

# Import AQI data

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*Friday, June 19, 2015*

## Insurance status predicts antiemetic use

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

## Data import

we load the original dataset and save as it as *PUF\_Q4\_2013.Rdata*

### Original Data

```
# run only once
PUF_Q4_2013 <- read.csv("Analysis/Data/PUF_Q4_2013_Antiemetic.csv")
save(PUF_Q4_2013, file="Analysis/Data/PUF_Q4_2013.Rdata")
```

### Load Rdata AQI raw dataset *PUF\_Q4\_2013.Rdata*

```
rm(list = ls())
load("Analysis/Data/PUF_Q4_2013.Rdata")
```

## Clean Data

### Predictor: insurance status

The predictor insurance status (*Payment*) is coded in 4 levels as Commercial, MEDICAID, Medicare, SELF, after we removed 265311 cases without insurance information, (originally coded as “”), with 176334 unique cases remaining. (False indicating no NAs in Payment: FALSE)

### Outcome variables: antiemetic administration

We focus on the antiemetics *ondansetron*, *dexamethason* and *droperidol*, the only agents with convincing evidence for effect.

Table 1: Cases with Ondansetron versus Dexamethason

	no Dex	Dex
no Ondan	79842	4873
Ondan	61191	30428

Table 2: Cases with Ondansetron versus Droperidol

	no Drope	Drope
no Ondan	84398	317
Ondan	89067	2552

Table 3: Cases with Dexamethason versus Droperidol

	no Drope	Drope
no Dex	140010	1023
Dex	33455	1846

The antiemetics *ondansetron* and *dexamethason* were sometimes administered together. This is coded in *ondan\_dex\_either*

## Potential confounders and other variables

### practice ID versus facility ID

	193055	691419	5013437	5610264
136085	0	23241	0	0
1116623	36127	0	0	0
9485541	0	0	0	94080
23100212	0	0	79	0
46100453	0	0	15040	0
53228659	0	0	1	0
71100339	0	0	7766	0

The table of facility ID versus practice ID suggests that five practices have only one facility ID and one practice (=5013437) has three (sub) facilities. We will simplify by using practice ID, which has no NA.

### case\_duration\_minutes

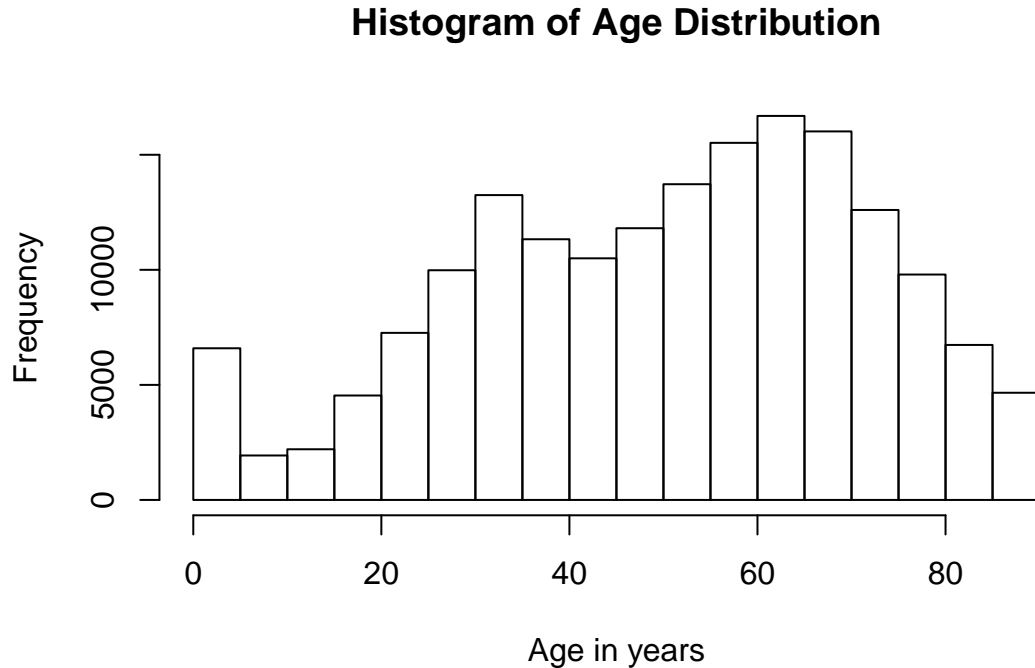
```
PUF_Q4_2013$case_duration_minutes [PUF_Q4_2013$case_duration_minutes == -1] <- NA
missing <- sum(PUF_Q4_2013$case_duration_minutes==-1)
PUF_Q4_2013 <- PUF_Q4_2013 [complete.cases(PUF_Q4_2013$case_duration_minutes),]
```

