# **Metric**Stream



# WHITE PAPER: AI and Contract Management Summer 2020

**Collaboration By: MetricStream and Golden Gate University** 

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#### **Foreword**

The age of artificial intelligence has made tools more accessible than ever before. Organizations are automating manual processes to achieve quicker time to insight and eliminating the tediousness of repetitive tasks. In the realm of Contract Lifecycle Management, organizations are utilizing natural language processing to extract relevant contract information to ensure information veracity and to capture anomalies that manual review might miss.

# **About Us (Golden Gate University ~ MetricStream Collaboration)**

MetricStream is an independent market leader in enterprise cloud applications for governance, risk, compliance (GRC), and quality management. Through the enterprise platform and cloud software, MetricStream enables enterprises across industries to drive exceptional business performance based on the foundation of good governance, trust and integrity.

Founded in 1901 in San Francisco, **Golden Gate University** offers undergraduate and graduate programs in business and management, accounting, taxation, and law.

#### **Contract Lifecycle Management**

Contracts are a legally binding agreement. Two different parties, such as companies, enter contracts usually to exchange of goods and services for currency. The starting point of a contract's life is usually a contract template for contracts entered into many times with different parties. The terms of the contract are sometimes negotiated between two parties and agreed upon. Once agreement occurs, the final contract version is signed and both parties are responsible for delivering what they agreed to in the contract. The legal department at a company usually drafts the contract before it is negotiated but contracts entered can impact many departments of a company. For example, sales, finance, and accounting, operations, supply chain, manufacturing and more. The life of a contract continues through to amending a contract, renewal, or termination of a contract. Contract Lifecycle Management is the management of the contracts through the process of initiation through execution, performance and renewal/expiry.

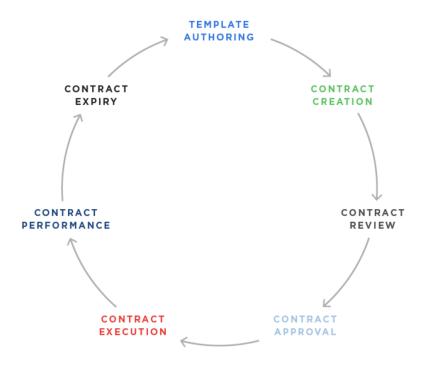


Figure 1: Steps in the Contract Lifecycle Management

The correct CLM solution connects the contracts process to order fulfillment and billing, eliminating errors, reducing risk and ensuring that revenue can be forecasted more accurately.

## **Introduction to Machine Learning**

Machine Learning is an application of Artificial Intelligence (AI) that helps the system to automatically learn, perform and improve from experience. Machine Learning focuses on the development of the computer programs that can access the data and use it for themselves. **Primary aim is to allow the computers to learn automatically without human intervention and act accordingly.** Machine

Learning enables analysis of massive quantities of data, while generating faster and more accurate results in order to identify profitable opportunities or dangerous risks pertaining to the organization. Combining ML with AI and cognitive technologies can make it even more effective in terms of processing large volumes of information.

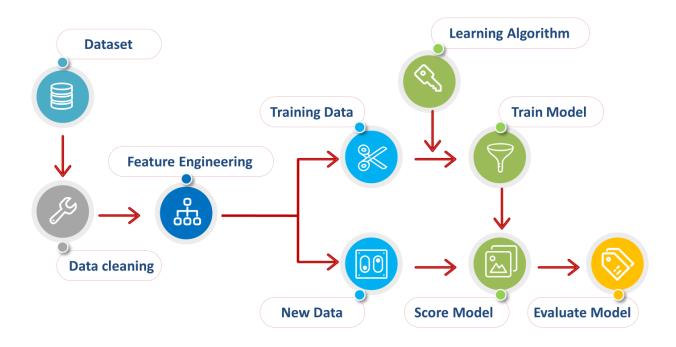


Figure 10: Machine Learning Process overview

It is very important for companies who practice CLM to understand the importance of applying Machine Learning combined with AI and produce customized outputs and perform the analysis.

#### **Artificial Intelligence and Contract Management**

Key terms and conditions in a contract typically include the contract name (i.e. the type of contract), the parties involved, effective date, any fees and / or rates, contract milestones, renewal or termination dates, and so on. Companies must understand the key terms and conditions after signing a contract to ensure that they are meeting their responsibility, especially as the company grows and enters more and more contracts. If they do not, they may incur legal fees, financial obligations, unsatisfied customers, early termination of contracts, and much more.

