

# Research of COVID-19 impact and trend in USA using statistical approach.

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**Abstract**—This paper presents a research of a public dataset United States COVID-19 Cases and Deaths by State over Time from Centers for Disease Control and Prevention containing 1 721 928 instances with goals to forecast a trend and to explore an impact on population of USA.

To determine research results statistical approach was applied which consists of methods such as attribute selection, aggregation, pattern recognition, feature classification, multiple regression models, k-means clustering and visualization.

To improve and to cross-validate accuracy of research results linear and isotonic regression models were used.

To determine deviations of inflated states data have been grouped and classified by mean feature.

Goals of visualization part of research are to identify US states based on density rates and to build cluster using k-means algorithm.

## I. INTRODUCTION

COVID-19 caused global pandemic mode to be enabled in 2019 [cov20] [arx].

The first case was identified in Wuhan, China, in December 2019.

SARS-CoV-2 is a cause which is defined as severe acute respiratory syndrome coronavirus 2.

COVID-19 is an effect of coronavirus disease 2019.

Previously it was discovered 4 pandemics for the last century: in 1918, 1957, 1968 and 2009 [cdc].

The research is dedicated to the fifth pandemic mode of century - COVID-19.

Technical part of research consists of applying following computer science methods over the United States COVID-19 Cases and Deaths by State over Time dataset.

- attribute selection
- aggregation
- pattern recognition
- mean feature classification
- linear and isotonic regression models
- density map plotting
- k-means clustering

## II. NORMALIZATION

Attribute Selection helps to clean data.

Dataset consists of 15 unique properties.

After applying method of attribute selection only 3 properties left.

Total number of confirmed instances with submission date allows to gain more accurate numbers for predictor model.

Other properties have missed or inconsistent values.

Property	Description	Type
submission_date	Date of counts	Date & Time
conf_cases	Total confirmed cases	Number
conf_death	Total number of confirmed deaths	Number

Table I: Selected dataset attributes [cdc20]

The dataset has 1 721 928 number of instances to work with.

In our research actual range was used - 04/15/20 - 12/27/20 with 0 as starting point at 04/01/20.

Total discovered points to works with is 55.

Selected attributes were grouped by 1, 5, 10, 15, 20, 25 days of each month with have medium values.

Numbers have cumulative feature which is shown in their increasing exponential progression III.

Recovery formula 1 has been introduced to increase accuracy of research and observe deviations.

$$(1) \quad Recovery = Case - Death$$

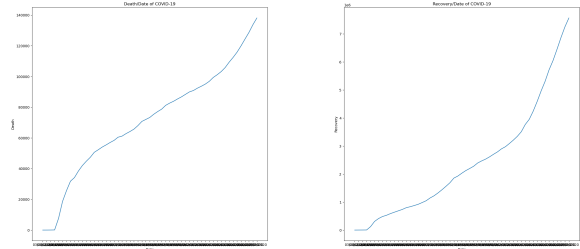


Figure 1: Raw Death/Recovery Rates

The slope of first image has sharp slope in the beginning and linear slope form in the tail.

Meanwhile beginning of another image has linear line slope and the tail has sharp curve form.

Both functions growth bounded above by  $n \log n$  and below by  $\sqrt{n}$  asymptotically.

$$(2) \quad f(x) \in O(n \log n)$$

$$(3) \quad f(x) \in \Omega(\sqrt{n})$$

To get rid off increasing progression that the slope is too steep the shatter semi-curve segmentation algorithm was developed 1 1.

