MACHINE LEARNED MODEL FOR CONTRACT GENERATION IN A DOCUMENT MANAGEMENT SYSTEM

Abstract

A document management system trains a machine learned model configured to facilitate contract generation. The document management system generates a training set for the machine learned model using user input. From a first set of agreement documents, the document management system identifies a first set of sentences relevant to clause terms input by the user. The document management system identifies, from a second set of agreement documents, a second set of sentences similar to the first set of sentences. In response to user feedback about the second set of sentences, the document management system updates the training set. The document management system retrains the machine learned model using the updated training set.

Inventors:	**Hegardh; Johan (Gothenburg, SE)**
Applicant:	**Docusign, Inc.** (San Francisco, CA)
Family ID:	**89984804**
Assignee:	**Docusign, Inc. (San Francisco, CA)**
Appl. No.:	**19/194434**
Filed:	**April 30, 2025**

Related U.S. Application Data

Publication Classification

U.S. Cl.:

CPC   **G06F16/93** (20190101); **G06Q50/18** (20130101);

Background/Summary

CROSS-REFERENCE TO RELATED APPLICATIONS [0001] This application is a continuation of and claims the benefit of priority to U.S. application Ser. No. 18/101,286, filed Jan. 25, 2023, the disclosure of which is incorporated herein by reference in its entirety.

TECHNICAL FIELD
[0002] The disclosure generally relates to the field of document management, and specifically to training a machine learned model for contract generation in document management systems.
BACKGROUND
[0003] Online document management systems can be used to create and review documents and may provide users with tools to edit, view, and execute the documents. Conventional document management systems require users to manually create documents. There is a need to provide users with improved and efficient document creation processes.
SUMMARY
[0004] A document management system uses machine learning to help improve the contract generation process.
[0005] The document management system identifies a first set of agreement documents associated with a received set of clause terms. The document management system identifies a first set of sentences within the first set of agreement documents that include a threshold number of the received set of clause terms. The document management system generates a training language set based on the identified first set of sentences and trains a machine learned language model using the training language set. From a second set of agreement documents associated with a subject, the document management system identifies a second set of sentences using the machine learned language model. The document management system presents, on a user interface, each of the second set of sentences to a user. The document management system updates the training language set by receiving a selection of a presented sentence from the second set of sentences, adding the selected sentence to the training language set, and replacing the selected sentence presented within the user interface with another sentence from the same agreement document. The document management system then retrains the machine learned language model using the updated training language set.

Description

BRIEF DESCRIPTION OF DRAWINGS
[0006] The disclosed embodiments have other advantages and features which will be more readily apparent from the detailed description, the appended claims, and the accompanying figures (or drawings). A brief introduction of the figures is below.
[0007] FIG. **1** is a high-level block diagram of a system environment for a document management system, in accordance with an example embodiment.
[0008] FIG. **2** is a high-level block diagram of a system architecture of the document management system, in accordance with an example embodiment.
[0009] FIG. **3** illustrates training and applying a machine learned model configured to rank predictive text suggestions for contract generation, in accordance with an example embodiment.
[0010] FIGS. **4**A-B illustrate an interface of the document management system with ranked

predictive text suggestions for contract generation, in accordance with an example embodiment.

[0011] FIG. **5** illustrates an example process for ranking predictive text suggestions for contract generation in the document management system, in accordance with an example embodiment.

[0012] FIGS. **6**A-B illustrate an interface of the document management system used to generate a training set for the machine learned model, in accordance with an example embodiment.

[0013] FIG. **7** illustrates an example process for training the machine learned model, in accordance with an example embodiment.

DETAILED DESCRIPTION

[0014] The Figures (FIGs.) and the following description relate to preferred embodiments by way of illustration only. It should be noted that from the following discussion, alternative embodiments of the structures and methods disclosed herein will be readily recognized as viable alternatives that may be employed without departing from the principles of what is claimed.

[0015] Reference will now be made in detail to several embodiments, examples of which are illustrated in the accompanying figures. It is noted that wherever practicable similar or like reference numbers may be used in the figures and may indicate similar or like functionality. A letter after a reference numeral, such as "120A," indicates that the text refers specifically to the element having that particular reference numeral. A reference numeral in the text without a following letter, such as "120," refers to any or all of the elements in the figures bearing that reference numeral.

