

Term Project Step 2

Burden Joshua

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Legalization of Marijuana

Introduction

With more and more states adding medical and recreational cannabis to their ballots, questions are now being asked more openly about the impact of cannabis on public health, crime statistics, and popularity of legalization. The data out there is sparse, but the data that has been collected could show trends and insights on the future of legislation and public sentiment of legalization, along with showing the potential for lowered crime and raised tax revenue in states and at the federal level.

Research questions

I want to know, based on the data that is out there, if legalization of marijuana:

- Reduces the number of arrests of all nonviolent offenses significantly
- Would increase the tax income for states and government entities
- Would drug usage go down by making it legal
- What is the state tax collections for legal rec states vs med only vs no rec/med
- What is the public support for legalization

Approach

I will look at the data that is out there for sales, crime, and public sentiment and see if there is anything that stands out to answer any of the questions above.

How your approach addresses (fully or partially) the problem.

The data will tell me whether or not the any or all of my questions can be answered and if there is discrepancies in the data, more research might be conducted to see if there is correlation.

Data (Minimum of 3 Datasets - but no requirement on number of fields or rows)

- <https://data.world/denver/marijuana-related-crime>
- <https://www.kaggle.com/datasets/tunguz/drug-use-by-age>
- <https://www.kaggle.com/datasets/mykeysw/marijuana-sales-forecasting-in-tx>
- <https://data.world/sya/marijuana-laws-by-state>
- <https://data.world/denver/marijuana-gross-sales>
- <https://data.world/health/support-for-legal-marijuana>
- <https://www.kaggle.com/datasets/terenceshin/historical-prices-for-biggest-weed-stocks>

Bad dataset

- <https://data.world/opensavannah/cannabis-justice>

Required Packages

- Gplot2

Plots and Table Needs

- Histograms
- Scatter plots
- CDF
- Linear Regressions

Questions for future steps

Is there more data out there that would help with this research, and how deep should these models go in terms of covering the questions?

Step 2

How to import and clean my data

Google is very good actually for making suggestions on cleaning datasets. That is my first step, by bringing these datasets into google sheets, I am effectively having google help me get the datasets clean. One other thing that I needed to do was to also parse what is important and what isn't based on the datasets. There were quite a number of datasets that were not helpful or would be considered a large reach for the dataset to fit into the story, so I removed those datasets from the list of datasets I had. I also made the decision early on to remove header column names that either didn't make sense or were not required for the dataset story.

What does the final data set look like?

Summaries of the dataset, as to not get overwhelmed by the vast amounts of data.

```
setwd('/Users/joshua/Documents/PERSONAL_GITHUB_REPOS/dsc520/TermProject')
cosalesrev <- read.csv("data/cleaned_COMJSalesMonthReports2019.csv")
cotaxrev <- read.csv("data/cleaned_COMJTaxMonthReports2019.csv")
costatepop <- read.csv("data/cleaned_COCountyPop20142018.csv")
drug_use <- read.csv("data/cleaned_drug-use-by-age.csv")
crime_denver <- read.csv("data/cleaned_crime_marijuana.csv")
legal_support <- read.csv("data/cleaned_legal_marijuana_support.csv")
```