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**Algorithm 1** Shatter semi-curve segmentation algorithm
 

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```

function SHATTERSEGMENTATION(DictionarySet  $F$ )
   $Result \leftarrow F$  ▷ Copy Data
  for ( $i = 0$ ;  $i < 1\,721\,928$ ;  $i++$ ) do
     $A \leftarrow F_i$  ▷ Current Day
     $B \leftarrow F_{i+1}$  ▷ Next Day
     $Result_i \leftarrow A - B$  ▷ Difference
  end for
  return  $Result$ 
end function

```

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Applied segmented algorithm results 2.

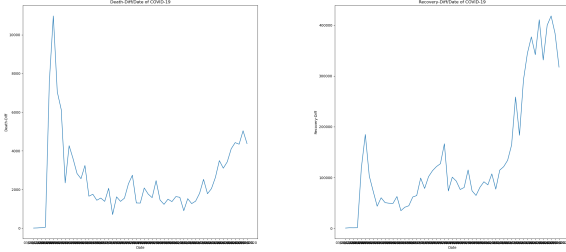


Figure 2: Segmented Death/Recovery Rates

The first image has strong noticeable wave in the beginning of the line.

The following distribution of the first image is almost normal except the rising wave in the tail.

The second image has 3 noticeable waves.

First and second waves are almost similar except the biggest wave size in the tail.

Alternatively, square-normalization approach was applied to reduce overall value of Y-axis and to preserve curve angles as a vector ratio scale 3.

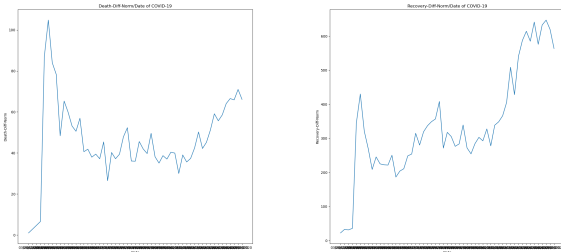


Figure 3: Square-normalization Death/Recovery Rates

Results of the applied shatter segmentation 1 and square-normalization algorithms.

Waves became less perceptible and more smoothly because of angles acute forms.

Furthermore, Segmentation algorithm 1 was applied to population dataset 4.

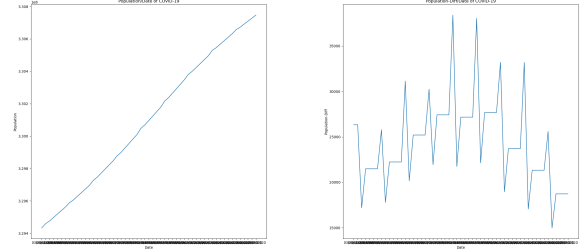


Figure 4: Raw and Segmented population of USA 2020

The first function bounded above by  $n \log n$  2 and below by  $\sqrt{n}$  3 asymptotically.

The second image has noticeable wave in the middle of image.

### III. RESULTS

#### A. COVID-19 impact on population of USA

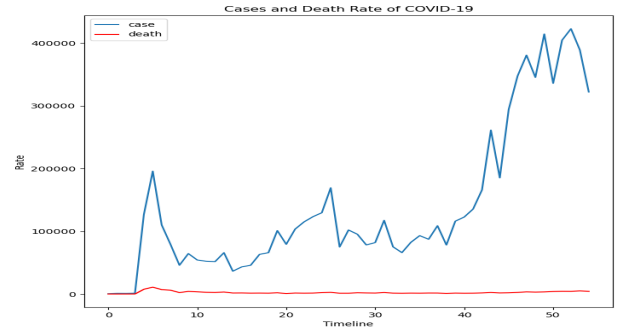


Figure 5: Case and Death Rate

Death rate curve has only one small wave in the beginning of the line.

Case rate has 3 waves with the strongest wave in the tail of the curve.

The death rate function has linear form with constant 4 asymptotic.

$$(4) \quad f(x) \in O(1)$$

Only small impact on population is noticeable in the beginning of the timeline 5.

The death rate function 6 has 1 wave in the beginning of the timeline.

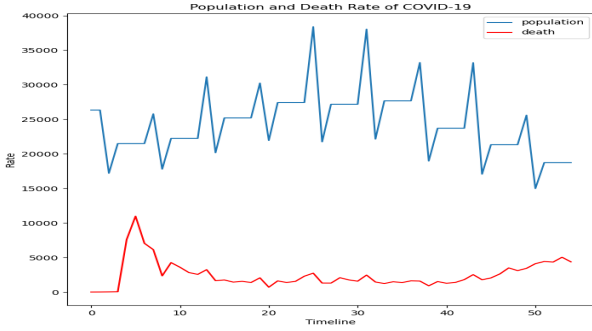


Figure 6: Population and Death Rate Correlation

Same as on the previous figure, the death rate function 6 has linear form with constant 4 asymptotic.

The population rate has 1 wave in the middle of the timeline.

