

# DME Integrative Assignment

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# 1 Task 1: Plan & Explore

Please note: I try to be precise but sometimes I will not describe certain features: the fact that a primary key is unique should be given etc. I try to describe everything but sometimes I just have to assume that the corrector knows what this is about.

## 1.1 Origin of Data & Purpose Introduction

**Introduction** Data drives security. The wide spread adoption of data information systems has been used to recognize crime hot-spots to increase policing efficiency and protect people. Thus, creating effective information systems for crime prevention is at the center of policy makers and executive branches of governments. This trend has led to the Chicago PD reaching out to this data management team with the request of creating a Data Management System for crime-data in Chicago.

## 1.2 Purpose of this assignment & data plan

To this end, a research database will be created in accordance to the FAIR principles (**Findability, Accessibility, Interoperability, and Reuse**). The expressed goal of this assignment is to offer a database and some fundamental queries for data exploration; i.e. a database with flexible example queries meant to further analysis of the trends discovered in the data. Additionally, any code produced during this endeavour will be made public on GitHub for transparency purposes.

**Data Description in Scope, Volume, and Format** The data was provided by the Chicago PD contains a sample of 730,900 registered crimes in the administrative districts of Chicago between the years of 2017 and 2021. The initial data source is in CVS-format and pertains to specific recorded crimes in the administrative jurisdiction of the Chicago PD in addition to time, location, crime type, and arrested or not. Thus, the data contact is to be found on the Chicago PD website (**CITE THE CHICAGO WEBSITE FOR THE DATA**). Considerations regarding the ethical (& GDPR compliant regardless of whether the data comes from the USA) use of the data will be discussed in part 5 of this assignment. It is notable that, while this data is only a subsample of the total database crime data recorded in this timeframe, the overall trends in the data are still maintained. Thus, the EDA for which this database can be used for further analysing trends uncovered during this assignment.

**Project time frame, researchers, and misc information** The set project time frame is the 27th of August 2022 till 2nd of October 2022. Involved in this project is only one student, Angelo Barisano. Additionally, this project is designed to comply with FAIR standards (**CITE FAIR STUFF HERE**).

**ETHICAL CONSIDERATIONS, FAIR, AND GDPR COMPLIANCE?!?!?**

## 1.3 Research Question

In an iterative (agile) development cycle the following questions have been set up: Chicago is known for its high homicide rate; this leads to policy makers focusing on this issue the most.

Upon conducting an initial Exploratory Data Analysis (EDA), three major categories were identified in the data:

- Location
- Crime type
- Crime record data (time, arrest, etc.)

Based on the data categories and EDA, the following research questions guide the creation of the database in increasing complexity. It may be noted, that the questions are meant to not be exact in their nature, but that the purpose of the assignment, database, and questions posed are of an explorative nature, evolving while answering said questions. Thus, this assignment is interpreted as an initial exploratory data analysis meant to uncover trends for further analysis in addition offering useful insights to policy makers.

**Question 1** Finding prevalent crime patterns in the data is a common starting point in exploring crime data. Crime trends take the form of temporal patterns distributed over a defined timeframe of interest. This kind of reporting is commonly used by policy makers (such as attorney generals) who are monitored based on their performance in terms of crime prevention and the overall trend that can be observed.

Thus, this question provides future research with an adjustable query to gain an overview over the general distribution of crime by type. Subsequently applying a temporal component (i.e. by day of week, month, year, season, etc.) to this initial distribution reveals trends and temporal patterns. This way, this question helps to answer questions to policy makers regarding general trends; such as how crime developed overall and by type.

More importantly, however, is that the descriptive analysis of temporal patterns may indicate areas of interest that may be examined more closely by further questions (Q2 & Q3) through its explorative lense. As such:

- What are the overall crime trends that can be observed in Chicago between 2017 and 2021? What crime types changed a lot? Which crime types increased during the timeframe

In order to keep the report in a reasonable frame, a primary focus will be laid on **WHAT??**

This question will be answered by using a view which contains the date per crime in a convenient manner, such as year, month, day, hour. This way, the analysis is decreased in complexity; providing future researchers with easy to use time dimensions. Moreover, in this question, a variety of window function will be used to group by the time dimension and return a distribution per time component (eg year).