In order to track these key terms and conditions, the company may manually input all that information somewhere to track it, such as into a system or spreadsheet. As the variety and number of contracts entered increases, the more time-consuming and tedious this task becomes. Importantly, customers may negotiate changes to the standard template which must be tracked and met on an on-going basis. This is how contract lifecycle management (CLM) becomes challenging. If this process is not well managed, a customer may not be billed or underbilled, a contract renewal may be missed, the services rendered may not meet a service-level agreement noted in the contract, or worse.

One solution to this problem is to automate the extraction of key words and phrases in contracts so that they are all shown in one file or output. These key words and phrases will be categorized by the relevant actor who is responsible for this area. For example, if a clause is mainly related to finance or a service-level

agreement then a team member on one of those teams or both should know about it and take appropriate action. Using this type of output, employees may be assigned tasks and be more aware of deliverable requirements before, during, and after the products and/or services are rendered, contracts that may be expiring soon, and more.

CLM is one use case in extracting key data from sometimes lengthy documents, including Microsoft Word documents or PDFs, and turning them into actionable insight potentially saving the company time and money in the process. This type of automated data extraction can be applied to other documents, such as lease agreements, line of credit agreements, invoices, purchase orders, financial statements, and so on.

# **Usefulness of Machine Learning Model to MetricStream**

Aside from the benefits of automating manual processes, machine learning is especially useful in domain specific oddities. Contracts often use highly specific terminology to define an action or state. Contracts are structured with odd breaks in lines and use a variety of punctuation and characters. Capturing all these structural issues in a set of rules can be difficult, but through machine learning, these oddities can be dealt with via training a contract specific model.

The out-of-box natural language processing components of the "spacy" library are utilized in several sections throughout our tool pipeline. Firstly, "sentencizer" is a natural language processing tool that breaks a document into sentence-like chunks rather than on specific demarcations. Named entity recognition uses specific data labeling that can be used to extract relevant metadata terminology. Dependency parsing looks at the relationship between words and aids in grabbing relevant terminology based on its relationship to other relevant terminology. Synonym usage and lemmatization allows us to broaden search and capture the relevant metadata terminology as it appears in contracts.

These out-of-box tools can be further trained and specialized towards our contracts as we generate labeled training sets for our model. This will further strengthen our tool the longer it is used.

"Sentencizer"	Breaks up document into sentence-like chunks
Named Entity Recognition	Label words with entity tags
Dependency Parser	Displays relationship between words
Synonym and Lemmatization	Finds synonym and similar words for search

**Progress to Date (Use cases of AI and Contract Management)** 

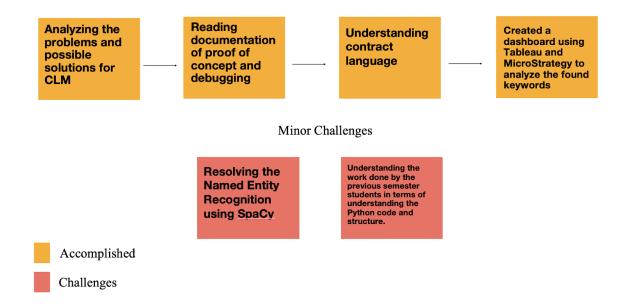


Figure 2: Starting Point

This phase highlights the initial stage of our internship wherein we were trying to understand the work done by the previous semester students as well as analyzing the possible problems and solutions for the contract lifecycle management. We read the documentation of the proof of concept and debugged the errors from the Python code/structure to suit the requirement. By gradually understanding the contract language, we created a drill down dashboard of the demo contracts using Tableau and MicroStrategy that helped us analyze the found keywords. Some minor challenges included resolving the named entity recognition using SpaCy library and deeper understanding of the work done by previous semester students in terms of the Python code and structure.

#### Phase 2:

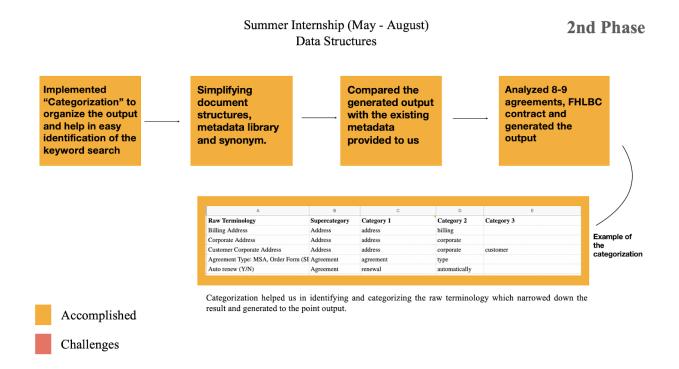


Figure 3: 2nd Phase

In this second phase, we concentrated categorizing various terminologies that will help us in the easy identification of the keywords search. Then we focused on simplifying the document structures and the metadata library. Categorizing the raw terminology narrowed down the result and generated the desired output.