It can be assumed because of that the coronavirus death rate has no effect on population of the USA, otherwise the death rate suppose to have a wave in the middle.

### B. Regressions Tendency

Multiple regression models were used in research to cross-validate estimated forecast results 7.

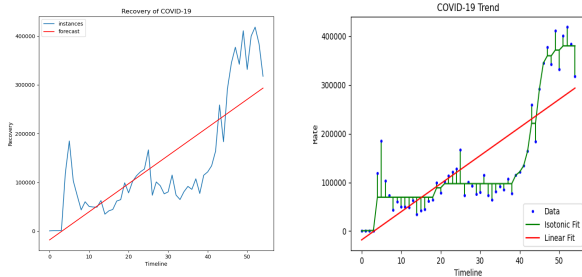


Figure 7: Recovery Rate Linear and Isotonic Regressions

Both models have similar results with small deviation.

Models show that recovery trend is growing up.

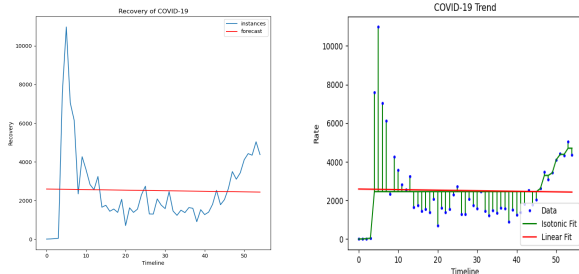


Figure 8: Death Rate Linear and Isotonic Regressions

Both functions strives for negative value and trend is death negative 8.

It can be concluded that trend is positive based on that

- death rate is going down 8
- recovery rate is going up 7
- population rate is going up 6

### C. Data Classification

Data II was grouped by state with total count of 26 states V.

Each state was classified by mean feature [wik20] for death and recovery attributes as threshold to identify states with inflated rates.

Death feature classification results: 17 state are mean and 9 are overmean V.

Recovery feature classification results: 15 state are mean and 11 are overmean V.

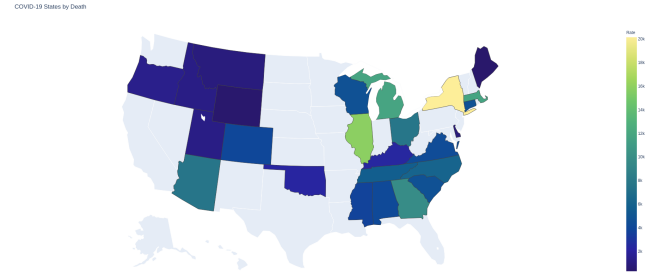


Figure 9: States Map By Death Attribute

New York state which is located in the north-east of the map has the highest death rate density between among states 9.

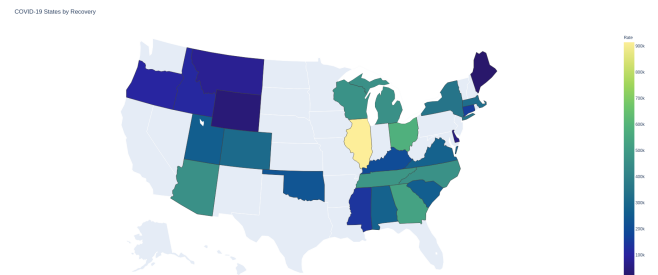


Figure 10: States Map By Recovery Attribute

Illinois state which is located in the middle of the map has the highest recovery rate 10.

### D. Clustering

In total 2 sub-clusters have been generated with 0 and 1 as cluster center points based on k-means algorithms using unsupervised learning 11.

The k-mean algorithm experienced error when more than 5 clusters were calculated.

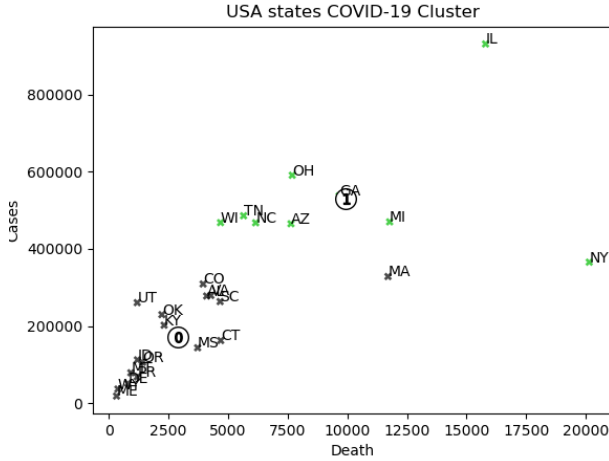


Figure 11: Cluster

#### IV. CONCLUSION

Number of positive cases compared to death rate has positive impact on recovery tendency and do not affect overall population tendency of USA.

However, COVID-19 is dangerous disease could affect every group of people - the president of USA was diagnosed positive with COVID-19 approximately 10/01/2020 [Tru20].

“Anyone who was a direct contact of President Trump or known COVID-19 cases needs to quarantine and should get tested [mn-]”.

“More than a dozen people in President Donald Trump’s circle – including Trump himself – have recently tested positive for Covid-19 [cnn]”.

The research does not exclude a small possibility of a second wave of coronavirus under new abbreviation of COVID-20 in 2021.