Summaries of datasets

```
# summaries of the datasets
summary(cosalesrev)
```

```
##      Month      MedicalSales      MSYearToDate      RetailSales
## Length:65      Min.      :24082927      Min.      : 25680596      Min.      : 14022213
## Class :character 1st Qu.:29238678      1st Qu.:100238704      1st Qu.: 44590853
## Mode  :character Median :32686869      Median :193094556      Median : 76018423
##              Mean  :32736868      Mean  :204611449      Mean   : 70423116
##              3rd Qu.:36010893      3rd Qu.:302926736      3rd Qu.: 96642441
##              Max.   :41056948      Max.   :445616062      Max.   :114317739
##      RSYearToDate      TotalSales      TSYearToDate
## Min.      :1.402e+07      Min.      : 45987045      Min.      :4.656e+07
## 1st Qu.:1.599e+08      1st Qu.: 78199969      1st Qu.:2.831e+08
```

```
## Median :3.088e+08 Median :109222331 Median :5.399e+08
## Mean :4.086e+08 Mean :103159984 Mean :6.132e+08
## 3rd Qu.:5.775e+08 3rd Qu.:127706166 3rd Qu.:8.809e+08
## Max. :1.214e+09 Max. :143107279 Max. :1.546e+09
## TotalToDate
## Min. :4.656e+07
## 1st Qu.:1.047e+09
## Median :2.644e+09
## Mean :2.895e+09
## 3rd Qu.:4.612e+09
## Max. :6.705e+09
```

```
summary(cotaxrev)
```

```
## Month SalesTax LicenseFees SalesTaxFees
## Length:65 Min. : 628947 Min. : 592661 Min. :1570401
## Class :character 1st Qu.:1014752 1st Qu.: 972122 1st Qu.:2137462
## Mode :character Median :1808419 Median :1063563 Median :3067394
## Mean :1970529 Mean :1097654 Mean :3068184
## 3rd Qu.:2763721 3rd Qu.:1221521 3rd Qu.:4016545
## Max. :3692930 Max. :1663120 Max. :4892115
## RetailSalesTax RetailExciseTax TotalTaxFees YearToDate
## Min. : 1401568 Min. : 195318 Min. : 3519756 Min. : 3519756
## 1st Qu.: 4394550 1st Qu.:2796865 1st Qu.:10856584 1st Qu.: 37511919
## Median : 7746575 Median :4683825 Median :17694953 Median : 76306924
## Mean : 8877802 Mean :4114956 Mean :16060942 Mean : 93255239
## 3rd Qu.:14608085 3rd Qu.:5598581 3rd Qu.:21622509 3rd Qu.:134971077
## Max. :18698640 Max. :7867853 Max. :26841073 Max. :266529637
## TotalToDate
## Min. :3.520e+06
## 1st Qu.:1.260e+08
## Median :3.555e+08
## Mean :4.161e+08
## 3rd Qu.:6.818e+08
## Max. :1.044e+09
```

```
summary(costatepop)
```

```
## County X2014Pop X2015Pop X2016Pop
## Length:64 Min. : 703 Min. : 689 Min. : 690
## Class :character 1st Qu.: 5692 1st Qu.: 5736 1st Qu.: 5636
## Mode :character Median : 14288 Median : 14358 Median : 14592
## Mean : 83526 Mean : 85076 Mean : 86472
## 3rd Qu.: 41946 3rd Qu.: 41975 3rd Qu.: 42592
## Max. :664715 Max. :683081 Max. :696347
## X2017Pop X2018Pop
## Min. : 714 Min. : 762
## 1st Qu.: 5837 1st Qu.: 5876
## Median : 14747 Median : 15014
## Mean : 87648 Mean : 88993
## 3rd Qu.: 43202 3rd Qu.: 43666
## Max. :705651 Max. :716492
```

```
summary(drug_use)
```

```
## age sample_size marijuana_use_by_percentage
```

```
## Length:17      Min.    :2223  Min.    : 1.10
## Class :character 1st Qu.:2469  1st Qu.: 8.70
## Mode  :character Median :2798  Median :20.80
##              Mean  :3251  Mean  :18.92
##              3rd Qu.:3058  3rd Qu.:28.40
##              Max.   :7391  Max.   :34.00
## marijuana_frequency_over_12_months
## Min.    : 4.00
## 1st Qu.:30.00
## Median :52.00
## Mean    :42.94
## 3rd Qu.:52.00
## Max.    :72.00
```

```
summary(crime_denver)
```

```
## REPORTDATE      OFFENSE_CODE  OFFENSE_TYPE_ID  OFFENSE_CATEGORY_ID
## Length:1254     Min.    :1006  Length:1254     Length:1254
## Class :character 1st Qu.:2203  Class :character Class :character
## Mode  :character Median :2203  Mode  :character Mode  :character
##              Mean  :2249
##              3rd Qu.:2206
##              Max.   :7399
## NEIGHBORHOOD_ID
## Length:1254
## Class :character
## Mode  :character
##
##
##
```

```
summary(legal_support)
```

```
##      Year      Month      Asked_half_sample  Yes_Legal
## Min.    :1969  Length:20      Length:20      Min.    :12.00
## 1st Qu.:1980  Class :character  Class :character  1st Qu.:25.00
## Median :2002  Mode  :character  Mode  :character  Median :34.00
## Mean    :1997                                Mean    :35.95
## 3rd Qu.:2011                                3rd Qu.:48.50
## Max.    :2016                                Max.    :60.00
## No_Illegal    No_Opinion  Percent_Yes      Percent_No
## Min.    :39.00  Min.    :1.00  Length:20      Length:20
## 1st Qu.:49.25  1st Qu.:2.00  Class :character  Class :character
## Median :63.00  Median :4.00  Mode  :character  Mode  :character
## Mean    :60.50  Mean    :3.45
## 3rd Qu.:70.75  3rd Qu.:4.25
## Max.    :84.00  Max.    :6.00
## Percent_No_Opinion
## Length:20
## Class :character
## Mode  :character
##
##
##
```

