

# Increases in Drug and Opioid Overdose Deaths — United States, 1999–2014

Arun Rajan, Charan Teja Lellaboyena

May 15, 2017

## 1 Introduction

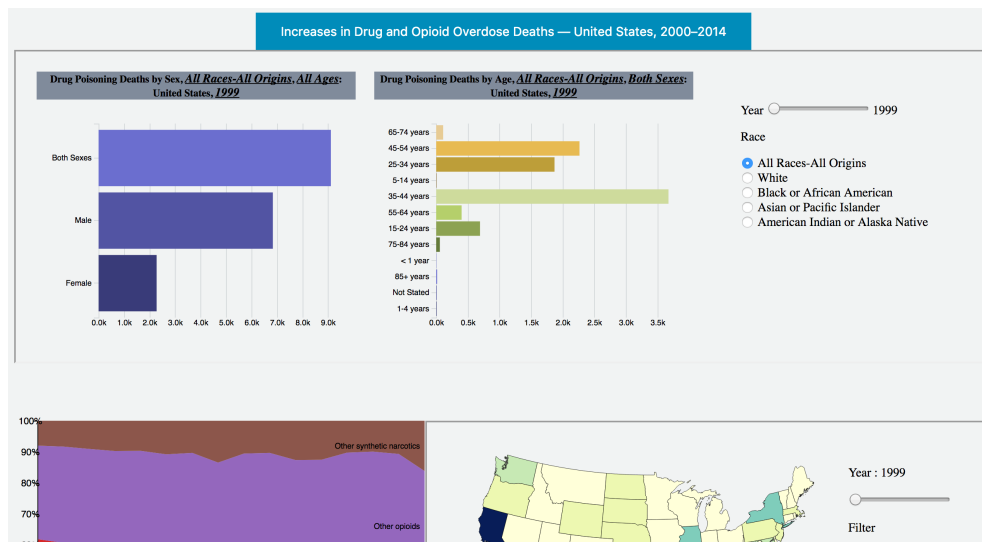
Opioid addiction and death rates in the U.S. and abroad have reached "epidemic" levels. Every single day, many lives are lost due to careless and unfortunate habits of drugs use. As per a report from 'Center for Disease Control Prevention' the United States is experiencing an epidemic of drug overdose (poisoning) deaths. Since 2000, the rate of deaths from drug overdoses has increased 137%, including a 200% increase in the rate of overdose deaths involving opioids[2] (opioid pain relievers and heroin)[1]. Especially, overdose death linked to prescription opioids is a growing public health crisis. The current epidemic accelerated during a fifteen-year interval between 1999 and 2014. In particular, the mortality rate from drug overdose tripled between 1999 and 2014 (CDC, 2015): by 2014, the number of overdose deaths in the United States exceeded the number of suicides; by 2014, they also exceeded the number of motor vehicle deaths. Prescription opioid analgesics, painkillers derived from opium or synthesized to possess narcotic properties similar to opioids, including oxycodone, hydrocodone and codeine, have driven this overdose epidemic (CDC, 2011; Warner et al., 2009). Prescription opioids may have been obtained legally through a prescription, or diverted through illegal distribution networks. This document captures the final report made by our team for the final project in CSE-564, Visualization. We discuss the different approaches, ideas and challenges that we met during its completion.

## 2 Progress

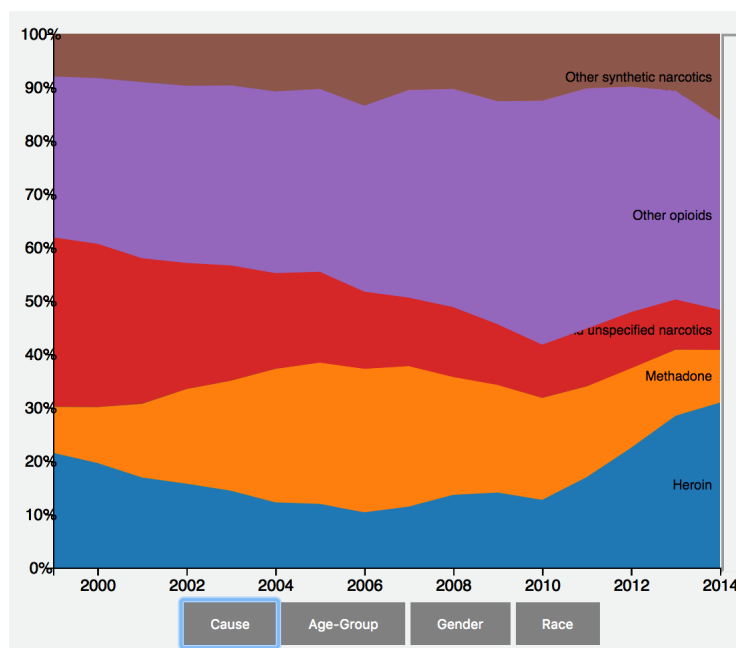
### 2.1 Insights & Visual Analysis

We brainstormed on the different dimensions that could be shown on the dashboard. These are majorly divided into Geographical and temporal divisions. Further, we have tried to correlate two different data sets, related to retail sales of opioid based analgesic drugs and deaths related to Opioid overdoses.