**Question 2** The next logical progression is to observe the location and crime dimension together. Certain crime trends, such as those of homicides, may be more localized; thus, certain neighborhoods and districts may be overrepresented in heavy crimes. The assumption is that certain districts and beats tend to be more prevalent in certain crime types. As such, in order to help the PD and policy makers to identify problematic areas with respect to certain crime hotspots, districts and beats are analysed with respect to trends identified by the temporal component on a localized level. Thus, question two follows:

- Based on overall trends discerned in question 1, what locations (beats & districts) are disproportionately represented overall and in selected subcategories of crimes (based on Q1)?

This question will be answered similarly as Q1. However, the main focus here is laid on using standard *GROUPBY* functions instead of window functions to demonstrate a variation in usage.

**Question 3** Finally, time, location, and crime type is triangulated. This enables policymakers to discern localized trends in the data in order to address crime patterns by distributing resources more efficiently; i.e. allocating more resources to dangerous beats during the night. Additionally, accessory dimensions will be integrated into the analysis to provide a holistic description. For instance assuming that crimes that lead to more arrests are more resource intensive, these crimes put a disproportionate strain on law enforcement. Thus, by triangulating arrests by location and e.g. time of day this will enable us to show areas that need more attention by law enforcement. Another angle would be a specific analysis of beats. **Beats are the smallest administrative unit of a police district; a beat is patrolled for one year by one unit and then transferred to another beat. Thus, it might be interesting to investigate the connection between a subset of beats that suddenly stop showing problems during one year and then re-appear in terms of crime in another year. The sub-question would, thus, investigate whether beats that usually persisted in crime only persist on a closed yearly basis. This way, effective police units might be identified and resources might be allocated more efficiently.**

- Triangulating time, crime-type, and location which areas persists in certain crimes wrt. time? In order to prevent homicides; which “beats“ are the most prevalent among homicides? During which time of day (for effective allocation of policing resources)?

#### **BASED ON THE LOCATION AND THE**

As such, the final question considers a variety of hypotheses that can be explored. Overall, these project based questions are constructed in such a way that they guide an external user (PD) through

the process of finding areas that need successively complex "sliced & diced" information and culminate in the creation of actionable policy implications regarding prime prevention through resource allocation.

AS such, by the descriptive and explorative nature of the questions posed, one can deduce that the entire analysis can be conducted using simple SQL functionalities.

**ADD PLOTS of EDA ALREADY HERE!!!**

## 2 Task 2: Design and Organize

**IMPORTANT: DISCUSS what 1st and second normal form means that that all entities are in 2nd normal form by default and due to the structure in 3rd nf!!!**

**IMPORTANT!!!! READ THE DEFINITION FOR 3rd NF AGAIN; MAYBE CRIME AND LOCATION QUALIFY TO BE 3rd NF... but maybe also not. STRESS AGAIN THAT THE INCLUSION OF BLOCK IS NOT NEEDED FOR THIS RESEARCH!!!**

The aforementioned questions in task 1 require three relevant datacategories, as identified in the data. These pertain to the 1) time dimension, 2) location dimension, and 3) the crime or case instance itself. Generally, these already display common characteristics of regular entities in a relational database (**CITE HERE BOOK FROM BACHELOR COURSE**). Generally, the design of the database followed the approach of combining logical structures (e.g. location) with an easy to use query design to explore the data.

We will start out with the case entity, as cases recorded in the original data create the centerpiece for all three posed questions.

**Entity 1: Case** The first component consist of the individual instances of cases. Conceptually, these are central to this project as they enable the creation of the frequency distributions conditional on time and/or location. The variables that define this entity are as follows:

- CaseId
- DateTime object implicitly containing day, month, year, and time of day
- Arrest Boolean
- **Location Description??**

It may be noted that items such as location description and location themselves, though listed, are not included in the final product in order to comply with the principle of data minimization (**CITE GDPR**). This is also reduces the possibility of mistakes from occurring.