glimpse of datasets

```
glimpse(cosalesrev)
```

```
## Rows: 65
## Columns: 8
## $ Month      <chr> "Jan 2014", "Feb 2014", "Mar 2014", "Apr 2014", "May 2014~
## $ MedicalSales <int> 32541720, 31738572, 34821878, 32686869, 31355208, 2995030~
## $ MSYearToDate <int> 32541720, 64280292, 99102170, 131789039, 163144247, 19309~
## $ RetailSales <int> 14022213, 14248473, 19881631, 20765986, 21375001, 2397808~
## $ RSTYearToDate <int> 14022213, 28270686, 48152317, 68918303, 90293304, 1142713~
## $ TotalSales <int> 46563933, 45987045, 54703509, 53452855, 52730209, 5392839~
## $ TSYearToDate <int> 46563933, 92550978, 147254487, 200707342, 253437551, 3073~
## $ TotalToDate <dbl> 46563933, 92550978, 147254487, 200707342, 253437551, 3073~
```

```
glimpse(cotaxrev)
```

```
## Rows: 65
## Columns: 9
## $ Month      <chr> "Feb 2014", "Mar 2014", "Apr 2014", "May 2014", "Jun 2~
## $ SalesTax    <int> 1330209, 1460429, 1569405, 1559710, 1569454, 1530968, ~
## $ LicenseFees <int> 592661, 857615, 902995, 761687, 940028, 1547853, 13795~
## $ SalesTaxFees <int> 1922870, 2318044, 2472400, 2321397, 2509482, 3078821, ~
## $ RetailSalesTax <int> 1401568, 1434916, 1898685, 2217607, 2070577, 2473627, ~
## $ RetailExciseTax <int> 195318, 339615, 609907, 734351, 1135648, 969637, 13979~
## $ TotalTaxFees <int> 3519756, 4092575, 4980992, 5273355, 5715707, 6522085, ~
## $ YearToDate  <int> 3519756, 7612330, 12593322, 17866677, 23582384, 301044~
## $ TotalToDate <int> 3519756, 7612330, 12593322, 17866677, 23582384, 301044~
```

```
glimpse(costatepop)
```

```
## Rows: 64
## Columns: 6
## $ County      <chr> "Adams", "Alamosa", "Arapahoe", "Archuleta", "Baca", "Bent", ~
## $ X2014Pop <int> 479477, 15758, 617498, 12240, 3576, 5777, 312588, 61617, 1845~
## $ X2015Pop <int> 489774, 15854, 628951, 12401, 3544, 5885, 318071, 64713, 1857~
## $ X2016Pop <int> 497419, 16006, 637266, 12839, 3522, 5664, 321363, 66399, 1907~
## $ X2017Pop <int> 503375, 16056, 643257, 13316, 3539, 5866, 322854, 68169, 1962~
## $ X2018Pop <int> 511868, 16683, 651215, 13765, 3585, 5882, 326078, 69267, 2002~
```

```
glimpse(drug_use)
```

```
## Rows: 17
## Columns: 4
## $ age          <chr> "12", "13", "14", "15", "16", "17", ~
## $ sample_size  <int> 2798, 2757, 2792, 2956, 3058, 3038, ~
## $ marijuana_use_by_percentage <dbl> 1.1, 3.4, 8.7, 14.5, 22.5, 28.0, 33~
## $ marijuana_frequency_over_12_months <int> 4, 15, 24, 25, 30, 36, 52, 60, 60, ~
```

```
glimpse(crime_denver)
```

```
## Rows: 1,254
## Columns: 5
## $ REPORTDATE    <chr> "2/27/2012", "8/6/2012", "9/18/2012", "8/19/2012", ~
## $ OFFENSE_CODE  <int> 2203, 2203, 2203, 5707, 2203, 2203, 2203, 2203, 22~
## $ OFFENSE_TYPE_ID <chr> "BURGLARY - BUSINESS BY FORCE", "BURGLARY - BUSINE~
## $ OFFENSE_CATEGORY_ID <chr> "Burglary", "Burglary", "Burglary", "All Other Cri~
```

```
## $ NEIGHBORHOOD_ID      <chr> "montclair", "five-points", "hampden-south", "mont-
glimpse(legal_support)

## Rows: 20
## Columns: 9
## $ Year                 <int> 2016, 2015, 2014, 2013, 2012, 2011, 2010, 2009, 200~
## $ Month                <chr> "Oct", "Oct", "Oct", "Oct", "Nov", "Oct", "Oct", "O~
## $ Asked_half_sample    <chr> "no", "no", "no", "no", "no", "no", "yes", "yes", "~
## $ Yes_Legal            <int> 60, 58, 51, 58, 48, 50, 46, 44, 36, 34, 34, 31, 25,~
## $ No_Illegal           <int> 39, 40, 47, 39, 50, 46, 50, 54, 60, 64, 62, 64, 73,~
## $ No_Opinion           <int> 1, 2, 2, 3, 1, 3, 4, 2, 4, 2, 4, 5, 2, 4, 5, 5, 6, ~
## $ Percent_Yes          <chr> "60%", "58%", "51%", "58%", "48%", "51%", "46%", "4~
## $ Percent_No           <chr> "39%", "40%", "47%", "39%", "51%", "46%", "50%", "5~
## $ Percent_No_Opinion   <chr> "1%", "2%", "2%", "3%", "1%", "3%", "4%", "2%", "4~
```