We started using a Top-down approach, first showing trends over the time period of 15 years. The visualization shown below, that is part of the dashboard,

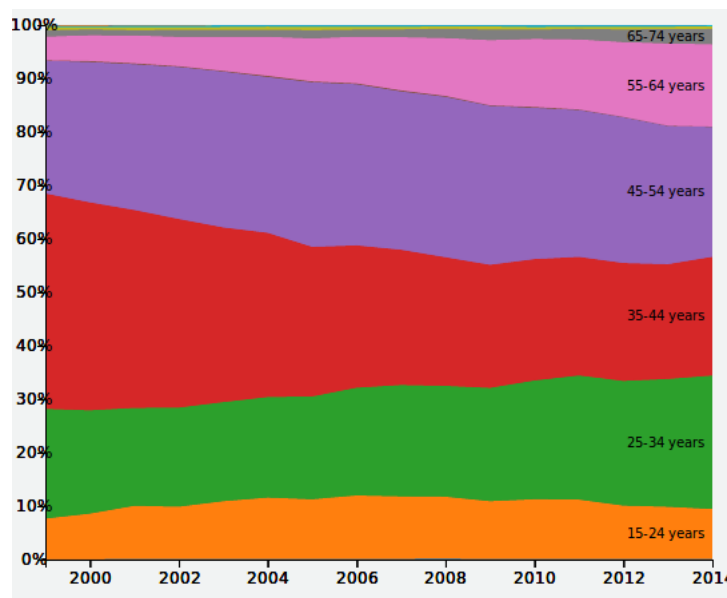


depicts the contribution in number of deaths along Y-axis by different elements like Cause, Age-Group, Gender and Race. One can easily notice that the use of Heroin increased after 2010.



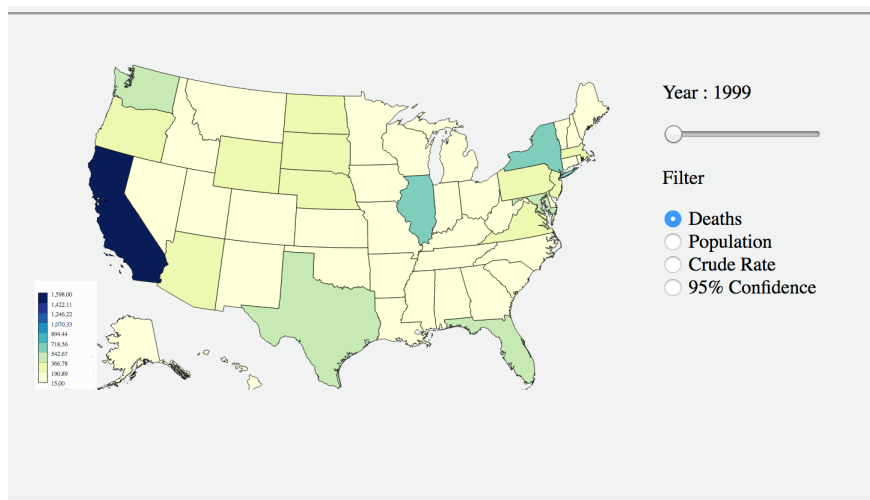
Adults aged 45–54 had the highest rate of drug overdose deaths in 2014. The rates of drug overdose deaths increased from 1999 to 2014 for all age groups. The greatest percentage increase occurred in the drug overdose death rate for adults aged 55–64, from 4.2 per 100,000 in 1999 to 21.8 in 2014, an average increase of 10.5% per year. In 2014, adults aged 45–54 had the highest death rate from drug overdose at 30 deaths per 100,000. In 2014, rates for adults aged 25–34, 35–44, 45–54, and 55–64 were more than twice the rate for younger adults aged 15–24, and more than 3.5 times the rate for adults aged 65 and over.

Clicking on the 'race' tab below the area-time graph, one can notice that 'White' race group has been affected the most by this epidemic.