The DateTime object will enable the clustering of crimes by the time dimension; which will be done via a view. Additionally, arrest information is used to further drill down the analysis and dissect the cases for more resource intensive cases. This is part of discentring interesting trends and exploring the data. The primary use case of case as an entity, however, is to make the entire database 1) work from a logical perspective (no crime distribution without each individual crime) and 2) to be the logical link between the type of crime committed in every case and where it happens. This way, the case entity is not only a natural entity, but it also fulfills the purpose of making the following two dimensions compliant to the 2nd normal form (and to some extend compliant with 3rd normal form with some caveats) by default. Imagine that if we were to leave out case as an entity and immediately match crime types and location, the resulting two entities could not comply with 3rd normal form as crime types and location (as categories) would produce a many-to-many relationship - thus, not minimizing storage space, increasing query diffucilty, and cause an inefficient relational database. As such, the case entity inadvertently functions as an associative entity, reducing many-to-many relationships to two many-to-one relationships.<sup>1</sup> As a consequence, case complies to the 3rd NF by definition as no stand alone entities are to be found in this entity (i.e. no transitive dependencies as will be described later), while requiring the other data categories to normalize as well.

It is notable, that while such design choices should be reflected by leaving out case as an entity in the conceptual model in part 3 and then include it as part of the logical model, but case is so pivotal to the functioning of the database in its purpose, that we will consider it along the way a a valid entity.

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<sup>1</sup>I will not further elaborate on this; this is a logical conclusion; in order to reach 2nd normal form this is a required step which is obvious

Please note that any datetime object should not get its own entity. This would violate rules set for relational databases **CITE BACHELOR STUFF HERE** as any instance created for case requires an instance to be created in a datetime relation; this would make the relation a one to one relationship, which are not efficient from a storage perspective.

Consequently, no problems of resolving many-to-many relationships are imposed on the design (i.e. no normalization issues later), while minimizing storage usage, in addition to adding an intuitive centerpiece to the database being implemented.

**Entity 2: Crime Type** Every Case requires a crime to be a valid instance. The raw data provides the IUCR, which identifies each unique combination of primary (e.g. homicide) and secondary crime category (e.g. first degree). Generally, most headlines only consider the primary type of crime, such as homicides, and generally disregard the secondary description of the data, such as first, second, or third degree in this case **CITE SOME EXAMPLE NEWSPAPERS**. However, for certain crimes, such as homicide, it might be useful to be more differentiated wrt. eg. gang related homicides. It is also the task of a FAIR database to enable future users to differentiate between different types of general crimes. Furthermore, it might be the case, that certain subcategories of crimes tend to be more prevalent than others. Subsequently, due to the inclusion of the secondary description of the crime, this necessitates that crime type is an entity on its own. Suppose only the primary category of any crime committed was to be analysed. This would imply that both the IUCR and secondary description would be superfluous. Subsequently, any entity with only one variable is a superfluous entity, taking up unnecessary storage space (similar to one to one relationships). The reason for this is that if a crime occurs, a foreign key would have to relate the primary crime type to the crime type entity. Obviously, the primary crime type foreign key would be the only variable (and primary key) in crime type. As such, it would not make sense from a relational perspective to split crime type into a separate entity. It would make more sense to integrate it directly into the case entity. However, now, due to the inclusion of secondary description as an attribute, crime type has to be its own entity in order to move the case table from 2nd to 3rd normal form (the crime type information would not be transitive). Subsequently, the creation of a crime type entity moves both the case entity and crime type entity into 3rd normal form. Moreover, IUCR then becomes a natural PK (and FK) in this relationship. Subsequently, through the initial design choice, any issues regarding normalization are being taken care of.

**Entity 3: Location** Finally, the location information is being considered. Here, a similar logic applies as with the crime type. In order to analyse certain administrative boundaries, such as districts etc., a location entity has to be included. Hypothetically, if we were to only include information in the database pertaining to in which district an instance of a crime was recorded, the aforementioned case entity would be the most optimal place for such a variable to be stored. However, this research database aims at offering the ability to further analyse certain subsections of particular police districts - their beats in order to offer better descriptive results. Subsequently, any instance of a crime recorded will have its district and beat included. Following hierarchically, beats are below district; one district containing multiple beats. This implies that the most optimal database structure is to use beat as the primary and foreign key in this relationship. This way, any normal form issues are resolved by default, setting both the case and location entity into 3rd normal form.