# **Overview of the Categorization List**

# **Initial Metadata Categorization**

A	В	С	D
Combined Metadata Categorization	- Category -	Sub-Category =	Category (Detailed)
Customer Corporate Address	Address	Customer	
Corporate Address	Address	Corporate	Organization
Billing Address	Address	Billing	String
Agreement Type: MSA, Order Form (SBD), SOW, CR, Work Order, amendments	Agreement	Amendments	Contract
Type of Support	Agreement		
Churn ACV	Agreement	Churn/AVC	
Churn Type (What is being churned)- text	Agreement	churn	type
breach	Agreement	Breach	Contract
backup	Agreement	backup	Contract
breach	Agreement	breach	Contract
Renewal Price Increase	Financial	Value	Renewal
Escrow Y/N	Financial	escrow	
uptime	Agreement/Time	Day/Date	
License Fee	Fee	Value	License
Cloud Fee	Fee	Value	
Platform Fee	Fee	Value	
Support Fee	Fee	Value	
Third party product fee	Fee	Value	
Auto fee increase (Y/N)	Fee	Value	

Figure 6: Categorization List (Initial Metadata Categorization)

# **Full Metadata Library**

Category Sub-Category		Metadata_Original	Metadata	Term1	Term2	Term3
Value		99	99	59		
/alue		99.8	99.8	99		
/alue		99.9	99.9	99		
Γime	Hours	99.99	99.99	99		
Гime	Days	12 hours	12 hours	hours	12	
l'ime	Hours/days	15 days	15 days	days	15	
Гime	Hours/days	24 hours	24 hours	hours	24	
Γime	Hours/Days	30 days	30 days	days	30	
Гime	Hours/Days	48 hours	48 hours	hours	48	
Person	Name	Account Owner	Account Owner	owner	account	
Agreement		Agreement Type	Master Service Agreement	Master Service Agreement		
Agreement		Agreement Type	Order Form	Order Form		
Agreement		Agreement Type	Statement of Work	Statement of Work		
Agreement		Agreement Type	Cost Reimbursable	Cost Reimbursable		
Agreement		Agreement Type	Amendment	Amendment		
Agreement		Agreement Type	Work Order	Work Order		
inancial	Value	Annual Renewal Value	Annual Renewal Value	value	renewal	annual
ree .	Value	Auto fee increase (Y/N)	Auto fee increase	fee	automatically	
ime .	String	Auto renew (Y/N)	Auto renew	renewal	automatically	
greement	backup	backup	backup	backup		
Address	Billing	Billing Address	Billing Address	address	billing	
Agreement	Breach	breach	breach	breach		
Agreement	Churn/AVC	Churn ACV	Churn ACV	churn	acv	
Гime	Date	Churn Effective Date	Churn Effective Date	date	churn	effective
Гime	Date	Churn Notification Date	Churn Notification Date	date	churn	notification
Agreement	churn	Churn Type (What is being churned)- text	Churn Type	churn	type	
Person	Name/ Yes or No	Client Consent required to assign - Y/N	Client Consent	client	consent	
erson	Name	Client Name	Client Name	client	name	
roduct	SKU	Cloud Application SKU	Cloud Application SKU	sku	cloud	application
ee e	Value	Cloud Fee	Cloud Fee	fee	cloud	
egal		Contract Term- 1 year, multiple years, optional years	Contract Term	contract	term	
egal	contract	Contractual Relationship	End User License Agreement	End User License Agreement		
egal	contract	Contractual Relationship	Cancellation Letter	Cancellation Letter		

Figure 7: Full Metadata library (including key terms and categorization)