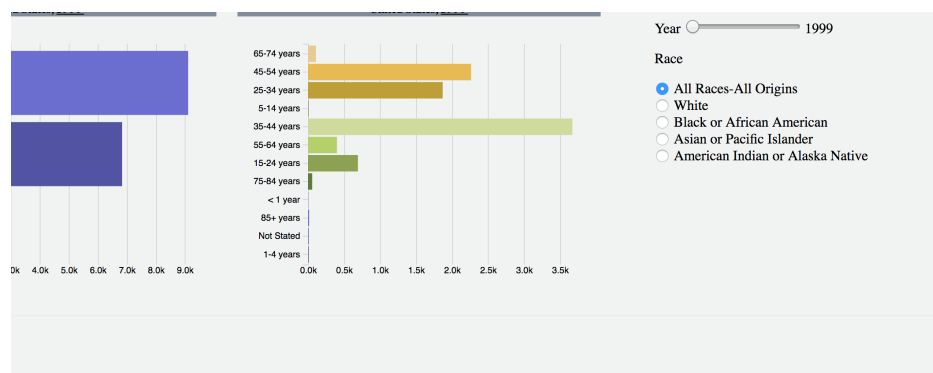


For the Geographical projection, we provided a US map that uses FIPS codes for different states. A time filter for different years is provided that helps seeing the trends in deaths, population, crude rate over different states. Below is the screen-shot. By paying attention to the crude rate as per different geographic regions, it is evident that the west region has constant high use of Heroin related drugs. However, in the states, West Virginia tops the list.

Now once we have seen the top-down trends, we can drill down and see details for different dimensions. A horizontal stacked bar chart, divided into portions,

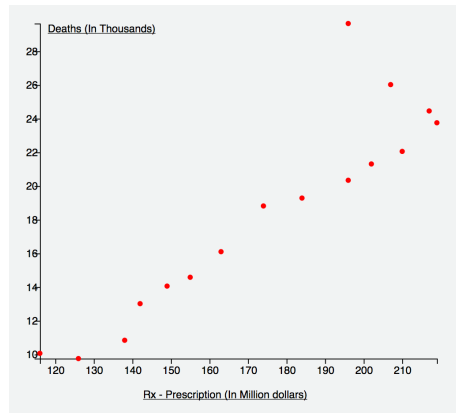


shows the number of deaths in different race groups, different sex over the span of 15 years. This visualization is interactive in nature. Due care has been taken while creating the dash-board and both the sides in this visualization are linked. By clicking on one, the changes are reflected in the other. Below is the screen-shot.

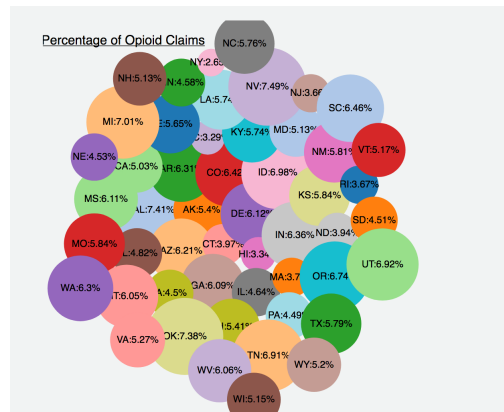


### 2.1.1 Correlation

We found a positive correlation between the number of prescription sold and deaths. It strengthens the argument that the people use drugs given by the doctors, first to ease the pain, but with time, they become addicted to the Opioid content present. Eventually, they get addicted to Heroin as the tolerance builds up. It is shown by a rather simple scatter plot shown below

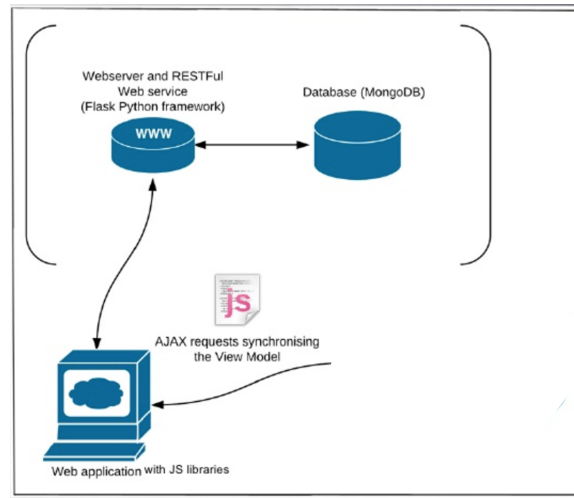


Finally, we also projected the number of Opioid based claims on the total number of claims filled in different states. We showed it using a bubble chart, where the radius explains the ratio proportionally. The bubble chart is shown below:



## 2.2 Framework Setup

The framework for the project as detailed in the primary proposal report uses Python-Flask as web server. For the database requirements, we set up a non-relational database, MongoDB. However, we are also using csv files in some part of the projects. To support the communication between Python code and MongoDB[4], we are using Pymongo. On the client side, we are using D3 for visualization. For asynchronous service calls from client to server (Flask) are made using AJAX[6] http requests with JSON responses to accommodate the REST requests(service calls).[8]



## 2.3 Data Cleaning

The originally gathered data [3] had many columns that contained null or invalid data. It was necessary to clean and fill the spaces with appropriate heuristics before loading the data in the mongoDB. For example, the value in the 'crude rate' column was filled by taking the average of 95% lower and upper crude rates.

## 2.4 Implementation

Our code implementation is in JavaScript, Python and JQuery. Most of the pre-processing part has been done in python. It includes data cleaning and restructuring such as a pivoting. We would like to mention that we used Geo-Map[7], a d3 extension library for geographical projections. Python-Flask framework was used to setup the endpoints for REST calls. This section includes the code snippets for different part of the projects.

Code Snippet for the database connection

```

4 from bson import json_util
5 from bson.json_util import dumps
6 import pandas as pd
7 import json
8 from pprint import pprint
9 from urlparse import urlparse
10
11 app = Flask(__name__)
12
13 MONGODB_HOST = 'localhost'
14 MONGODB_PORT = 27017
15 DBS_NAME = 'final'
16 COLLECTION_NAME = 'dataset'
17 FIELDS = {'_id': False}
18
19 """ Main method - CT """
20 @app.route("/")
21 def index():
22     cleanAndLoad()
23     return render_template("index_main.html")
24
25
26 @app.route("/dashboard")
27 def dashboard():
28     return render_template("dashboard1.html")
29
30 @app.route("/bubbleData")
31 def bubbleData():
32     client = MongoClient()
33     db = client.final
34     coll = db.geodata
35     data = coll.find( {}, { 'state_abr': 1, 'percent_opioid_claim': 1, '_id': 0 } )
36     json_data = []
37     for d in data:
38         json_data.append(d)
39     json_data = json.dumps(json_data, default=json_util.default)
40     client.close()
41     return json_data

```

Code Snippet for illustrating how to access MongoDB and use Python Flask

```

87 @app.route("/getGenderDataByAge")
88 def USSexDataByAge():
89     client = MongoClient()
90     db = client.final
91     coll = db.nation wide
92     race = urlparse(request.args.get('race', 'White', type=str)).path
93     year = request.args.get('year', type=int)
94     age = urlparse(request.args.get('age', 'All', type=str)).path
95     print year
96     if race == 'All Races-All Origins':
97         print "all"
98         pipe = [ {'$match': { '$and': [{'age_group': age, 'year': year}]} },
99                 {
100                     '$group':
101                     {
102                         '_id': '$gender',
103                         'totalAmount': { '$sum': '$deaths' }
104                     }
105                 },
106                 {'$project': {'label': '$_id', 'value': '$totalAmount', '_id': 0}}
107             ]
108     else:
109         print 'else'
110         pipe = [ {'$match': { '$and': [{'race': race, 'year': year, 'age_group': age}]} },
111                 {
112                     '$group':
113                     {
114                         '_id': {'gender': '$gender', 'race': '$race', 'year': '$year'},
115                         'totalAmount': { '$sum': '$deaths' }
116                     }
117                 },
118                 {'$project': {'label': '$_id.gender', 'race': '$_id.race', 'value': '$totalAmount', '_id': 0}}
119             ]
120
121 data = list(coll.aggregate(pipeline=pipe))
122 json_data = []
123 count = 0
124 for d in data:
125     count += d['value']
126     json_data.append(d)

```

Code Snippet for illustrating how to clean data using Python Pandas and dump json into MongoDB

```

296 """ Data Cleaning and Load in the mongoDB - AR """
297 @app.route("/cleanAndLoad")
298 def cleanAndLoad():
299     print("hello")
300     df = pd.read_csv('./static/data/dataset.csv')
301     for j in range(3,9):
302         f=pd.to_numeric(df[df.columns[j]],errors="coerce")
303         for i, row in df.iterrows():
304             oldValue = row[df.columns[j]]
305             if(oldValue=="Suppressed" or oldValue=="Unreliable"):
306                 df.set_value(i,df.columns[j],f.mean())
307
308
309 """Setting up the crude rate - AR"""
310 print(df.dtypes)
311 df[['Population', 'Crude Rate', 'Crude Rate Lower 95% Confidence Interval', 'Crude Rate Upper 95% C
312 df[['Deaths']] = df[['Deaths']].astype(int)
313 print(df.dtypes)
314 for i, row in df.iterrows():
315     lower = row[df.columns[6]]
316     upper = row[df.columns[7]]
317     df.set_value(i,df.columns[5],(lower+upper)/2)
318 data_json = json.loads(df.to_json(orient='records'))
319 """ Putting clean data in mongodb -AR"""
320 client = MongoClient()
321 coll = client.final.dataset
322 coll.drop()
323 coll.insert(data_json)
324 """ Just saving a copy of db collection to the file- AR"""
325 df.to_csv('./static/data/copyOfMainData.csv', columns=['Id', 'State', 'Year', 'Deaths', 'Population
326

