Logistic regression AQI data

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Insurance status predicts antiemetic use

\$ sex

\$ ASA

We investigate the Hypothesis that insurance status predicets antiemetic use the population in the Public Use File of the Anestehsia Quality Institute with electronic anesthesia records recording antiemetic use.

Load cleaned dataset fullAQI_4_14 and prov1_AQI_4_14

we load the cleaned larger dataset without and with provider information $fullAQI_4_14.Rdata$, $prov1_AQI_4_14$, which we generated in $import_AQI_14Jul2015.Rmd$

```
rm(list = ls())
load("Analysis/Data/fullAQI_4_14.Rdata")
load("Analysis/Data/prov1_AQI_4_14.Rdata")
str(fullAQI_4_14)
## 'data.frame':
                    173133 obs. of 12 variables:
               : Factor w/ 2 levels "no Ondan", "Ondan": 2 1 2 1 1 2 2 2 1 2 ...
##
   $ ond
## $ dex
               : Factor w/ 2 levels "no Dex", "Dex": 1 1 1 1 1 1 1 1 1 1 ...
## $ drop
               : Factor w/ 2 levels "no Drope", "Drope": 1 1 1 1 1 1 1 1 1 1 ...
               : Factor w/ 2 levels "neither", "either": 2 1 2 1 1 2 2 2 1 2 ...
## $ any
## $ pay
               : Factor w/ 4 levels "Commercial", "MEDICAID", ...: 1 1 1 3 1 3 2 1 3 1 ...
## $ age
               : int 50 53 58 73 64 73 19 27 85 59 ...
## $ age_group: Factor w/ 6 levels "19 - 49", "50 - 64",...: 2 2 2 3 2 3 1 1 4 2 ...
               : Factor w/ 2 levels "female", "male": 2 2 1 2 2 1 1 1 2 2 ...
## $ sex
## $ ASA
               : Factor w/ 6 levels "1", "2", "3", "4", ...: 2 3 3 2 3 3 2 2 3 2 ...
## $ duration : int 59 43 190 56 37 116 93 108 70 93 ...
## $ anes_type: Factor w/ 7 levels "General", "Neuroaxial", ..: 1 4 1 2 3 1 2 1 1 1 ...
## $ practice : Factor w/ 4 levels "A", "B", "C", "D": 2 2 2 2 2 2 2 2 2 2 ...
str(prov1_AQI_4_14)
## 'data.frame':
                    146879 obs. of 13 variables:
## $ ond
               : Factor w/ 2 levels "no Ondan", "Ondan": 2 1 2 2 2 2 1 1 2 2 ...
               : Factor w/ 2 levels "no Dex", "Dex": 1 1 1 1 1 2 1 1 1 2 ...
##
   $ dex
##
               : Factor w/ 2 levels "no Drope", "Drope": 1 1 1 1 1 1 1 1 1 1 ...
   $ drop
## $ any
               : Factor w/ 2 levels "neither", "either": 2 1 2 2 2 2 1 1 2 2 ...
               : Factor w/ 4 levels "Commercial", "MEDICAID", ...: 3 2 1 1 3 1 1 1 3 1 ...
##
  $ pay
##
               : int 73 31 56 59 64 49 36 45 62 26 ...
## $ age_group: Factor w/ 6 levels "19 - 49", "50 - 64", ...: 3 1 2 2 2 1 1 1 2 1 ...
```

\$ anes_type: Factor w/ 7 levels "General", "Neuroaxial",..: 1 1 1 1 1 1 1 1 1 1 ...

\$ practice : Factor w/ 4 levels "A", "B", "C", "D": 2 2 2 2 2 2 2 2 2 2 ...

: Factor w/ 2 levels "female", "male": 1 1 2 2 1 2 2 1 1 2 ...

\$ duration : int 172 80 172 122 133 83 136 81 112 110 ...

: Factor w/ 6 levels "1", "2", "3", "4", ...: 4 3 2 2 3 1 2 2 3 2 ...

\$ provider : Factor w/ 720 levels "5622", "5623", ...: 161 155 152 163 156 156 161 161 156 153 ...

Logistic Regression

Logistic Model 1 Link: logit

We first fit a logistic regression model with the a \mathbf{logit} link, saved in $LogisticRegression_AQIFixedEfffects_LogitLink$ as shown below:

Summary: Controlling for age, sex, facilty and case duration, antiemetic administration is strongly associated with insurance status as a marker of SES, but the OR is not as strong as previously estimated controlling for fewer variables. it makes sense that controlling for likely confounders like anesthesia type and ASA status reduces the effect estimate.

Logistic Model 2 Link: log

We try to fit a logistic regression model with the a log link.

Problem: We are unable to fit a log link model with all predictors, because we get the error message:

Error: no valid set of coefficients has been found: please supply starting values

if we use more predictors than a few predictors. Two solutions:

- supply starting values for all coefficients, (factors need one for each level)
- consider using the package biglm, specifically for log link with large data set

Starting values for log link glm

Estimate fit with **one less** predictor:

Use estimated coefficients as **starting values** for model with one additional predictor. Starting values for additional predictors are set to zero.

	OR	X2.5	X97.5	p_values
(Intercept)	0.702	0.692	0.711	0.000
payMEDICAID	0.838	0.825	0.850	0.000
payMedicare	0.989	0.977	1.002	0.113
paySELF	1.080	1.038	1.120	0.000
$age_group50 - 64$	1.143	1.130	1.156	0.000
$age_group65$ - 79	1.007	0.993	1.022	0.325

	OR	X2.5	X97.5	p_values
age_group80+	0.819	0.799	0.839	0.000
$age_groupUnder 1$	0.337	0.311	0.364	0.000
$age_group1-18$	1.070	1.052	1.089	0.000
sexmale	1.003	0.994	1.012	0.570
ASA2	0.824	0.813	0.836	0.000
ASA3	0.803	0.791	0.815	0.000
ASA4	0.367	0.358	0.376	0.000
ASA5	0.025	0.015	0.038	0.000
ASA6	0.032	0.002	0.137	0.001