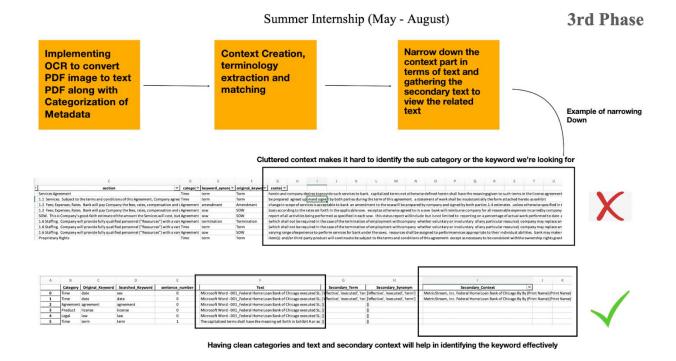


Figure 4: 3rd Phase

The third phase focused on narrowing down the context part of the output generated. The cluttered context made it hard to identify the categories as well as the keywords we were looking for. After running the code, the output generated had the text, secondary term, secondary synonyms and the secondary context. This was done in order to gather the surrounding text which the original text/context missed out. This gathering of the secondary text will help in better understanding of the output. `

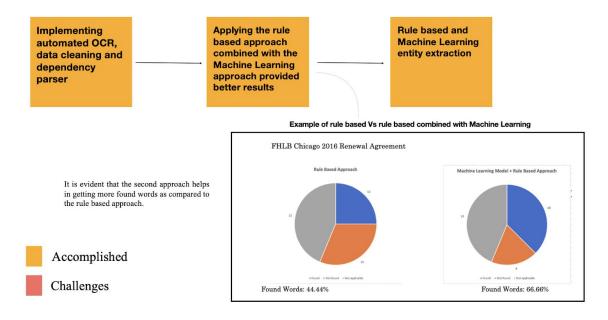


Figure 5: 4th Phase

This phase was more focused on implementing an automated OCR system. After implementing that, there were two approaches to provide the output: a) Rule based; b) Rule based combined with Machine Learning entity extraction. After applying both the approaches, the results were better in the second approach (as mentioned in the figure).

# **Overview of the Categorization List**

## **Initial Stage**

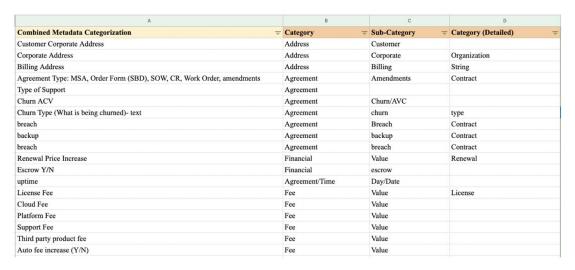


Figure 6: Categorization List (Initial Stage)

# After improvising

Category Sub-Category		Metadata_Original	Metadata	Term1	Term2	Term3	
Value		99	99	99			
Value		99.8	99.8	99			
Value		99.9	99.9	99			
Time	Hours	99.99	99.99	99			
Time	Days	12 hours	12 hours	hours	12		
Time	Hours/days	15 days	15 days	days	15		
Time	Hours/days	24 hours	24 hours	hours	24		
Γime	Hours/Days	30 days	30 days	days	30		
Time	Hours/Days	48 hours	48 hours	hours	48		
Person	Name	Account Owner	Account Owner	owner	account		
Agreement		Agreement Type	Master Service Agreement	Master Service Agreement			
Agreement		Agreement Type	Order Form	Order Form			
Agreement		Agreement Type	Statement of Work	Statement of Work			
Agreement		Agreement Type	Cost Reimbursable	Cost Reimbursable			
Agreement		Agreement Type	Amendment	Amendment			
Agreement		Agreement Type	Work Order	Work Order			
Financial	Value	Annual Renewal Value	Annual Renewal Value	value	renewal	annual	
Fee	Value	Auto fee increase (Y/N)	Auto fee increase	fee	automatically		
Γime	String	Auto renew (Y/N)	Auto renew	renewal	automatically		
Agreement	backup	backup	backup	backup			
Address	Billing	Billing Address	Billing Address	address	billing		
Agreement	Breach	breach	breach	breach			
Agreement	Churn/AVC	Churn ACV	Churn ACV	churn	acv		
Γime	Date	Churn Effective Date	Churn Effective Date	date	churn	effective	
Γime	Date	Churn Notification Date	Churn Notification Date	date	churn	notification	
Agreement	churn	Churn Type (What is being churned)- text	Churn Type	churn	type		
Person	Name/ Yes or No	Client Consent required to assign - Y/N	Client Consent	client	consent		
Person	Name	Client Name	Client Name	client	name		
Product	SKU	Cloud Application SKU	Cloud Application SKU	sku	cloud	application	
Fee	Value	Cloud Fee	Cloud Fee	fee	cloud		
Legal		Contract Term- 1 year, multiple years, optional years	Contract Term	contract	term		
Legal	contract	Contractual Relationship	End User License Agreement	End User License Agreement			
Legal	contract	Contractual Relationship	Cancellation Letter	Cancellation Letter			