Besides population impact COVID-19 has impact on US businesses which could be used as a theme for future research.

Many companies claim Chapter 11 due to COVID-19 impact on business. “For some, the pandemic was a chance to open a new chapter. But for many businesses, the swift and stark economic shutdown led straight to Chapter 11. [for]” ■.

#### REFERENCES

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#### APPENDIX

##### OBTAINED RESULTS DATA

The gained data from experiment.

Property	Description	Type
submission_date	Date of counts	Date & Time
state	Jurisdiction	Plain Text
tot_cases	Total number of cases	Number
conf_cases	Total confirmed cases	Number
prob_cases	Total probable cases	Number
new_case	Number of new cases	Number
pnew_case	Number of new probable cases	Number
tot_death	Total number of deaths	Number
conf_death	Total number of confirmed deaths	Number
prob_death	Total number of probable deaths	Number
new_death	Number of new deaths	Number
pnew_death	Number of new probable deaths	Number
created_at	Date and time record was created	Date & Time
consent_cases	Composite field*	Plain Text
consent_deaths	Composite field*	Plain Text

Table II: Columns in the dataset [cdc20]

Date	Case	Death	Recovery
03/25/2020	504	1	503
04/01/2020	1584	9	1575
04/05/2020	2585	31	2554
04/10/2020	3908	74	3834
04/15/2020	130619	7672	122947
04/20/2020	326443	18652	307791
04/25/2020	436615	25705	410910
05/01/2020	515576	31842	483734
05/05/2020	561470	34182	527288
05/10/2020	625946	38446	587500
05/15/2020	680173	42034	638139
05/20/2020	732398	44858	687540
05/25/2020	784045	47415	736630
06/01/2020	849940	50655	799285
06/05/2020	886338	52305	834033
06/10/2020	929645	54055	875590
06/15/2020	975503	55495	920008
06/20/2020	1038759	57050	981709
06/25/2020	1104655	58432	1046223
07/01/2020	1205765	60491	1145274
07/05/2020	1284999	61192	1223807
07/10/2020	1388488	62810	1325678
07/15/2020	1503448	64192	1439256
07/20/2020	1626670	65743	1560927
07/25/2020	1756264	68034	1688230
08/01/2020	1925549	70771	1854778
08/05/2020	2000485	72072	1928413
08/10/2020	2102576	73367	2029209
08/15/2020	2197800	75444	2122356
08/20/2020	2276011	77204	2198807
08/25/2020	2358032	78780	2279252
09/01/2020	2475466	81236	2394230
09/05/2020	2550763	82689	2468074
09/10/2020	2616739	83918	2532821
09/15/2020	2699039	85415	2613624
09/20/2020	2792140	86785	2705355
09/25/2020	2879397	88414	2790983
10/01/2020	2988233	90009	2898224
10/05/2020	3066371	90909	2975462
10/10/2020	3182475	92432	3090043
10/15/2020	3305166	93699	3211467
