

# Midterm Project - Dangerous Workplaces

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## I. Introduction

What is the most dangerous workplace in Massachusetts? To answer this, we use the U.S. Dept. of Labor's OSHA database to compile, organize, and present accident and violation data for over 1000 workplaces in MA. The data span more than four decades of statistics, starting with the 1970s. The goal is to organize these data into meaningful sets that an analyst can manipulate without encountering errors. Such data cleaning mainly involves tidying data tables, removing/changing missing values, separating data into defined categories, etc. The following report is the steps that have been taken to clean up several tables that originally contained raw data. A short exploratory analysis is performed to gauge the cleanliness of the data.

## II. Data Cleaning

### A. Raw data

The files used for this preparation are:

1. accid.dbf
2. viol.dbf
3. osha.dbf
4. acc.dbf
5. hazsub.dbf
6. hzs.dbf
7. sic.dbf

The first three files provide the details of OSHA inspections. All of the data are coded (i.e. body parts are coded with numbers 1-31, injuries are coded 1-18, etc.). 'acc.dbf' contains the references for these codes. 'hazsub.dbf' and 'hzs.dbf' give the codes for hazardous substances, and 'sic.dbf' gives the industry type (fishing, repair, construction, etc.)

### B. Cleaning up accid.dbf

This file contains 2147 observations of accidents and related details. With 16 variables, the table holds a large amount of information; however, not all variables are needed.

After checking that every single value in the STATE column was the same, the column was removed. All of these incidents occurred in MA.

The DBF file has several columns but the ones that were appropriate for describing an accident are:

1. ACTIVITYNO: unique identifier for a given record
2. DEGREE: fatal or non-fatal
3. NATURE: type of injury
4. BODYPART: body part(s) affected
5. EVENT: circumstances of injury
6. SOURCE: source of injury (equipment, vehicle, etc.)
7. ENVIRON: location of injury

8. HUMAN: presence/absence of human error
9. TASK: regularly scheduled task or not
10. HAZSUB: hazardous substance present or not

Upon inspection, the last column, HAZSUB, seemed to have missing values. It was assumed that if HAZSUB = NA, then no hazardous substance was present during the accident. The total NA's in each column was counted. The table was filtered and all incidents containing hazardous substances was moved to a new DBF file. Hereafter, both the original accidents DBF (without hazardous substances) and the new file (with hazardous substances) are worked with simultaneously.

Both tables were arranged by DEGREE (1, 2, 3). It was found that the accidents without hazardous substances contained incidents of zero degree. From the README files for the original data sets, zero values were undefined. It was inferred that a degree of zero represents an incident without injuries. These rows were moved to a new table and removed from the original tables.

Using the hzs.dbf file, the coded HAZSUB column was replaced by its corresponding hazardous substances. Not all hazardous substances had a matching code so this joining procedure produced missing values. These rows cannot be omitted as they can be used for analysis so the table was arranged such that these missing values would be towards the end of the table.

### C. Cleaning up viol.DBF

This table contains the violations for each company as well as more than 30 other variables. To determine the hazardous conditions of a site, only the following columns were maintained:

1. ACTIVITYNO: same unique identifier as in accid.DBF
2. ISSUEDATE: date of issue, useful for time series
3. ITEMNO: tally of repeat violations
4. GRAVITY: seriousness of violation
5. VIOLTYPE: type of violation (serious, willful, repeat, etc.)
6. INSTANCES: total number of violations

Some of the violations in this table were marked as “deleted due to some reason other than failure to submit information.” These rows were removed.

After performing the above cleanup, all rows were checked for missing values; none were found, except in the GRAVITY column. Some duplicates were found, and were deleted. The GRAVITY column provides meaningful information so as we did with the accidents file, rows containing GRAVITY = NA were moved to the bottom of the table.

### D. Cleaning up osha.DBF

This is likely the most comprehensive of all of the tables. It contains summaries of all of the other tables. The columns of interest are:

1. ACTIVITYNO: unique identifier
2. ESTABNAME: name of workplace
3. SITEADD: address
4. SITEZIP: zipcode
5. SIC: industry code
6. ACCID\_: number of accidents (in accid.dbf)
7. VIOLS\_: number of violations (in viol.dbf)
8. HAZSUB\_: number of hazardous substances

If a place has no violations, no accidents, and no hazardous substances, then we can assume it is a safe place. These rows were deleted because they would not assist in determining how dangerous other places are.

Using sic.dbf, the sic codes were replaced with their industry (i.e. the coded numbers were changed to strings such as “FISHING” or “REPAIR”).