[0016] The figures depict embodiments of the disclosed system (or method) for purposes of illustration only. One skilled in the art will readily recognize from the following description that alternative embodiments of the structures and methods illustrated herein may be employed without departing from the principles described herein.

Document Management System Overview

[0017] A document management system enables a party (e.g., individuals, organizations, etc.) to create and send documents to one or more receiving parties for negotiation, collaborative editing, electronic execution (e.g., via electronic signatures), contract fulfilment, archival, analysis, and more. For example, the document management system allows users of the party to create, edit, review, and negotiate document content with other users and other parties of the document management system.

[0018] The system environment described herein can be implemented within the document management system, a document execution system, or any type of digital transaction management platform. It should be noted that although description may be limited in certain contexts to a particular environment, this is for the purposes of simplicity only, and in practice the principles described herein can apply more broadly to the context of any digital transaction management platform. Examples can include but are not limited to online signature systems, online document creation and management systems, collaborative document and workspace systems, online workflow management systems, multi-party communication and interaction platforms, social networking systems, marketplace and financial transaction management systems, or any suitable digital transaction management platform.

[0019] Conventional systems require that users manually type out contract documents. The methods described herein use machine learning to improve the document generation experience for users of the document management system. A user provides initial text input into a contract document. The document management system uses a machine learned model to rank a set of predictive text suggestions based on a likelihood that each suggestion will be selected to complete the initial text input. The document management system presents the top ranked text suggestions to the user, who selects at least one of the text suggestions. The document management system modifies the document interface to include the selected suggestion.

[0020] The document management system generates the language set used to train the machine learned model with the user's input. The document management system finds clauses in existing documents that are contextually related to the initial text input, receives the user's feedback on

those clauses, and adds user approved contextually related clauses to the training set. The machine learned model is retrained as the training set is updated.

[0021] FIG. **1** is a high-level block diagram of a system environment **100** for a document management system **110**, in accordance with an example embodiment. The system environment **100** enables users **130**A-B to generate documents more efficiently with the document management system **110**. As illustrated in FIG. **1**, the system environment **100** includes a document management system **110**, which has a target contract document **120** and historical contract documents **125**, users **130**A-B, and corresponding client devices **140**A-B, each communicatively interconnected via a network **150**. In some embodiments, the system environment **100** includes components other than those described herein. For clarity, although FIG. **1** only shows two users **130**A-B and two client devices **140**A-B, alternate embodiments of the system environment **100** can have any number of users and client devices. For the purposes of concision, the web servers, data centers, and other components associated with an online system environment are not shown in FIG. **1**.

[0022] The document management system **110** is a computer system (or group of computer systems) for storing and managing documents for the users **130**A-B. Using the document management system **110**, users **130**A-B can collaborate to create, edit, review, store, analyze, manage, and negotiate documents, including the target contract document **120** and historical contract documents **125**. The target contract document **120** is a new document that a user (e.g., one of the users **130**A-B) seeks to generate. The user and/or document management system **110** designates characteristics of the target contract document **120**. For example, the target contract document **120** may be a specific type of contract, such as an employment agreement, purchase agreement, service agreement, financial agreement, master services agreement, intellectual property licensing agreement, and so on. In other embodiments, the target contract document **120** is another type of document, such as a press release or a technical specification. The document management system **110** stores one or more historical contract documents **125**. These historical contract documents **125** may be specific to the users **130**A-B or an entity associated with the document management system **110**.

[0023] The document management system **110** can be a server, server group or cluster (including remote servers), or another suitable computing device or system of devices. In some implementations, the document management system **110** can communicate with client devices **140**A-B over the network **150** to receive instructions and send documents (or other information) for viewing on client devices **140**A-B. The document management system **110** can assign varying permissions to individual users **130**A-B or groups of users controlling which documents each user can interact with and what level of control the user has over the documents they have access to. The document management system **110** will be discussed in further detail with respect to FIG. **2**.