The reason to exclude block as an entity of analysis is twofold. 1) blocks overlap to some extent with different districts and beats increasing the level of analysis unnecessarily. 2) Beats and districts are administrative units, while blocks require local knowledge, which is what we try to use to answer for the aforementioned questions.

**Logging & Master Table** Finally, a small master table will be included which tracks any information regarding formatting the data, data retrieval, and restrictions/ triggers called (not included in part 3 due to relevancy). Such relations are not part of the standard relationship model and are, thus, excluded from the ERD development in part 3.

**Conclusion** Based on the questions posed during task 1 in addition to the fundamental nature of the data provided, the structure described in this section offers an intuitive optimal solution for any normal form problems that might arise. Particularly the inclusion of the case entity as an integral part of the relationship model automatically resolves any issues pertaining to the inclusion of location and crime type information on a case by case level.

### 3 Part 3: ERD

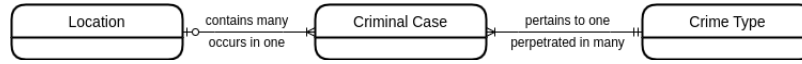


Figure 1: Conceptual Diagram

#### 3.1 Conceptual Model

As mentioned in part 2, the individual criminal case instance is central to the database design. Contemporaneously, the case entity resolves all problems regarding normal form (all tables are in 3rd NF) and many-to-many relationships. Subsequently, the conceptual model describes the relation of the three entities described in part 2:

- Location
- Case
- Crime Type

Disregarding information regarding keys, the structure outline in part 1 & 2 is described here from left to right. To start with, the location entity describes where any crime instance is taking place in terms of beat and implicitly the corresponding district. Consequently, location is related to criminal case with a one (non-mandatory) to many (mandatory) relationship, read from left to right. The reason for this specific choice of cardinality is simple: in order to be included in the data, a district or beat must have had at least one crime happening in it; otherwise it would not show up in the raw data. As we are not creating a true transactional database, but rather a research project database, we do not have to care for the hypothetical case of a district being included just to make sure you can create future crime instances in this specific location instance. In order to be included in the raw data, a location (be it beat or district) must have experienced at least one crime instance. Thus, for this specific case of a research project driven database, the *model* is requiring any location to have at least one crime instance. A similar argument will be used further down as well. Contrarily, certain crimes do not have a location; think of financial crimes. It would be wrong to remove these records. Thus, a criminal case instance does not need a location.

As such this is then read as: one location must have experienced at least one or many criminal cases; One criminal case can occur in one and only one, but not mandatory, location instance.

Subsequently, we need the criminal case entity, which contains all case related information on an individual subject basis. Continuing to the right of it, case and crime type is related via a mandatory one or many and mandatory one and only one relationship. Crime type gives the general description to each criminal case (description is the same as in part 1 and 2). The choice for this cardinality is similar to the aforementioned relationship. A criminal case must have a crime type, otherwise it is invalid as no crime would have been committed. Contrarily, a crime type, in order to be included in the initial raw data, must have been committed once at least. This is actually the case here. Certain IUCRs, which exist in the penal code in Illinois, were not committed and are thus not included in the raw datafile.

As such this is then read as: one criminal case pertains to one (mandatory) and only one crime type; one crime type can be perpetrated in one or many (mandatory) criminal cases.

**Intermezzo:** It may be notable that the model shows mandatory relationships for crime type and location. Technically correct would be to use non mandatory relationships for location and crime type. For example a district or crime may not have been perpetrated and, thus, would not have a record. This research does not consider this case as the goal of this data base is to use the existing data for analysis. As such, in order for a district/ crime to be included in the database, a crime instance must have been created for these instances. As such, the choice was made to include a mandatory relationship for these cardinalities.