## **E. Joining to make comprehensive tables**

All three tables have a common identifier of ACTIVITYNO. Using this, the site names and addresses and industries (and all other relevant information) were attached to each table. The result was four large comprehensive tables:

1. Accident
2. Accidents (without hazardous substances)
3. Violations
4. Summary (from osha.DBF)

## **F. Replacing the codes**

As we did with the SIC codes and HAZSUB codes, using the acc.dbf code lookup file, all coded values in the accidents and violations files were replaced by their corresponding strings. The data tables are now large tables of categorical data. This may make it difficult to work with, but one can easily factor each column and assign a numerical value to each level as needed.

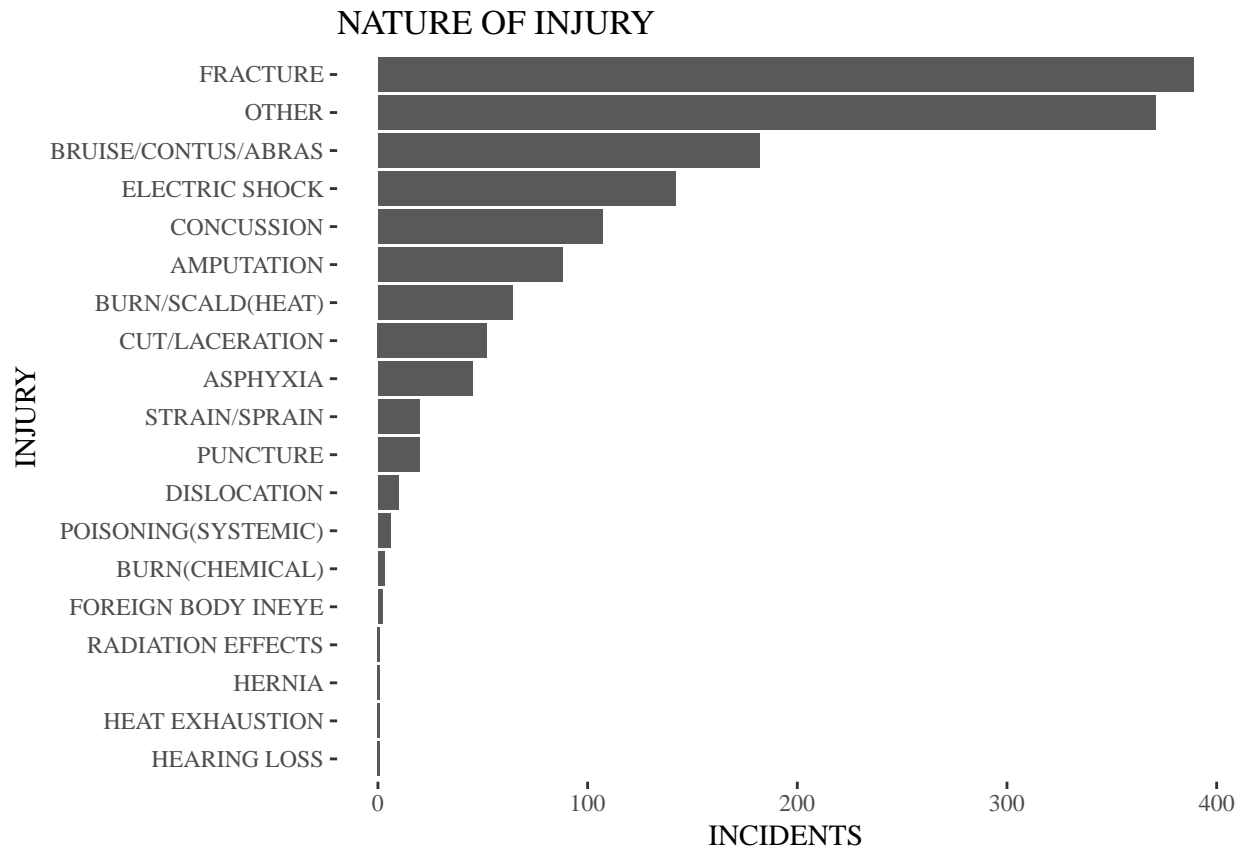
# **III. Exploratory Analysis**

## **A. Frequency of appearance**

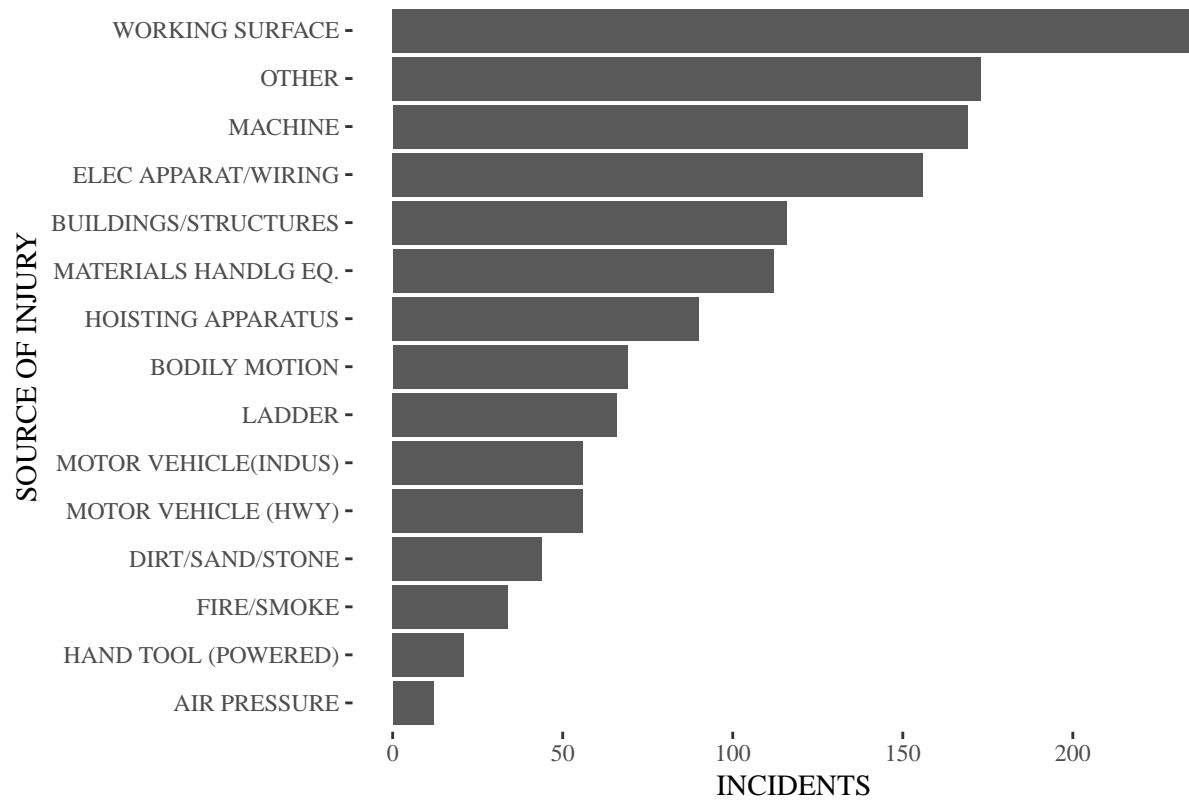
The frequency of appearance for the variables in accid.dbf were measured. Histograms of a few variables are shown below.

Note: For variables with more than 30 levels, only the top 15 are shown in the plots below to reduce clutter.