Figure 7: Categorization List (After Improving)

# Sample work of the output

#### **Initial Stage**

section	category	keyword_synonym	original_keyword	context
Services Agreement	Time	term	Term	herein and company desires toprovide such services to bank, capitalized terms not otherwise defined herein shall have the meaningsgiven to such terms in the license a
1.1 Services. Subject to the terms and conditions of this Agreement, Company agree	Time	term	Term	be prepared agreed up anand signed by both parties during the term of this agreement. a statement of work shall be insubstantially the form attached hereto as exhibit
1.2 Fees; Expenses; Rates. Bank will pay Company the fees, rates, compensation and	Agreement	amendment	Amendment	change in scope of services is acceptable to bank an amendment to the sowwill be prepared by company and signed by both parties. 1.4 estimates. unless otherwise spi
1.2 Fees; Expenses; Rates. Bank will pay Company the fees, rates, compensation and	Agreement	sow	sow	basis according to the rates set forth in the applicable sow. exceptas otherwise agreed to in a sow bank will reimburse company for all reasonable expenses incurredby
SOW. This is Company's good-faith estimate of the amount the Services will cost, but	Agreement	sow	sow	report of all activities being performed as specified in each sow. this status report willinclude but is not limited to reporting on a percentage of actual work performed
1.6 Staffing. Company will provide fully qualified personnel ("Resources") with a var-	Agreement	termination	Termination	(which shall not be required in the case of the termination of employment withcompany whether voluntary or involuntary of any particular resource) company may r
1.6 Staffing. Company will provide fully qualified personnel ("Resources") with a var-	Time	term	Term	(which shall not be required in the case of the termination of employment withcompany whether voluntary or involuntary of any particular resource) company may r
1.6 Staffing. Company will provide fully qualified personnel ("Resources") with a var-	Agreement	sow	sow	varying range of experience to perform services for bank under the sows. resources shall be assigned to perform services appropriate to their individual abilities. bank m
Proprietary Rights	Time	term	Term	item(s) and/or third-party product will continue to be subject to the terms and conditions of this agreement except as necessary to be consistent with the ownership ri-
Proprietary Rights	Names	owner	Owner	of this agreement except as necessary to be consistent withthe ownership rights granted herein, for purposes of this agreement "deliverable" means all right title and i
(a) Bank acknowledges that from time to time Company develops certain reusable co	Agreement	amendment	Amendment	applicable sow, nevertheless the parties agree that company may request amendment of a sow if a gs deliverable is identified following commencement of work under:
(a) Bank acknowledges that from time to time Company develops certain reusable co	Agreement	sow	sow	parties, company shall identify any gs deliverables in the applicable sow.company agrees to use reasonable efforts to identify such gs deliverables prior to the execution
2.2 Company Proprietary Items. Company's processes, including but not limited to	Agreement	sow	sow	in the course of its performance under this agreement or any sows and that is (i) not a deliverable (ii) developed independently of this agreement including pre-existing i
Delivery; Warranty	Agreement	sow	sow	the deliverables will be made as specified in the applicable sow. (a) all software work products provided to bank will be deemed accepted upon installation of the delive
3.4 No Warranty. Company does not provide any warranty or representation whatso	Agreement	sow	sow	responsibilities specified in the "bank responsibilities" section of the applicable sow as well as those specified in sections 5.3 and 5.5 below, the sow may also contain a
5.2 Company Responsibilities. Company will perform those tasks and fulfill those re-	Agreement	sow	sow	tasks and fulfill those responsibilitiesspecified in this agreement and/or any sow. company understands that its performance of servicesshall be completed in a timely n
5.4 Cooperation. Bank and Company mutually agree to cooperate to see that the Ser	Agreement	sow	sow	agree to cooperate to see that the services specified in the sow are successfully completed.5.5 third party vendors, bank represents and warrants to company that bank
Limitations of Liability	Legal	damages	Indemnity	to company for the applicable statement of work 6.2 no consequential damages, neither party will be liable to the other forany indirect incidental punitive or conse-
Limitations of Liability	Agreement	sow	sow	deliverables or anyother items furnished under this agreement or any sow will not exceed the compensation paid including any due and owing by the bank to compan