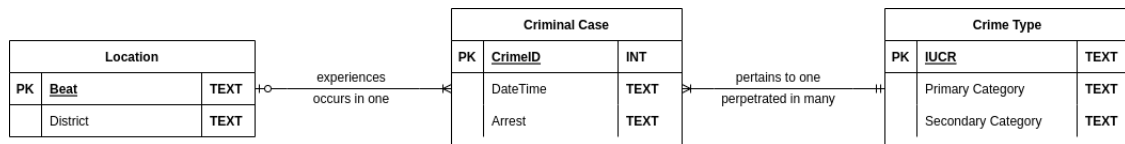


Figure 2: Logical Diagram

### 3.2 Logical Model

Figure 2 describes the logical model. For location, "beat" is chosen as primary key because beat (as a stand alone entity) would be hierarchically below District; one district - many beats. As such, each district is uniquely identifiable by its beat in case we want to aggregate, which is why beat is the only natural (primary) key in this relation. In reality (with more variables in location) this relation is not in 3rd normal form as district is usually an entity on its own. However, due to no more data/ variables being available on beat or district, this relation does not contain any transitive dependencies and is, thus, in 3rd NF. The meaning of each attribute here is selfexplanatory. More importantly, the primary key beat is in TEXT form (also district). The reason for this is that beats are defined in the first two characters by their district and the latter two describe the beat. Thus, a beat instance may start with a "0". In many scripting languages and some databases, this leads to problems and the removal of the "0", as the underlying software does not interpret the "0". As such, TEXT was chosen for data quality reasons. Please note that the same argument applies to crime type and IUCR.

Following, CrimeID was chosen to be the primary key of case, due to it being a natural primary key. This is a standard auto-incremental integer key and can, thus, be treated as an integer; though care should be applied when reading the data into scripting languages. In addition to each CrimeID identifying each instance of a crime committed, DateTime and Arrest are included as attributes. DateTime has TEXT as dtype as SQLite does not support a conventional datetime object, which then defaults to a string value. DateTime provides year, month, day, hour, minute, second of the instance 8 as any other datetime dtype does by definition). Arrest has the dtype TEXT for a similar reason; SQLite does not support boolean dtypes. A possible alternative would have been to use 1 or 0, but a simple string input TRUE or FALSE has the same functionality in the and while being more explicit. Arrest describes the state of each criminal case whether or not an arrest was made.

Finally, crime type's primary key is the IUCR key, a natural primary key. It uniquely identifies each crime category plus its secondary description. Its dtype is TEXT for the same reason as for why beat's dtype is TEXT. Following, primary category has dtype TEXT and 'describes the overall category of the crime committed. Secondary category then further describes each primary category; thus, it also has a TEXT dtype.

### 3.3 Physical Model

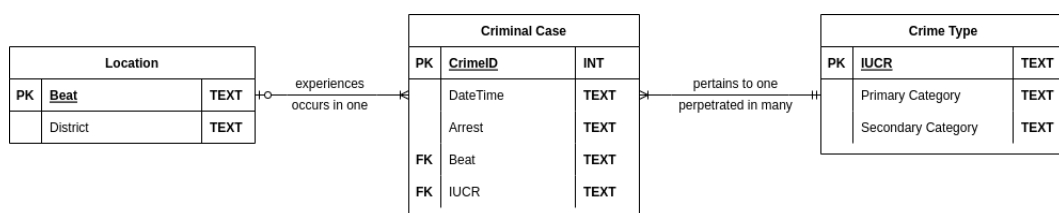


Figure 3: Physical Diagram

As mentioned in part 2, the inclusion of case as a full entity through the entire process automatically removes any many-to-many relationships. Additionally, this simplified the choice in appropriate foreign keys to describe the relationships. Remember, foreign keys must be placed in the adjacent entity on the many side (the other way around would be counterproductive). Thus, beat is the foreign key in criminal case for location and IUCR is the foreign key in criminal case for crime type. The reason for this should be selfexplanatory. It may be noted that CrimeID is still the unique identifier in criminal case as it is unique by its own right.

Thus, planing in part 1 and 2 simplified this part considerably by default.

## 4 Part 4- Data Quality checks and preparation

In the process of preparing the data to be loaded into the research DB, python was the scripting language of choice.