Questions for future steps.

I am in a situation where there is almost too much data, but not enough connections as there

What information is not self-evident?

I was going to add this datasets to other states, but for many states there is not an aggregate set of information out there for just marijuana related offenses, and without being specific, it becomes difficult to see patterns and trends when the net is cast to wide.

What are different ways you could look at this data?

Data like this is not very connected. Making a story out of the datasets requires looking at each dataset as a piece of a puzzle and not trying to force the datasets to work with each other but rather, answer a question and check if the answer relates to the next question.

How do you plan to slice and dice the data?

For one set of data, I plan to see the sales trend for Colorado based on sales, and tax data. I will also look at if crime went up specifically in Denver, and then see if there is another data set out there to see if there is a relationship with higher sales, with uprising crime and youth uses.

What types of plots and tables will help you to illustrate the findings to your questions?

barcharts and lineplots seem to make the most sense in these instances. I might also look at doing some data plotting with a scatterplot.

Do you plan on incorporating any machine learning techniques to answer your research questions? Explain.

I am unsure at this time I will add a machine learning algorithm to these datasets. I would like to say yes, but I am still parsing data, and narrowing my search queries to something smaller than my original scope.

Questions for future steps.

Is there more data out there that I just have not found yet that might give me a better and more up to date dataset for the questions I am posing?