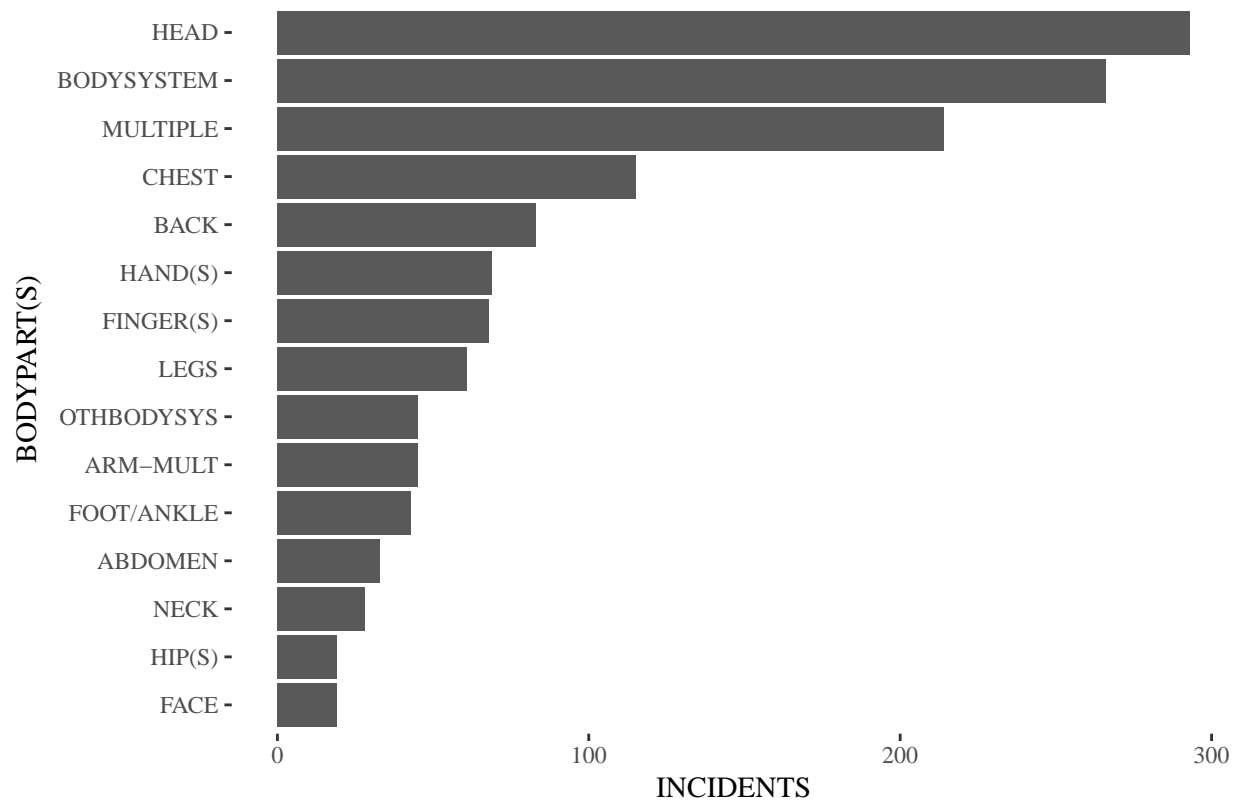


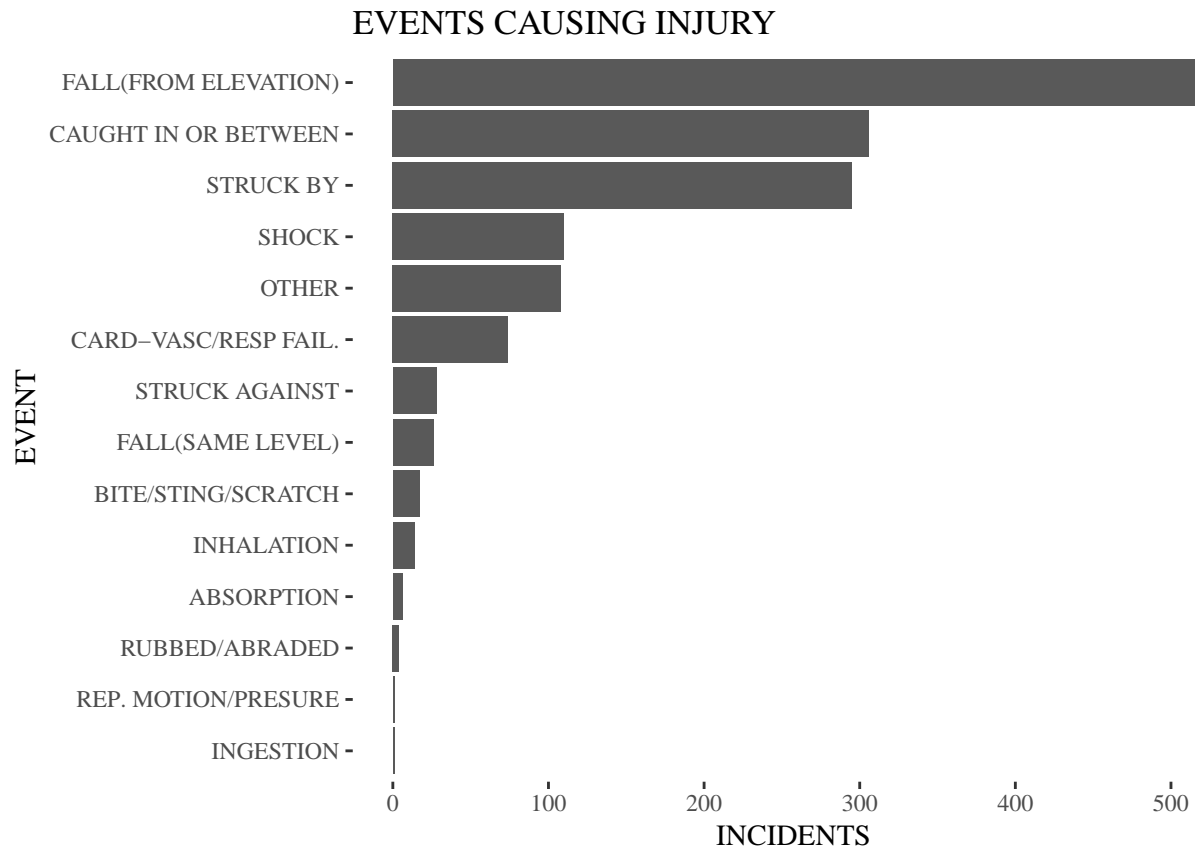


## TOP 15 SOURCES OF INJURY

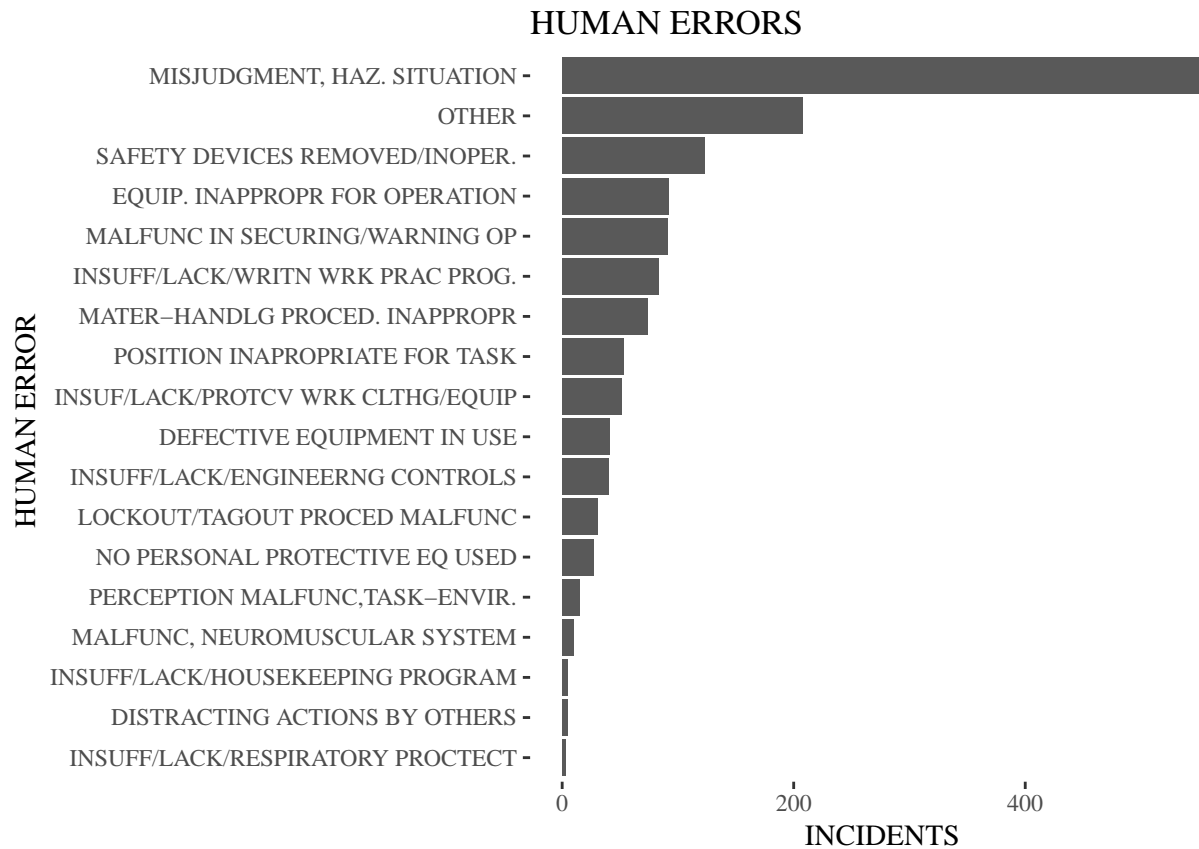


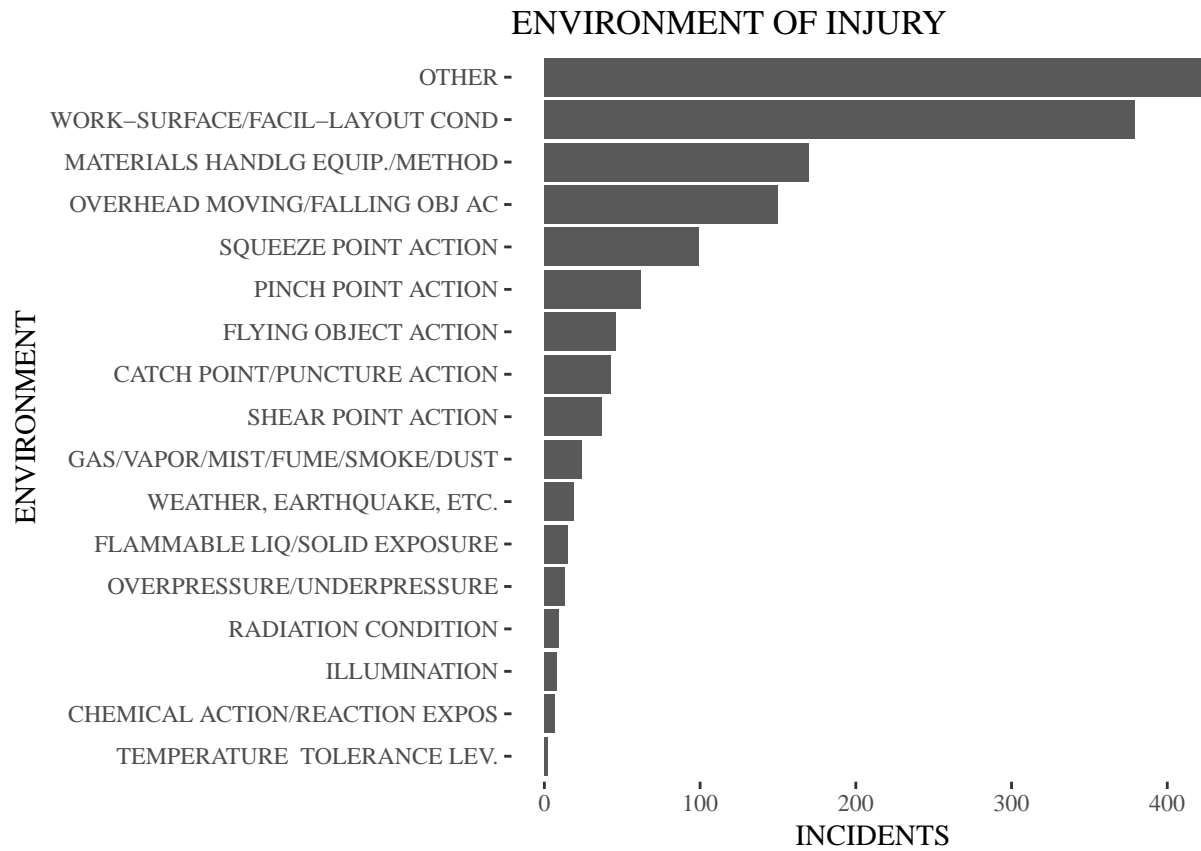
## TOP 15 BODYPARTS AFFECTED

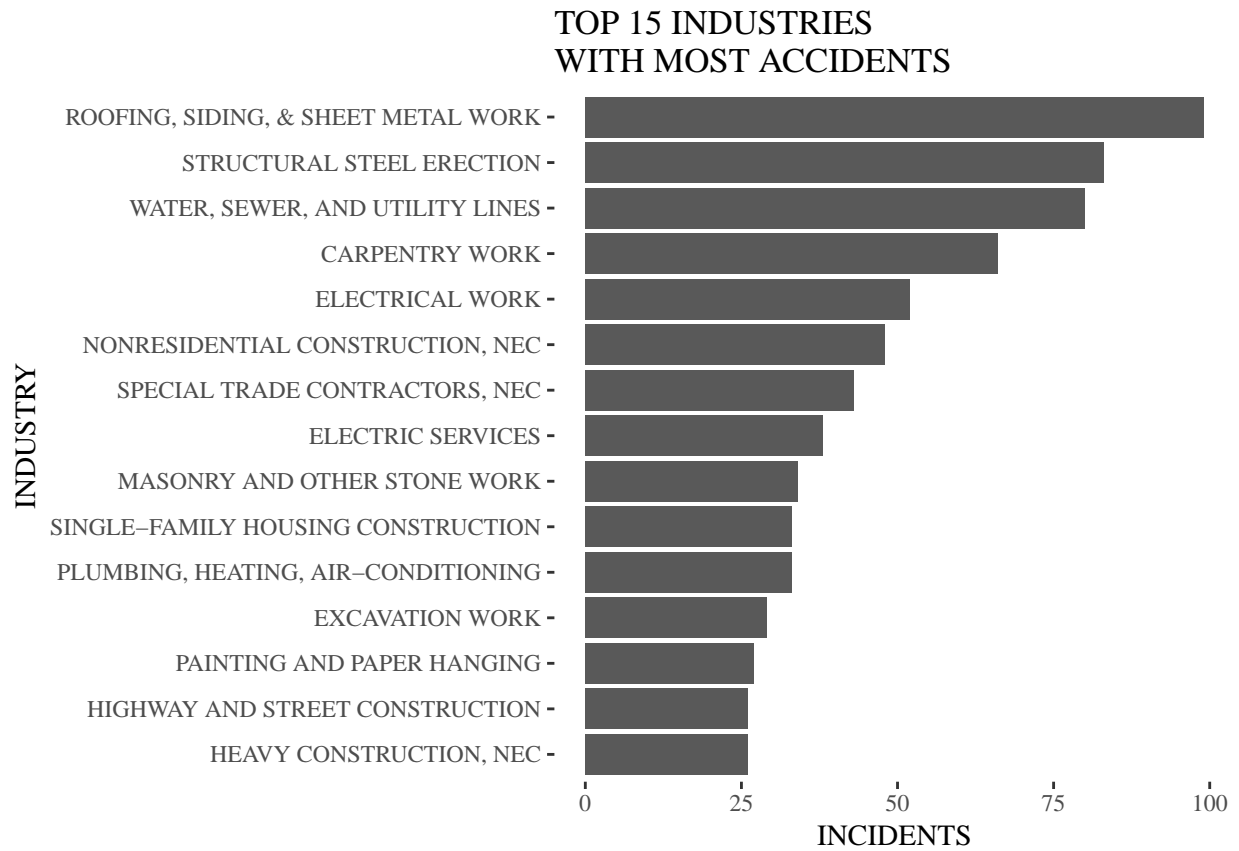


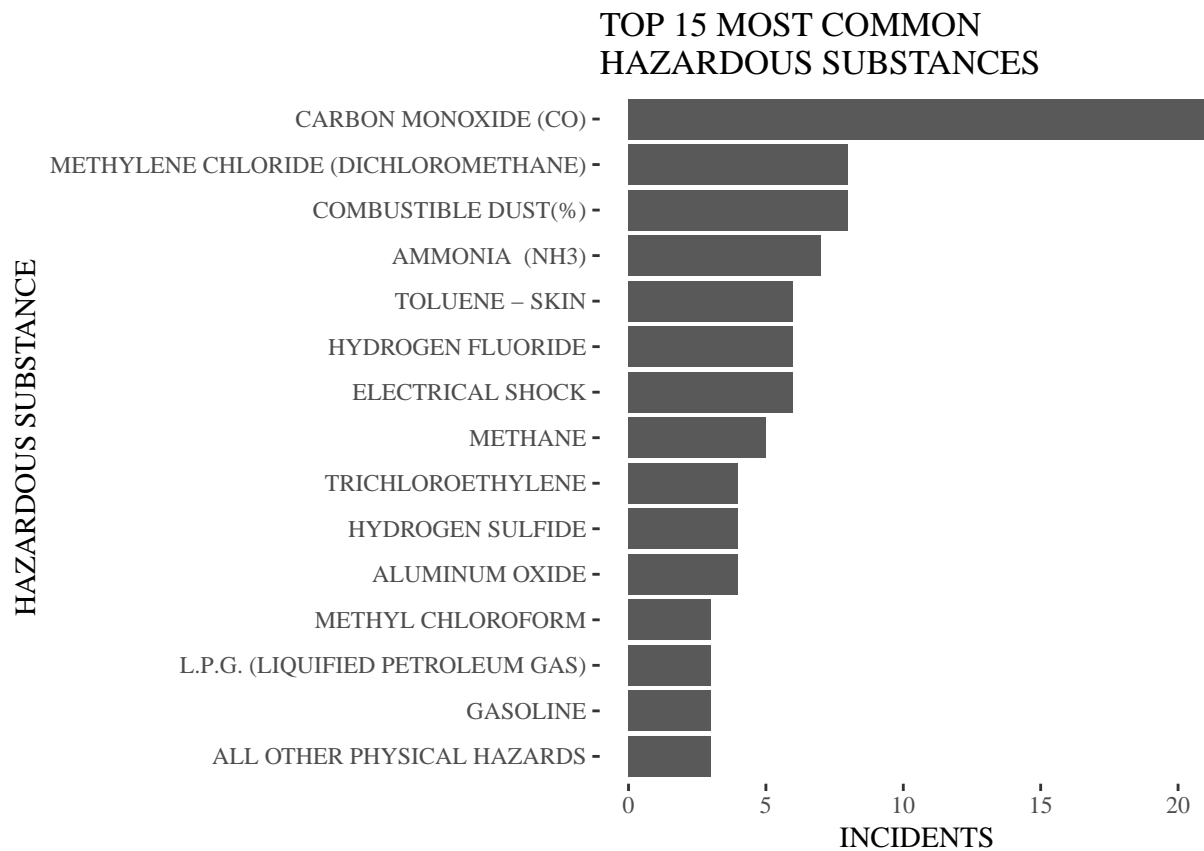




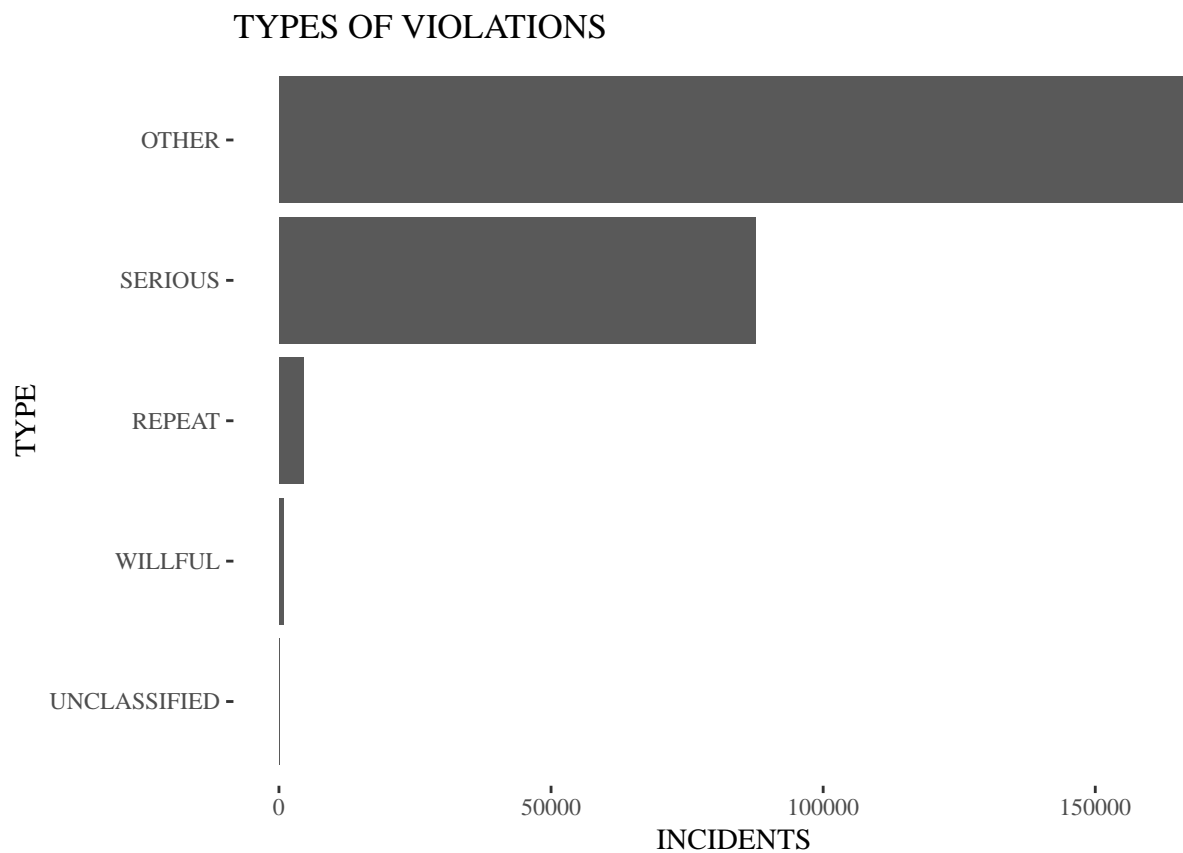








The frequencies of the violation types are shown below. Most of the violations fall into the “OTHER” category in the raw data file.



## B. Most dangerous places by total number of deaths

Here are the top five establishments with the most deaths

SITE	INCIDENTS
GENERAL DYNAMICS QUINCY SHIPBU	6