In order to conduct the data cleaning, the GDPR's dataminization principle was used (the others, though important, are less applicable here **CITE GDPR HERE**. Dataminization states that only data that is actively being used and has a purpose of being stored should be included in a database. While this project is based on a public database from the US (so technically GDPR does not apply; particularly because no individual is identifiable), GDPR still offers guiding principles/standards on how to fair with data. Additionally, dataminization has a practical usage point: less data means fewer errors.<sup>2</sup> Thus, not all data in the raw data is included. This also means that considerations regarding FAIR were made (see part 1).

In order to prepare the data, the **data triangle was used (CITE THE LECTURE HERE**.

**DATA TRIANGLE** in lecture!!! **ALSO THE META DATA AND ETHICAL CONSIDERATIONS!! REGARDING GDPR** that data quality is important to be complied by!!! This is also important for fair wrt interpretability and reusability

After the setup of the data management plan in part 1 to 3, the implementation requires data preparation and data quality check. **In this context, the data triangle was used.** Generally, the data quality of the provided sample is very good; less than 0.01% of observations display any abnormality. This is attributable to the origin of the data being of administrative nature by public offices. The operations performed specifically relate to the relevant data for this endeavour mentioned in part 3. Firstly, 3 instances contained no casenumbers. These records were broken overall and were, thus, removed. Secondly, 16 instances missed date. As date is a central to the analysis, these records have to be removed. Moreover, 79 instances had non-correspondent beats and districts. While it would have been possible to impute the missing data based on the block or lat-lon information, one must assume that all location related information wrt. these observations is compromised. Similarly as before, location quality is central to answering the questions in this assignment; thus, these instances were removed. Finally, 149 perfectly duplicate items were removed.

In terms of data transformations, date was transformed into SQL castable form to run functions on date. Moreover, beat and district instances were repaired wrt. the preceding "0" filler, creating instances of length four and two respectively. Finally, arrest ("true", "false") was cast into boolean form.

Regarding non-crucial abnormality, 15 instances miss arrest; but this information is only used for one subquery once. Additionally, seven records of casenumbers display noncompliance. However, CaseID is a perfect reflection for this issue. Considering CaseID, set-tests were made to warrant that no IDs were missing (see appendix) beyond the removed records. Further, one instance of lat-lon showed an extreme outlier which was converted to NaN, due to the beat and district information appearing properly. **All instances with minor problems were marked, which will be represented in the MasterTable of this project.** It may also be notable that the absence of information is still information; i.e. financial crime might not have a location.

Overall, the consequences of removing these observations will be minimal. Of a total of 730900 observations, we were left with 730654 records.

**Considering the IUCR, there are some duplication values. It must be noted that most instances contain typos in the description; as such, the IUCR is always compliant. Thus we use a right join to remove all duplications of IUCR which will not change the data**

## 5 Part 5

DESCRIBE THE RESTRICTION THAT YOU INTEGRATED IN THE DATA!! ENCRYPTION OF DATABASE!!!!!!!

**NOT: ALSO ADD THE DISTINCT IUCR TO THE CRIME TYPE TABLE! CHANGE THE TABLENAMES IN THIS ENTIRE FILE!!!**

Notes: - reading data

## 6 Part 6

To do: create master table plus some triggers, add the stuff regarding index and use it, then finish q3.

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<sup>2</sup>Please note that I studied the GDPR at UvA and had to work with it; so I can speak from experience.



**It may be noted that we cannot query all possible outcomes It may be noted that we cannot query all possible outcomes It may be noted that we cannot query all possible outcomes It may be noted that we cannot query all possible outcomes**

Note: the goal of this assignment is to do most if not all data transformation in SQL and let python handle only minor conversions such as into the right dtype object (eg GEOPANDAS)

**IMPORTANT! READ DATABASE INTO PYTHON HERE LIKE YOU DID AT VYTAL!!!**

Note: Sadly in order to create an index, one cannot use subqueries due to SQLLITE (this feature is present in other versions of SQL tho); as such, the index that increases speed for arrest heavy crimes does not work here.