Case duration in minutes (*case\_duration\_minutes*) is an integer and has NA missing values coded as -1, which we recoded as NA and removed from the dataset, leaving 175123 unique cases.

### patient age

```
PUF_Q4_2013$patient_age[PUF_Q4_2013$patient_age==-1] <- NA
hist(PUF_Q4_2013$patient_age,
```

```
main = "Histogram of Age Distribution",
xlab = "Age in years")
```



```
missing <- sum(patient_age== -1)
PUF_Q4_2013 <- PUF_Q4_2013[complete.cases(PUF_Q4_2013$patient_age),]
```

Patient age (*patient\_age*) is an integer with a distribution above and has 8 missing values coded as -1, which we recoded as NA and removed from the dataset, leaving 175115 unique cases.

**patient\_age\_group**

```
levels(PUF_Q4_2013$patient_age_group)[2] <- "1-18"
levels(PUF_Q4_2013$patient_age_group)[1] <- NA
missing <- sum(is.na(PUF_Q4_2013$patient_age_group))
which(is.na(PUF_Q4_2013$patient_age_group))
```

```
## integer(0)
```

```
PUF_Q4_2013 <- PUF_Q4_2013[complete.cases(PUF_Q4_2013$patient_age_group),]
```

Patient age group (*patient\_age\_group*) is a factor with 6 levels: 1-18, 19 - 49, 50 - 64, 65 - 79, 80+, Under 1; it has 0 missing values, leaving 175115 unique cases.

(Missing values were initially coded as -1, which we recoded as NA and removed as a level; we corrected the miscoding from "18-Jan" to "1-18").

### **patient\_sex**

Patient gender (*patient\_sex*) is recoded as factor with the two levels female, male and 18 NAs, which are removed from the dataset, leaving 175097 unique cases.

### **in\_or\_out\_patient**

in- or outpatient status (*in\_or\_out\_patient*) is recoded as a factor with the two levels Outpatient, Inpatient and 62432 NAs, which are too numerous to include this variable in the dataset.

### **surgical\_cpt code**

We considered to control with a random effect for *surgical\_cpt* code but 58916 cases do not have a *surgical\_cpt* code defined, which are too many to exclude.

### **combined\_cpt code**

We considered to control with a random effect for *combined\_cpt* code but 71552 cases do not have a *combined\_cpt* code defined, which are too many to exclude.

### **reported\_anesthesia\_code**

We considered to control with a random effect for *reported\_anesthesia\_code* code but 174153 cases do not have a *reported\_anesthesia\_code* code defined, which are too many to exclude.

### **primary\_anesthesia\_type**

*primary\_anesthesia\_type* is recoded as a factor with 7 levels [General, Epidural/Spinal, Regional, Monitored Anesthesia Care, Sedation, Local, Other]. We considered to control with a fixed or a effect for *primary\_anesthesia\_type* code but 1914 cases do not have a *primary\_anesthesia\_type* code defined, which may be too many to exclude.

We did exclude NA leaving us with 173183 unique cases.

### **procedure\_status**

It would make sense to try to control for *procedure\_status*, (which indicates if the case was Emergency or Elective); but 139046 of the remaining cases do not have a *procedure\_status* code defined, which obviously are too many to exclude.

### **case\_type**

It would make sense to try to control for *case\_type*, (which indicates if the case was Non - OR, OB/GYN NON Surgical, OB/GYN Surgical, OR, OTHER ...), but 57743 of the remaining cases do not have a *case\_type* code defined, which obviously are too many to exclude.

**asaps\_imputed**

**asaps**

It would make sense to try to control for *asaps* or *asaps\_imputed*, (ASA Status, which indicates how sick a patient was, and only 50 of the remaining cases do not have an ASA status recorded; so we exclude them, leaving us with 173133 unique cases with also *asaps* as a predictor.