[0024] Users **130**A-B of the client devices **140**A-B can perform actions relating to documents stored within the document management system **110**. Each client device **140**A-B is a computing device capable of transmitting and/or receiving data over the network **150**. Each client device **140**A-B may be, for example, a smartphone with an operating system such as ANDROID® or APPLE® IOS®, a tablet computer, laptop computer, desktop computer, or any other type of network-enabled device from which secure documents may be accessed or otherwise interacted with. In some embodiments, the client devices **140**A-B include an application through which the users **130**A-B access the document management system **110**. The application may be a stand-alone application downloaded by the client devices **140**A-B from the document management system **110**. Alternatively, the application may be accessed by way of a browser installed on the client devices **140**A-B and instantiated from the document management system **110**. The client devices **140**A-B enables the users **130**A-B to communicate with the document management system **110**. For example, the client devices **140**A-B enables the users **130**A-B to access, review, execute, and/or analyze documents within the document management system **110** via a user interface. In some implementations, the users **130**A-B can also include AIs, bots, scripts, or other automated processes

set up to interact with the document management system **110** in some way. According to some embodiments, the users **130**A-B are associated with permissions definitions defining actions users **130**A-B can take within the document management system **110**, or on documents, templates, permissions associated with other users and/or workflows.

[0025] The network **150** transmits data within the system environment **100**. The network **150** may be a local area or wide area network using wireless or wired communication systems, such as the Internet. In some embodiments, the network **150** transmits data over a single connection (e.g., a data component of a cellular signal, or Wi-Fi, among others), or over multiple connections. The network **150** may include encryption capabilities to ensure the security of customer data. For example, encryption technologies may include secure sockets layers (SSL), transport layer security (TLS), virtual private networks (VPNs), and Internet Protocol security (IPsec), among others.

[0026] FIG. **2** is a high-level block diagram of a system architecture of the document management system **110**, in accordance with an example embodiment. To facilitate contract generation, the document management system **110** includes a database **205**, a model generator **220**, a model store **230**, a text suggestion module **240**, and a user interface module **250**. Computer components such as web servers, network interfaces, security functions, load balancers, failover servers, management and network operations consoles, and the like may not be shown so as to not obscure the details of the system architecture. The document management system **110** may contain more, fewer, or different components than those shown in FIG. **2** and the functionality of the components as described herein may be distributed differently from the description herein.

[0027] The database **205** stores information relevant to the document management system **110**. The database **205** can be implemented on a computing system local to the document management system **110**, remote or cloud-based, or using any other suitable hardware or software implementation. The data stored by the database **205** may include, but is not limited to, the target contract document **120**, the historical contract documents **125**, portions of text corresponding to the historical contract documents **125**, information about users (e.g., the users **130**A-B), information about one or more entities associated with users, client device identifiers (e.g., of the client devices **140**A-B), document clauses, document templates, version histories, and other information stored by the document management system **110**. In some embodiments, the database **205** stores metadata associated with documents or portions of documents, such as users who modified the documents, entities associated with the documents, parties who signed the documents, and so on. The document management system **110** can update information stored in database **205** as new information is received, such as new documents and feedback from users received via the user interface module **250**. Updates to machine learned models are also stored in the database **205**.

[0028] The model generator **220** trains machine learned models. To facilitate contract generation, the model generator **220** uses the historical contract documents **125** to train a machine learned model that is configured to rank predictive text suggestions based on their relevance to initial user input. Specifically, the machine learned model ranks the text suggestions based on a likelihood that each text suggestion will be selected to complete a user's initial text portion into a contract document. In some embodiments, the machine learned model ranks the text suggestions based on characteristics of the contract document. In some embodiments, the model generator **220** retrains models stored in the model store **230** periodically, or as new training data is received. Additional details about the machine learned model are provided with respect to FIG. **3**.

[0029] The model store **230** stores machine learned models for the document management system **110**, including those generated by the model generator **220**. In some embodiments, the model store **230** may store various versions of models as they are updated over time. In other embodiments, the model store **230** may store multiple versions of a type of model, for example, to apply to different document types or to other variations of available inputs. In the example herein, the model store **230** stores the machine learned model configured to rank predictive text suggestions based on initial user input to facilitate contract generation.

[0030] The text suggestion module **240** uses the trained machine learned model, stored in the model store **230**, to generate contract documents in the document management system **110**. The text suggestion module **240** receives user input of an initial text portion into a contract document. The initial text portion may be, for example, a word, a sentence, or a paragraph corresponding to a contract. The text suggestion module **240** also identifies one or more characteristics of the contract document, such as a type of the contract document, one or more entities or parties associated with the contract document, and so on.