Please note that many queries, particularly those meant for visualization were read into python immediately (so they use queries to get the results needed and then let python do the rest). Additionally, the database was not saved directly in the repository due to security concerns (.gitignore). Moreover, no virtual environment was set up to facilitate the project. Finally, it is notable that in order to answer the questions, multiple different approaches in writing the queries were chosen, ranging from joining subqueries to window functions in order to demonstrate knowledge of all functions. Another reason is that I wanted to try the speed of different styles of code (particularly the procedurally generated optimization behind window functions).

## 6.1 Question 1:

The first question considered how overall crime patterns are distributed with respect to the temporal component. To this end the following queries were constructed

- Distribute Total Crimes per year to see the overall trend of crimes committed. This is relevant for policy makers such as attorneys generals which are responsible for crime levels in the city and their persecution.
- Further, this analysis is then extended to primary crime-categories by year and month in order to provide a more drilled down perspective to this question regarding overall specific crime. This is relevant as certain crimes are more relevant in the public eye, such as homicides.
- these crimes are then subsectioned into resource intensive crimes compared to non resource intensive crimes; this is decided on whether the crime type overall leads to a more than 50% arrest quota in addition to more than 500 total arrests for this category over the entire timeframe
- bad crimes by time of day are also addressed
- Finally, a window function is used in order to find the yearly change in crime by primary crime type

Moreover, in order to ease the analysis, a view with transformed date attributes (i.e. year, month, day, hour of day) for each crime was created.<sup>3</sup> Additionally, the analysis evolved organically and, thus, is not attached to the questions themselves.

- How do crime types (e.g. homicides) patterns distributed by a temporal component? How do certain crime types (homicide) patterns distribute by seasonality patterns, by time of day?

Very simple: -report how total crimes went down - then report the resource intensive crimes how they behaved (500 total arrests needed because 100 per year minimum) - then mention the crime category that showed the largest reduction among all crimes - then report one drill down into homicides for the secondary category to be used (report how each secondary category evolved)

**Figure 4** contains the overall analysis of the time component by crime. Primarily, these graphics are focused on providing policy makers of Chicago information regarding the overall crime level of the city

**Part 1** As can be seen in **Figure 4 a)** overall crime levels reduced during the observation timeframe of 2017 to 2021. In 2017, the overall level of crimes recorded was at 161304, which gradually reduced to 124558 crimes committed. Overall, this resulted in a reduction of 22.78% in total crimes recorded. Thus, on a superficial level, crime tends into the right direction. It may be noted, that a view (*view\_convenient\_time*) was created in order to facilitate an easier handling of the data and due to this query being performed frequently.<sup>4</sup>

<sup>3</sup>Please note that views should only be used for frequently used queries and converted variables

<sup>4</sup>Aforementioned variable transformations are supposed to be used in such views

**Part 2** However, overall crime levels do not consider the severity of the crime committed. In this context, this analysis focuses on resource intensive crimes. These were defined by calculating the proportion of arrests in this specific overall (primary) category. It was decided due to brevity concerns that if a crime category displayed more than 50% arrest rate in addition to more than 500 total arrests over the period of the five recorded years (100 per year), that this crime poses a reasonably higher strain on law enforcement; this is because if such a crime occurs, two law enforcement officers have to respond only to arrest the suspect in question and bring it to jail while not being able to respond to other calls and if they occur not frequently enough they do not pose a strain on law enforcement. Overall, of the 34 primary categories defined in the data seven fulfill this classification. **Figure 4 b)** shows the development over time of these categories. Primarily Narcotics, while increasing initially, Narcotics related crimes dropped by 56.92%, posing the most frequent category in this category, starting at 6946 in 2017 and dropping to 2992 records. The largest reduction was observed in Prostitution related crimes from 438 in 2017 to 53 in 2021, which constitutes a reduction 89.90%. Finally, the only category that showed an increase in this subcategory was Weapons Violation, which displayed an increase of 93.89% from 2814 to 5456 recorded crimes.

Overall, resource intensive crimes showed an overall drop of 29.95% from 15958 to 11178. Combined with the insight that overall crime levels have dropped, this implies that law enforcement is less strained in 2021 than in 2017. It may be noted that the final drop may be explained by the outbreak of the Corona Virus.