Step 3

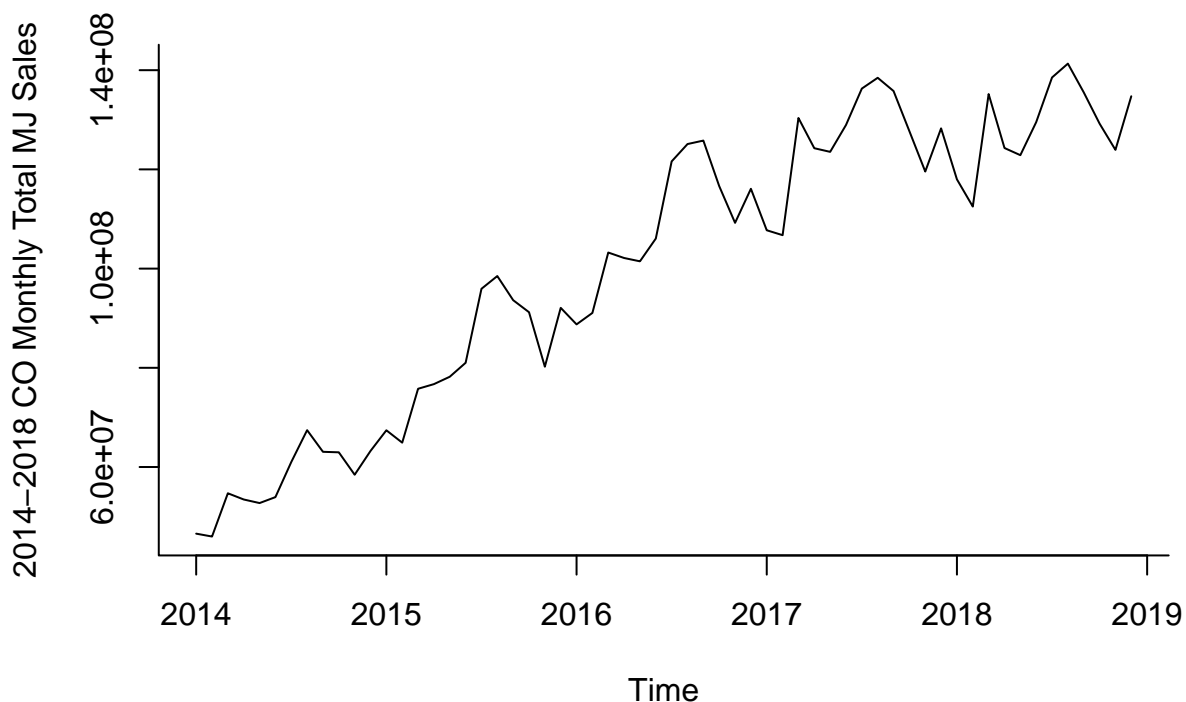
Introduction

The data collected is attempting to find out if the sale and legality of cannabis shows an increase in crime and use in teens. The data collected shows the sales of cannabis based on annual sales since its legalization recreationally in 2014 in the state of Colorado, and the statistics of crime in the city of Denver. Data that did not support the investigation was scrubbed from the datasets and a new aggregate dataset was formed using the data that best fits the model example we are trying to follow. Results are interesting after running the models and building out visualizations.