**prov1**

It would be great to control for individual provider behavior, to show variability among providers in their propensity to administer antiemetics contingent on insurance status. There are 720 different *prov1* levels, I believe they are coding for individual providers. 26254 of the remaining cases do not have the *prov1* recorded; if we exclude them, it leaves us with 146879 unique cases with provider coded as *prov1* as predictor.

## Save cleaned datasets

### Larger dataset without provider information in *fullAQI\_4\_14*

a clean dataframe without provider info but more unique cases is saved as *fullAQI\_4\_14.Rdata*

```
## 'data.frame': 173133 obs. of 12 variables:
## $ ondansetron : Factor w/ 2 levels "no Ondan","Ondan": 2 1 2 1 1 2 2 2 1 2 ...
## $ dexamethason : Factor w/ 2 levels "no Dex","Dex": 1 1 1 1 1 1 1 1 1 1 ...
## $ droperidol : Factor w/ 2 levels "no Drope","Drope": 1 1 1 1 1 1 1 1 1 1 ...
## $ ondandex_either : Factor w/ 2 levels "neither","either": 2 1 2 1 1 2 2 2 1 2 ...
## $ Payment : Factor w/ 4 levels "Commercial","MEDICAID",...: 1 1 1 3 1 3 2 1 3 1 ...
## $ patient_age : int 50 53 58 73 64 73 19 27 85 59 ...
## $ patient_age_group : Factor w/ 6 levels "1-18","19 - 49",...: 3 3 3 4 3 4 2 2 5 3 ...
## $ patient_sex : Factor w/ 2 levels "female","male": 2 2 1 2 2 1 1 1 2 2 ...
## $ asaps_imputed : Factor w/ 5 levels "3","4","5","6",...: 5 1 1 5 1 1 5 5 1 5 ...
## $ case_duration_minutes : int 59 43 190 56 37 116 93 108 70 93 ...
## $ primary_anesthesia_type: Factor w/ 7 levels "General","Epidural/Spinal",...: 1 4 1 2 3 1 2 1 1 1 ...
## $ practiceID : Factor w/ 4 levels "193055","691419",...: 2 2 2 2 2 2 2 2 2 2 ...
```

### Smaller dataset with provider information in *prov1\_AQI\_4\_14*

a more limited dataset with the individual provider as predictor is saved (after removing cases with *prov1* NA) in *prov1\_AQI\_4\_14.Rdata*

```
## 'data.frame': 146879 obs. of 13 variables:
## $ ondansetron : Factor w/ 2 levels "no Ondan","Ondan": 2 1 2 2 2 2 1 1 2 2 ...
## $ dexamethason : Factor w/ 2 levels "no Dex","Dex": 1 1 1 1 1 2 1 1 1 2 ...
## $ droperidol : Factor w/ 2 levels "no Drope","Drope": 1 1 1 1 1 1 1 1 1 1 ...
## $ ondandex_either : Factor w/ 2 levels "neither","either": 2 1 2 2 2 2 1 1 2 2 ...
## $ Payment : Factor w/ 4 levels "Commercial","MEDICAID",...: 3 2 1 1 3 1 1 1 3 1 ...
## $ patient_age : int 73 31 56 59 64 49 36 45 62 26 ...
## $ patient_age_group : Factor w/ 6 levels "1-18","19 - 49",...: 4 2 3 3 3 2 2 2 3 2 ...
## $ patient_sex : Factor w/ 2 levels "female","male": 1 1 2 2 1 2 2 1 1 2 ...
## $ asaps_imputed : Factor w/ 5 levels "3","4","5","6",...: 2 1 5 5 1 5 5 5 1 5 ...
```

```
## $ case_duration_minutes : int 172 80 172 122 133 83 136 81 112 110 ...
## $ primary_anesthesia_type: Factor w/ 7 levels "General","Epidural/Spinal",...: 1 1 1 1 1 1 1 1 1 1 .
## $ prov1                  : Factor w/ 720 levels "5622","5623",...: 161 155 152 163 156 156 161 161 1
## $ practiceID             : Factor w/ 4 levels "193055","691419",...: 2 2 2 2 2 2 2 2 2 ...
```