# Import AQI data

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# Insurance status predicts antiemetic use

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.

# 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_Antimetic.csv")
save(PUF_Q4_2013, file="Analysis/Data/PUF_Q4_2013.Rdata")</pre>
```

# 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<br>Ondan | 84398<br>89067 | $\frac{317}{2552}$ |

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),]
```

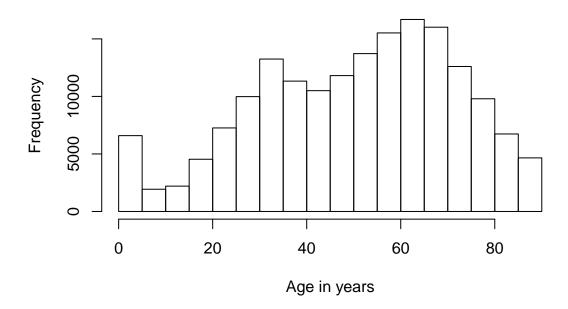
Case duration in minutes (case\_duration\_minutes) is an integer and has NA missing values coded as -1, which we recoded as NA and removed from the dataset, leaving 175123 unique cases.

#### patient age

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

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

# **Histogram of Age Distribution**



```
missing <- sum(patient_age==-1)
PUF_Q4_2013 <- PUF_Q4_2013[complete.cases(PUF_Q4_2013*patient_age),]</pre>
```

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))</pre>
```

## 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 exclude.

#### surgical\_cpt code

We considered to control with a random effect for  $surgical\_ctp$  code but 58916 cases do not have a  $surgical\_ctp$  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.

# ${\bf primary\_anesthesia\_type}$

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 admister 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

```
173133 obs. of 12 variables:
## 'data.frame':
##
   $ ondansetron
                              : Factor w/ 2 levels "no Ondan", "Ondan": 2 1 2 1 1 2 2 2 1 2 ...
                              : Factor w/ 2 levels "no Dex", "Dex": 1 1 1 1 1 1 1 1 1 1 ...
## $ dexamethason
## $ droperidol
                              : Factor w/ 2 levels "no Drope", "Drope": 1 1 1 1 1 1 1 1 1 1 ...
## $ ondan_dex_either
                              : Factor w/ 2 levels "neither", "either": 2 1 2 1 1 2 2 2 1 2 ...
                              : Factor w/ 4 levels "Commercial", "MEDICAID", ...: 1 1 1 3 1 3 2 1 3 1 ....
## $ Payment
## $ patient_age
                              : int 50\ 53\ 58\ 73\ 64\ 73\ 19\ 27\ 85\ 59\ \dots
                              : Factor w/ 6 levels "1-18",
"19 - 49",...: 3 3 3 4 3 4 2 2 5 3 ....
   $ patient_age_group
##
## $ patient_sex
                              : Factor w/ 2 levels "female", "male": 2 2 1 2 2 1 1 1 2 2 ...
## $ asaps
                              : Factor w/ 6 levels "1", "2", "3", "4", ...: 2 3 3 2 3 3 2 2 3 2 ...
## $ 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 ...
                             : Factor w/ 2 levels "neither", "either": 2 1 2 2 2 2 1 1 2 2 ...
## $ ondan_dex_either
## $ 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 ...
                             : Factor w/ 6 levels "1-18", "19 - 49", ...: 4 2 3 3 3 2 2 2 3 2 ...
## $ patient_age_group
                             : Factor w/ 2 levels "female", "male": 1 1 2 2 1 2 2 1 1 2 ...
## $ patient_sex
## $ asaps
                             : Factor w/ 6 levels "1","2","3","4",..: 4 3 2 2 3 1 2 2 3 2 ...
```