weight.kg., qresiduals, xlab="Weight", ylab="Quantile Residuals")
abline(h = 0, col = "red")
plot(athletes$Total.number.of.races, qresiduals, xlab="Total number of races", ylab="Quantile Residuals abline(h = 0, col = "red")
```



```
max_position = which.max(qresiduals);max_position
## [1] 110
max_value = max(qresiduals);max_value
## [1] 2.690111
athletes[110,]
##
       Date.of.birth
                                   Name
                                          PΒ
                                                 ID Height.cm. X.Weight.kg.
## 111
          1986-11-11 Ben Youssef Meïté 9.96 67421
                                                           179
##
       Races.under.10.seconds Total.number.of.races no..Date.of.birth NA.
## 111
                                                  110
                                                                   31727 111
# Assuming m1 is your fitted model
predicted_value <- predict(m1, newdata = athletes[19, ], type = "response")</pre>
print(predicted_value)
##
         20
## 7.428351
# Cook's distance
cooks_d <- cooks.distance(m1)</pre>
plot(cooks_d, ylab="Cook's distance", xlab="Index")
```



#abline(h=c(0.5, 1), col="red")

Variance inflation factor vif(m1)

##	noDate.of.birth Total	.number.of.races	X.Weight.kg.
##	1.276440	1.223739	1.981531
##	Height.cm.	РВ	
##	1.802255	1.112980	