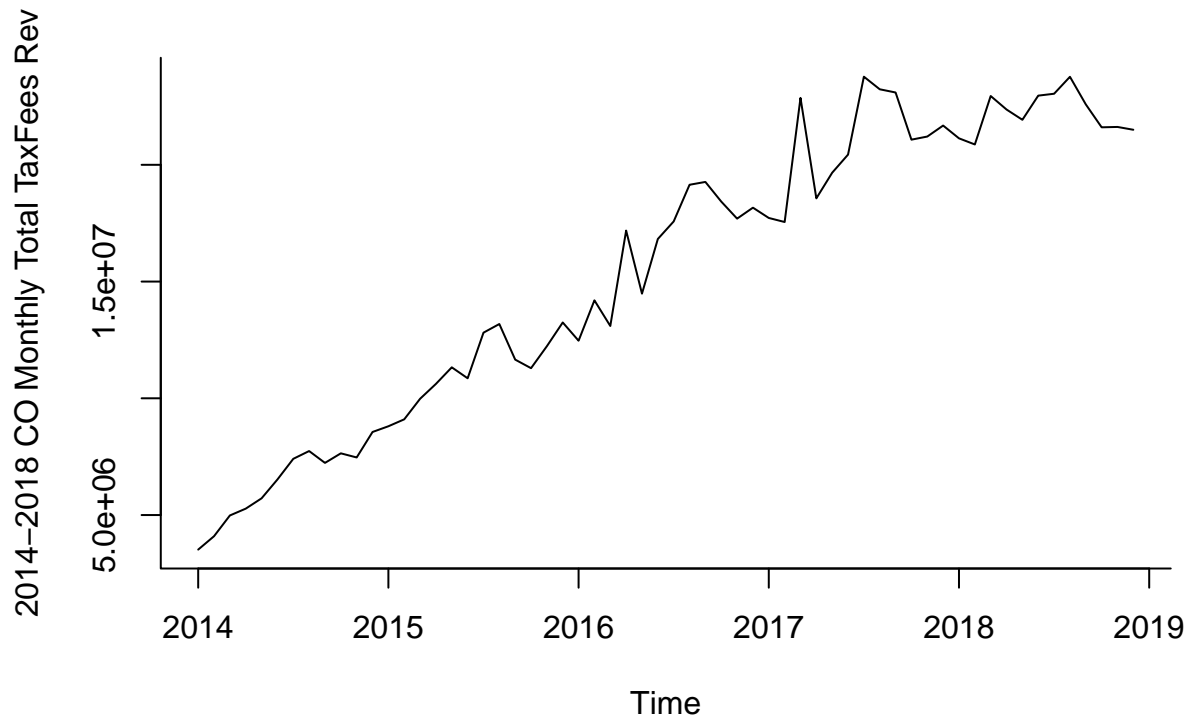
Findings

found this:

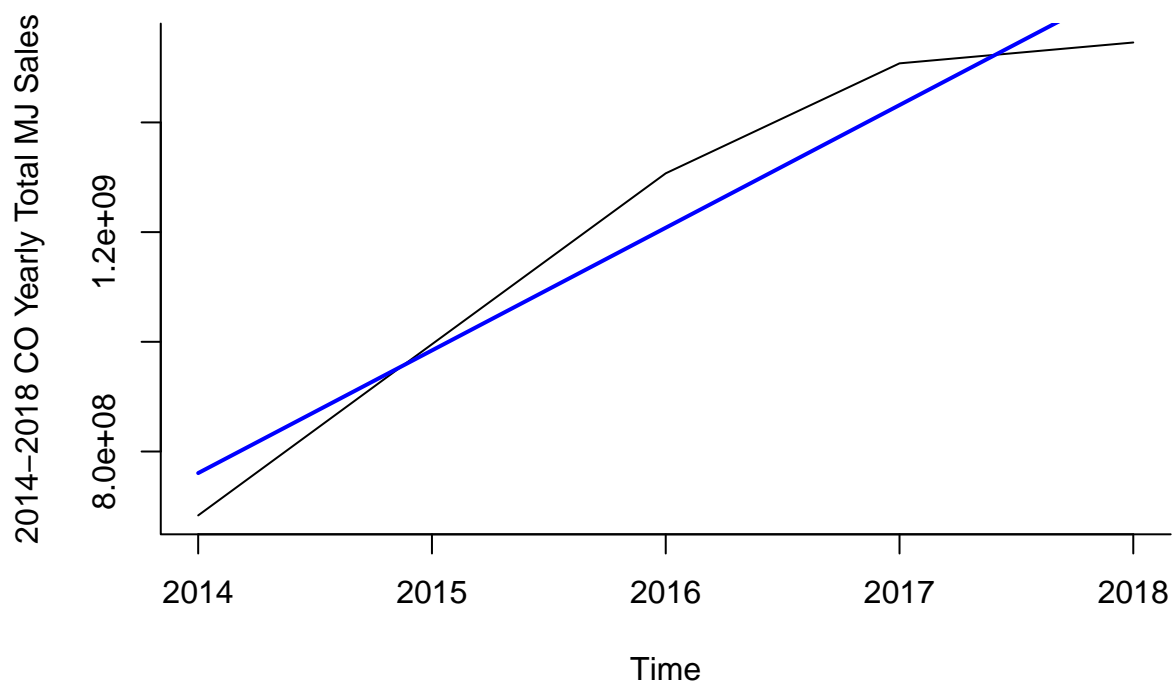
```
# Exploratory Analysis CO 2014-2019 marijuana sales & tax revenue
cosalesrev <- read.csv("data/cleaned_COMJSalesMonthReports2019.csv")
cotaxrev <- read.csv("data/cleaned_COMJTaxMonthReports2019.csv")
# Visualize CO 2014-2019 monthly total marijuana sales as a time series
cosalesrev.ts <- ts(cosalesrev$TotalSales, start = c(2014,1), end = c(2018,12), freq = 12)
plot(cosalesrev.ts, xlab = "Time", ylab = "2014-2018 CO Monthly Total MJ Sales", bty = "l")
```