6.3 Independent Allocations of Risk. EACH PROVISION OF THIS AGREEMENT THAT	Legal	liability	Liability	each provision of this agreement that provides for a limitation of liability disclaimer of warranties or exclusion of damages is to allocate the risks of this agreement betw
6.3 Independent Allocations of Risk. EACH PROVISION OF THIS AGREEMENT THAT	Legal	damages	Indemnity	for a limitation of liability disclaimer of warranties or exclusion of damages is to allocate the risks of this agreement between the parties, this allocation is reflected in the
Indemnification	Legal	damages	Indemnity	fromand against any and all claims suits liabilities judgments losses damages fines costs and expenses(including reasonable attorneys' fees and expenses) ("losses") re
Confidentiality	Time	term	Term	confidential information shall also include all nonpublic personal information (as that term is defined in thegramm-leach-bliley act) and all similar categories of information
This Section will not be interpreted or construed to prohibit: (a) any use or disclosur	Agreement	termination	Termination	completion of the various sow(s) issued hereunder or as provided hereunder. 9.2 termination without cause, either party may terminate this agreement at any time up
This Section will not be interpreted or construed to prohibit: (a) any use or disclosure	Time	term	Term	to injunctive relief and other equitable relief to enforce such obligations termination 9.1. term, this agreement shall remain in effect until or until
This Section will not be interpreted or construed to prohibit: (a) any use or disclosur-	Agreement	sow	sow	obligations or exercise of its rights under this agreement or any sow or any other agreement between the parties; (b) anyuse or disclosure required by applicable law pro
9.3 Termination with Cause. Either party may terminate this Agreement or any SOW	Agreement	sow	sow	with cause, either party may terminate this agreement or any sow if the other party is in material breach or default of any obligation hereunder, which breach or default
9.4 Termination for Lack of Payment. If Bank fails to perform any payment obligation	Agreement	termination	Termination	any sow(s) upon written notice. 9.5 payment of accrued fees upon termination, upon any termination of this agreement or anysow bank shall pay all of company's ung
9.4 Termination for Lack of Payment. If Bank fails to perform any payment obligation	Time	term	Term	any sow(s) upon written notice.9.5 payment of accrued fees upon termination, upon any termination of this agreement or any sow bank shall pay all of company's ung
9.4 Termination for Lack of Payment. If Bank fails to perform any payment obligation	Agreement	sow	sow	such failure bycompany company may suspend its performance of any sow(s) until payment is received orterminate this agreement and any sow(s) upon written notice
termination of the parties' respective activities under the terminated Agreement and	Agreement	termination	Termination	under the terminated sow after the effective date of the termination; (e) any and alliabilities accrued prior to the effective date of the termination will survive; and (f) the
termination of the parties' respective activities under the terminated Agreement and	Time	term	Term	suchitems or products delivered by bank to company under the terminated agreement and/or sow and deliver to bank a written certification of such return signed by an
termination of the parties' respective activities under the terminated Agreement and	Agreement	sow	sow	of the parties' respective activities under the terminated agreement and/or sow; (b)company shall promptly return to bank any and all bank proprietary items and any c
11. Miscellaneous	Agreement	sow	sow	fulfill its obligations hereunder 11.3 entire agreement, this agreement and any sow(s) including all attachments and exhibits thereto is the complete agreement between
11.8 Surviving Sections. Whenever the context requires, any commitment or obliga	Agreement	termination	Termination	any commitment or obligation provided for in this agreement shall survive termination or expiration hereof.11.9 no third-party beneficiaries. this agreement is for the
11.8 Surviving Sections. Whenever the context requires, any commitment or obliga		term	Term	any commitment or obligation provided for in this agreement shall survive termination or expiration hereof.11.9 no third-party beneficiaries, this agreement is for the
	Time	term	Term	federal home loan bank of chicago (the "bank") agree that the terms of this statement of work shall be subject to the terms of the services agreement between the part
Exhibit 1	Agreement	amendment	Amendment	the parties:deliverables may be adjusted in the form of an amendment to this statement of work duration of work assignment: 2010 to 20 estimated ba
	-		_	

Figure 8: Sample output (Initial stage)