```

Code Snippet for illustrating Geo Map and time slider

```

1 d3.select("#map").html("")
2 var globalYear = 1999
3 $('#sliderYear').on('input change', function () {
4     $('#rangeText').text($(this).val());
5 });
6
7 d3.selectAll('input[name="sliderYear"]').on("change",function(){globalYear=this.value;
8     $.getJSON($SCRIPT_ROOT + '/filterDataByYear', {
9         year: this.value
10     }, function(data) {
11         var filterType = d3.select('input[name="filter"]:checked')
12
13         var map = d3.geomap.choropleth()
14             .geofile('./static/d3-geomap/topojson/country
15             .colors(colorbrewer.YlGnBu[9])
16             .projection(d3.geo.albersUsa)
17             .column(filterType)
18             .unitId('Id')
19             .scale(600)
20             .legend(true);
21
22         d3.csv('./static/data/temp-map-data/temp'+globalYear+'.cs
23
24         d3.select('#map').html("")
25         .datum(data)
26         .call(map.draw, map);
27     });
28 });
29 return false;
30 });

```

Code Snippet for illustrating how to use JQuery AJAX calls

```

$.getJSON($SCRIPT_ROOT + '/getGenderData', {race : $('#input[name="filter"]:checked').val(),year:$('input
[name="sliderYear"]').val()), function(d) { horbar1(d,"#barGender");});
$.getJSON($SCRIPT_ROOT + '/getAgeGroupData', {race : $('#input[name="filter"]:checked').val(),year:$('
input[name="sliderYear"]').val()), function(d) { horbar1(d,"#barAge");});
d3.selectAll('input[name="filter"]').on("change", change);
$('#sliderYear').on('input change', function () {
    $('#rangeText').text($(this).val());
});

```

Code Snippet for illustrating how to link two visualizations



```

90 function change() {
91     d3.selectAll("#agespan").html('All Ages');
92     $.getJSON($SCRIPT_ROOT + '/getAgeGroupData', {race : $('input[name="filter"]:checked').val(), year: $('
93     input[name="sliderYear"]').val(), function(d) { horbar1(d, "#barAge");});
94     $.getJSON($SCRIPT_ROOT + '/getGenderData', {race : $('input[name="filter"]:checked').val(), year: $('
95     input[name="sliderYear"]').val(), function(d) { horbar1(d, "#barGender");});
96 }
97
98 function changeGenderByAge(age) {
99     $.getJSON($SCRIPT_ROOT + '/getGenderDataByAge', {race : $('input[name="filter"]:checked').val(), year:
100     $('input[name="sliderYear"]').val(), age: age, function(d) { horbar1(d, "#barGender");});
101 }
102
103 function changeAgeByGender(gender) {
104     $.getJSON($SCRIPT_ROOT + '/getAgeGroupDataByGender', {race : $('input[name="filter"]:checked').val(),
105     year: $('input[name="sliderYear"]').val(), gender: gender, function(d) { horbar1(d, "#barAge");});
106 }

```

### 3 References

- 1 [<https://www.cdc.gov/mmwr/preview/mmwrhtml/mm6450a3.htm>]
- 2 [<https://en.wikipedia.org/wiki/Opioid>]
- 3 [<https://data.world/health/opioid-overdose-deaths>]
- 4 [<https://en.wikipedia.org/wiki/MongoDB>]
- 5 [[https://en.wikipedia.org/wiki/Flask\\_\(web\\_framework\)](https://en.wikipedia.org/wiki/Flask_(web_framework))]
- 6 [<https://cdnjs.cloudflare.com/ajax/libs/jquery/3.0.0-beta1/jquery.min.js>]
- 7 [<https://d3js.org/>]
- 8 [<https://image.slidesharecdn.com/finalpresentation-120501061120-phpapp01/95/listentoitlater-final-year-project-presentation-13-728.jpg?cb=1365141016>]