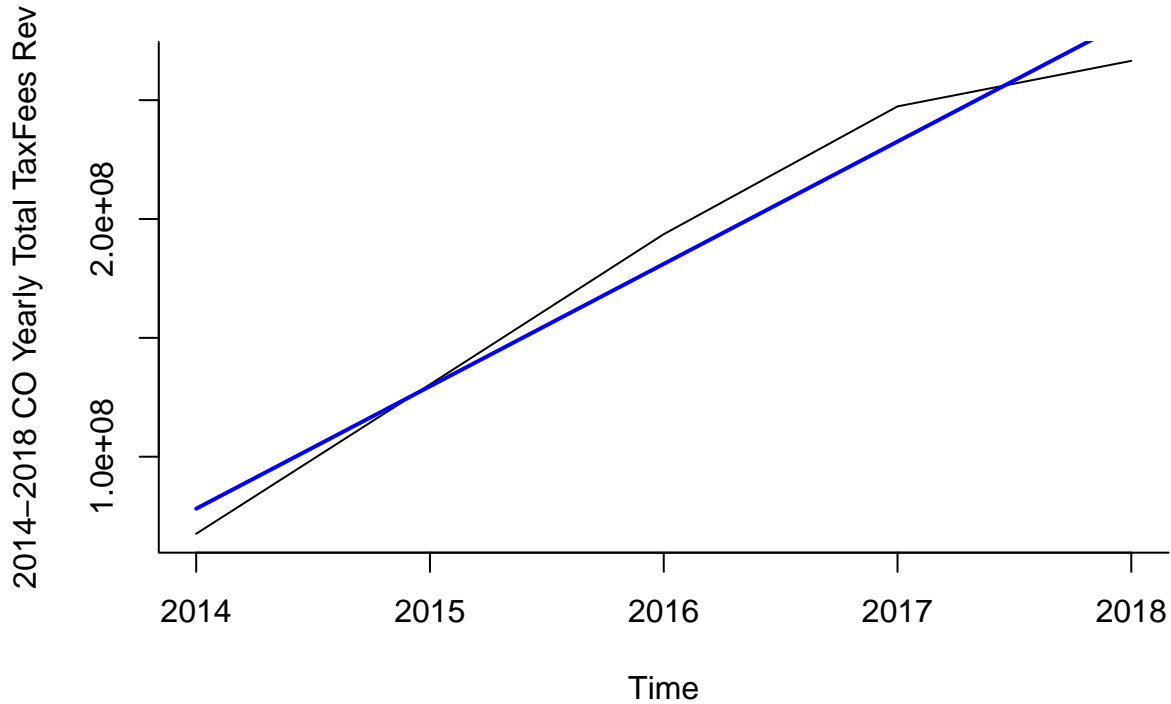
```
# Visualize CO 2014-2019 monthly total tax fees revenue as a time series
cotaxrev.ts <- ts(cotaxrev$TotalTaxFees, start = c(2014,1), end = c(2018,12), freq = 12)
plot(cotaxrev.ts, xlab = "Time", ylab = "2014-2018 CO Monthly Total TaxFees Rev", bty = "l")
```



```
# Exploratory Analysis CO 2014-2018 marijuana sales & tax summaries
cosalessummary <- read.csv("data/cleaned_COMJSalesSummary2019.csv")
cotaxsummary <- read.csv("data/cleaned_COMJTaxSummary2019.csv")
# Visualize CO 2014-2018 total marijuana sales summary by years as time series
cosalessummary.ts <- ts(cosalessummary$TotalSales, start = c(2014), end = c(2018), freq = 1)
cosalessummary.lm <- tslm(cosalessummary.ts ~ trend)
plot(cosalessummary.ts, xlab = "Time", ylab = "2014-2018 CO Yearly Total MJ Sales", bty = "n")
lines(cosalessummary.lm$fitted, col="blue", lwd = 2)
```




```
# Visualize CO 2014-2019 monthly tax summary by years as a time series
cotaxsummary.ts <- ts(cotaxsummary$TotalTaxFees, start = c(2014), end = c(2018), freq = 1)
cotaxsummary.lm <- tslm(cotaxsummary.ts ~ trend)
plot(cotaxsummary.ts, xlab = "Time", ylab = "2014-2018 CO Yearly Total TaxFees Rev", bty = "l")
lines(cotaxsummary.lm$fitted, col="blue", lwd = 2)
```



The problem statement you addressed.

Addressing the data

Analysis

Implications

Limitations

Concluding Remarks