#### **After improving**

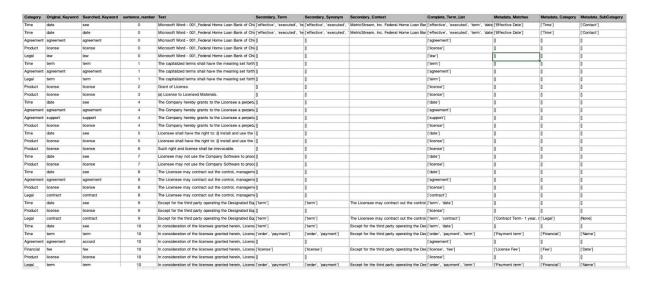
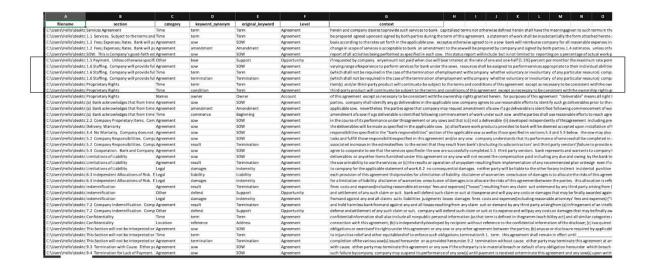


Figure 9: Sample output (After improving)

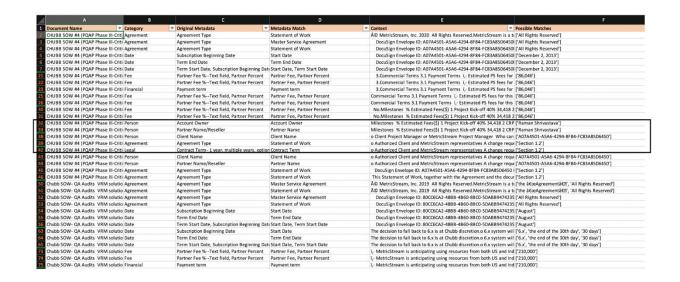
After improving, we have categorized the metadata field into several columns such as category, original key word, searched keyword, text, secondary term, secondary synonym, secondary context etc which helped in gathering the surrounding text

which the original context/synonym did not pick up. Apart from this, when this output was analyzed in Tableau, it not only helped in better understanding of the output but also gave it a cleaner look.

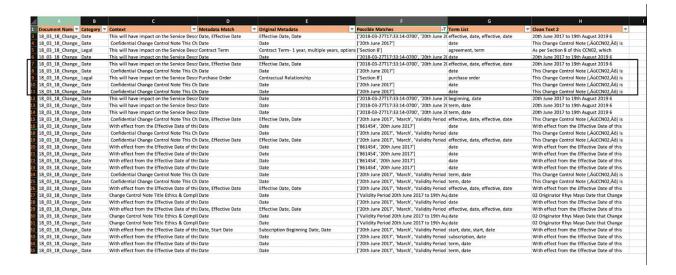
# **Key Findings:**



**Initial Stage:** In the initial stage, we were able to extract the section of the document along with its category, keyword synonym, original keyword and the context. The context section however needed to be changed and to be on point.



**Mid-Point:** In the mid-point, we focused on getting the possible matches which will help in identifying the keyword we're looking for. For example, we want to know the Account Owner of which is mentioned in the contract, rather than reading the entire context, the possible matches helped us in identifying the account owner. One can surely read the context for more information.

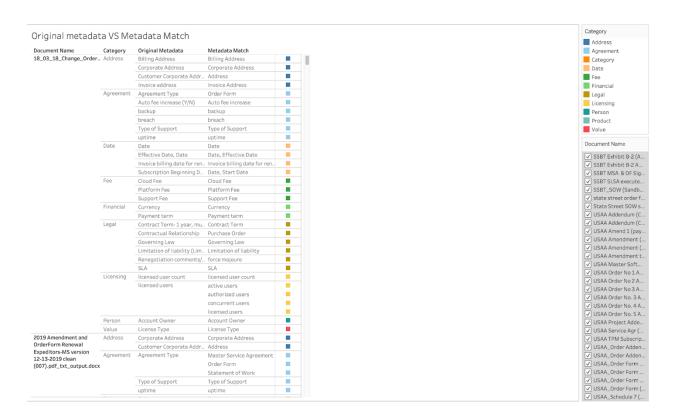


**End Point:** In the final stage, we focused on getting the outcome as precise as possible. For example, if we are looking for "effective date", the outcome will help

in identifying the possible matches (as date) and the term list. Along with the context, it also shows the clean text which helps in understanding the outcome in a better way.