BOSTON EDISON CO	5
GENERAL ELECTRIC CO	4
MASSACHUSETTS ELECTRIC CO	3
PERINI CORP	3

Here are the top five establishments with the most deaths not caused by human error. This table was produced after filtering accid.DBF by human factors not related to human error. These were:

1. EQUIP. INAPPROPR FOR OPERATION
2. INSUF/LACK/PROTCV WRK CLTHG/EQUIP
3. INSUFF/LACK/ENGINEERNG CONTROLS
4. INSUFF/LACK/HOUSEKEEPING PROGRAM
5. INSUFF/LACK/RESPIRATORY PROTECT
6. INSUFF/LACK/WRITN WRK PRAC PROG
7. DEFECTIVE EQUIPMENT IN USE
8. MALFUNC IN SECURING/WARNING OP
9. MATER-HANDLG PROCED. INAPPROPR
10. SAFETY DEVICES REMOVED/INOPER.

SITE	INCIDENTS
GENERAL DYNAMICS QUINCY SHIPBU	2
GENERAL ELECTRIC CO	2
PAR ELECTRICAL CONTRACTORS INC	2
PLYMOUTH & BROCKTON STREET RAI	2
A & M ROOFING & SHEETMETAL CO.	1

## C. Interpretation

All of the distributions seem to be skewed. However, they are skewed towards the incident that is most logical in the context of these workplaces. Most of these companies handle building or repair or technical/vocational work that involve physical labor. As a result, one expects fractures, falls, damage to upper body/head, to be most common. The most common incidents are indeed what is expected. Accidents involving hazardous substances a slightly difficult to analyze because one cannot be sure what substance is most toxic. Still however, that carbon monoxide is the most toxic substance agrees with the fact that carbon monoxide is one of the more prevalent toxic gases in any establishment.

Based on the plots above the frequency of incidents, we may infer that most accidents are due to falls from elevated working surfaces that fractures mostly the upper body and head, and are due to workers' misjudgment of hazardous situations in companies that specializes in roofing, siding, and sheet metal work.

That a "misjudgement of hazardous situation" is by far the most common human error provides a great deal of insight as to whether a not a workplace is truly dangerous.

That the 'OTHER' violation type is the most common points to the incompleteness of the original raw data files. More data is needed to assign violation type(s) to these incidents.

The two tables on the previous page are an example of how the data set can be manipulated to produce data tables. One can filter deaths by removing human error and then calculate the proportion of deaths that are not caused by human error in determining the hazards in a given workplace. Based on how these data are cleaned, the analyst can choose from a variety of approaches to analyze the data.