10/20/2020	3440605	95094	3345511
10/25/2020	3606521	96895	3509626
11/01/2020	3867726	99417	3768309
11/05/2020	4052847	101196	3951651
11/10/2020	4346956	103238	4243718
11/15/2020	4694410	105859	4588551
11/20/2020	5075047	109352	4965695
11/25/2020	5420314	112452	5307862
12/01/2020	5834653	115881	5718772
12/05/2020	6170432	119987	6050445
12/10/2020	6574954	124411	6450543
12/15/2020	6997693	128753	6868940
12/20/2020	7386553	133790	7252763
12/25/2020	7708482	138161	7570321

Table III: Grouped selected attributes by [cdc20]

Date	Population
midrule 03/25/2020	329433166
04/01/2020	329459499
04/05/2020	329476690
04/10/2020	329498178
04/15/2020	329519667
04/20/2020	329541155
04/25/2020	329562644
05/01/2020	329588430
05/05/2020	329606219
05/10/2020	329628455
05/15/2020	329650692
05/20/2020	329672928
05/25/2020	329695164
06/01/2020	329726295
06/05/2020	329746456
06/10/2020	329771658
06/15/2020	329796860
06/20/2020	329822061
06/25/2020	329847263
07/01/2020	329877505
07/05/2020	329899443
07/10/2020	329926866
07/15/2020	329954289
07/20/2020	329981711
07/25/2020	330009134
08/01/2020	330047526
08/05/2020	330069263
08/10/2020	330096434
08/15/2020	330123605
08/20/2020	330150776
08/25/2020	330177947
09/01/2020	330215986
09/05/2020	330238125
09/10/2020	330265798
09/15/2020	330293471
09/20/2020	330321145
09/25/2020	330348818
10/01/2020	330382026
10/05/2020	330400989
10/10/2020	330424693
10/15/2020	330448397
10/20/2020	330472101
10/25/2020	330495805
11/01/2020	330528990
11/05/2020	330546051
11/10/2020	330567378
11/15/2020	330588705
11/20/2020	330610031
11/25/2020	330631358
12/01/2020	330656950
12/05/2020	330671932
12/10/2020	330690661
12/15/2020	330709389
12/20/2020	330728117
12/25/2020	330746845

Table IV: Population of USA [pop20]

State	Death	DeathClass	Recovery	RecoveryClass
NC	6152	OVERMEAN	461328	OVERMEAN
ID	1191	MEAN	111206	MEAN
WY	373	MEAN	36327	MEAN
NY	20151	OVERMEAN	345322	OVERMEAN
OH	7687	OVERMEAN	582943	OVERMEAN
CT	4686	MEAN	157763	MEAN
MS	3718	MEAN	140018	MEAN
AZ	7635	OVERMEAN	457378	OVERMEAN
IL	15799	OVERMEAN	915050	OVERMEAN
VA	4272	MEAN	274881	MEAN
TN	5646	OVERMEAN	480082	OVERMEAN
ME	314	MEAN	18203	MEAN
MT	916	MEAN	78013	MEAN
CO	3948	MEAN	304959	OVERMEAN
MA	11706	OVERMEAN	316601	OVERMEAN
WI	4679	MEAN	463220	OVERMEAN
AL	4096	MEAN	273658	MEAN
MI	11775	OVERMEAN	458153	OVERMEAN
OR	1407	MEAN	104591	MEAN
OK	2218	MEAN	227244	MEAN
KY	2312	MEAN	199732	MEAN
DE	791	MEAN	50695	MEAN
GA	9656	OVERMEAN	527423	OVERMEAN
SC	4662	MEAN	258730	MEAN
PR	1187	MEAN	67274	MEAN
UT	1182	MEAN	259407	MEAN

Table V: Classified states by mean feature