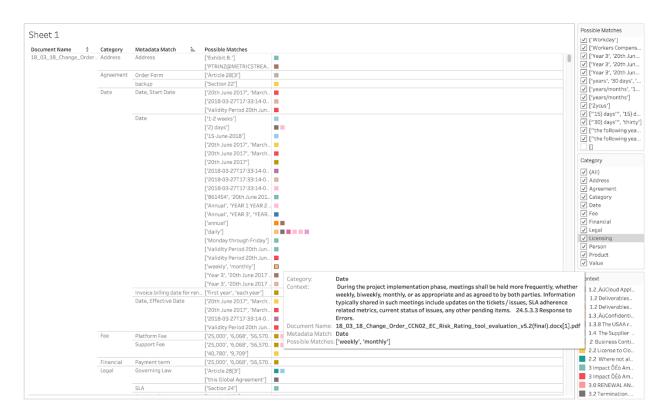
As compared to the initial stage, we're in a better position to identify the outcome. Instead of reading the entire context, we have the possible matches along with the context and the clean text.

#### **Tableau Dashboard**



This dashboard focused on the Original Metadata Vs the Metadata Match. We categorized the terminology into various categories and assigned them metadata's. This metadata was then matched by the original metadata and the outcome was as

follows which had the context, along with the metadata matches and possible matches.



The aim of the dashboard is to be able to identify the categories we're looking for according to the document/contract name and find the keyword/synonym. The dashboard helps us to select the category through the filter and find the context in the colored box for more reference.

#### **Conclusion**

Applying AI to CLM has already been occurring. In fact, AI can be used to manage a contract's life from beginning to end. From drafting a contract, obtaining approval, alerting and reminding teams of action items on an on-going basis, and more. This can be done by integrating an automatic extraction

algorithm with other systems such as a CRM system. Beyond applying automatic extraction to other types of documents (I.e. invoices, purchase orders, financial statements, etc.), this process can be pushed to the edge of computing, so that the extraction is performed locally.

#### **Future Work and Implementations**

We suggest performing these next steps in the provided order.

#### 1.) OCR Solution

The current implementation of Tesseract OCR has several faults. Depending on document quality, the output text may be not be represented in text properly. If we continue to utilize open source technology, future work must be done to improve the text output. Improving the actual optical character recognition technology itself is not a feasible goal and therefore we are limited to what we are able to accomplish with the current output.

One feature of the current tool searches for characters that are clearly incorrect and either removes or replaces them. Future work can look to implement a form of spell checking and word recognition to improve upon current text output.

#### 2.) Customization of metadata library and synonym finder

Current metadata library consists of the initial set of metadata terms provided to us from the contract desk team. An exhaustive list is yet to be

compiled. Adding new metadata fields is as easy as adding a new row to the metadata library excel file.

The terminology list and synonym finder however can be developed as well. For example, synonyms for words such as 'date' or 'uptime' may be relevant to our metadata field. However, synonyms for 'cloud' in our circumstances would not be relevant and therefore updating the synonym library to be exclusive to contract language may be an avenue worth exploring. The nltk wordnet synonym library contains a large database of synonyms and the current tool only makes use of up to 3 synonyms. In certain cases this is too many and in other cases, too few. A more sophisticated method of synonym usage may help in finding more diverse sets of text for a given metadata field.

#### 3.) Named Entity Recognition, Extraction, and the Dependency Parser

Named entity recognition applies labels to words. For example, it will look for dates and apply the tag DATE. In simple cases, this feature of the spacy library works well. However, in cases that are more specific to contracts such as 'limitation of liability' the entity recognizer has no tag capable of finding relevant terms. The contract itself must be manually labeled as training data and fed into the spacy machine learning model updater so that these terms can be automatically labeled in the future.

The ability to automatically label very specific pieces of text allow the dependency parser to answer questions directly. While the current implementation of the spacy model allows one to label monetary values and extract them, if the nearby text is unclear about what the monetary value relates to, we must manually perform that. For example, a piece of text may contain the word 'cloud fee', 'license fee', '\$10,000', '\$20,000'. The tool can find these two words and two dollar amounts but determining which word relates to which dollar amount is a manual process. Training the dependency parser using user labeled data will allow the tool to automatically relate the correct term to the correct dollar amount.