**Part 3** When considering all crime categories again, the question which categories, regardless of whether they are resource intensive or not, have shown the largest rise and reduction may be considered. **Figure 4 c)** displays the top five categories with the largest reduction, while **Figure 4 d)** shows those categories with the largest increase in records. While the overall trend was positive in terms of overall crime levels reducing, crime categories such as Homicide and Human Trafficking show a 19.94% and 80.0% increase respectively. This implies, that while the overall trend in the crime is positive, certain subcategories display a severe increase.

## 6.2 Question 2:

- Do certain crime patterns (e.g. for homicides) persist by location such as certain districts/ beats/ blocks?

When using the insights gained from the first question (regarding overall decrease in intensive crimes but a tragic increase in homicides etc, we can then proceed by examining those locations that are indeed struggling with these crimes mentioned above

overall points: - show a plot that shows that overall levels of total crime levels are similar in most districts - then proceed with a slight drill down into the data by looking at how crime types distribute: meaning that eg there might be a certain district with very high homicide rates based on the percentage of a certain crime type by district – look specifically for those crimes discovered in question 1.

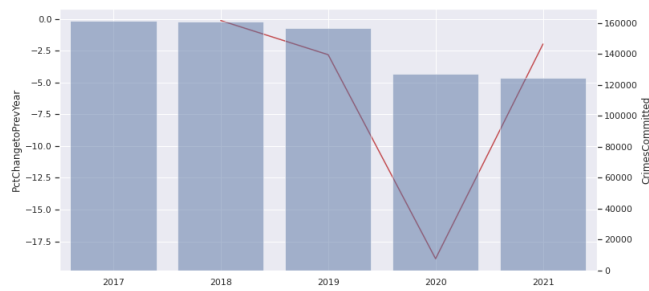
USE AN INDEX ON HOMICIDES AND OVERALL BAD CRIMES

## 6.3 Question 3:

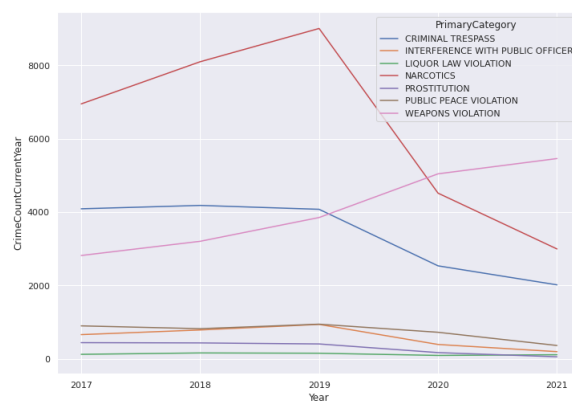
Finally, based on the trends observed in question 1 and the locations in question 2, we will look at how we can prevent these crimes by allocating resources efficiently to those locations by time of day during time of day (eg morning)

- Triangulating time, crime-type, and location which areas persist in certain crimes wrt. time? In order to prevent homicides; which “beats“ are the most prevalent among homicides? During which time of day (for effective allocation of policing resources)?

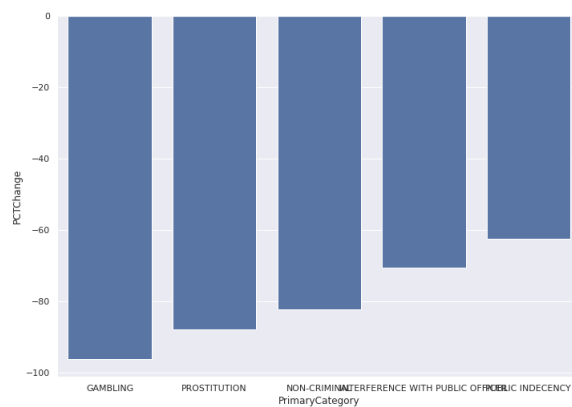
## 6.4 Conclusion



(a) Total Crimes Recorded



(b) Resource intensive Crimes Annual Change



(c) Crime Categories with the greatest Reduction