[0031] The text suggestion module **240** generates a database of contract text portions from the historical contract documents **125** and from this database, identifies a candidate set of text portion suggestions that are relevant to the user's initial text portion. In some embodiments, the database of contract text portions is specific to the user who provided the initial text portion. In other embodiments, the database of contract text portions is specific to an entity associated with the document management system **110** or to which the user belongs. The text suggestion module **240** may use a machine learned model to identify the relevant candidate set of text portion suggestions.

[0032] The text suggestion module **240** subsequently applies the trained machine learned model to the candidate set of text portion suggestions. The trained machine learned model takes as additional input the user's initial text portion and characteristics of the contract document, generating a set of top ranked text portion suggestions. The ranking is based on a likelihood that each text portion suggestion will be selected to complete the user's initial text portion, while factoring in the characteristics of the contract document. The text suggestion module **240** presents the top ranked text portion suggestions to the user and in some embodiments, the level of risk associated with each of these suggestions. After a user selects at least one of the top ranked text portion suggestions, the text suggestion module **240** modifies the contract document with the selected text portion suggestion. In some embodiments, the user provides feedback as to the accuracy and relevance of the top ranked text portion suggestions. Based on the feedback, the text suggestion module **240** modifies the ranking of the text portion suggestions or provides the feedback back to the model generator **220** to retrain the machine learned model.

[0033] For example, the text suggestion module **240** may receive, from a user, a first sentence of an indemnity clause in a contract document designated to be a Master Services Agreement. The first sentence of the indemnity clause is the initial text portion; the type of the document (i.e., the Master Service Agreement) is a characteristic of the contract document. The text suggestion module **240** generates a candidate set of text portion suggestions relevant to the first sentence of the indemnity clause. The machine learned model ranks each of the text portion suggestions based on a likelihood that the user will select each text portion suggestion to complete the remainder of the indemnity clause in the Master Services Agreement. The text suggestion module **240** subsequently presents the top ranked text portion suggestions to the user, who selects at least one suggestion. The text suggestion module **240** modifies the contract document to include the selected suggestion, completing the indemnity clause.

[0034] The user interface module **250** generates user interfaces for users (e.g., the users **130**A-B) to interact with the document management system **110**. The user interface module **250** receives input from the user and presents text suggestions received from the text suggestion module **240** to the user. The input from the user includes the initial text portion as well as feedback as to the relevance and accuracy of the text portion suggestions. The user interface module **250** also provides a user interface for users to add, delete, or modify the contents of the target contract document **120**, the historical contract documents **125**, and other documents stored in the document management system **110**. In some embodiments, the user interface module **250** may provide a user interface that allows users to modify content such as text, images, links to outside sources of information such as databases, and the like.

Generating Predictive Text Suggestions Using Machine Learning

[0035] FIG. **3** illustrates training and applying a machine learned model **300** configured to rank

predictive text suggestions for contract generation, in accordance with an example embodiment. As described with respect to FIG. **2**, the document management system **110** receives initial text from a user, input into the target contract document **120**. The document management system **110** (e.g., via the text suggestion module **240**) identifies a candidate set of text suggestions **305** from the historical contract documents **125**. Each of the text suggestions in the candidate set of text suggestions **305** is relevant to the initial text from the user. The machine learned model **300** (e.g., trained by the model generator **220** and stored in the model store **230**) takes, as input, the set of candidate text suggestions **305** as well as characteristics **360** of the target contract document **120**. The machine learned model **300** ranks the candidate set of text suggestions **305** based on the characteristics **360** of the target contract document **120** and a likelihood that each text portion suggestion will be selected to complete the initial text.

[0036] The model generator **220** uses a training set **310** to train the machine learned model **300**. The training set **310** comprises the historical contract documents **125**, each of which includes one or more portions of text labeled as initial text **330** or completed text **340**. The initial text **330** may be a word, a phrase, a sentence, a clause, a paragraph, a heading, and so on. The completed portion **340** corresponds to and follows the initial text **330**. For example, the completed portion **340** is the remainder of the phrase, sentence, or clause that was started by the initial text **330**. Each historical contract document **125** is also associated with one or more characteristics **350**. Characteristics include, for example, a type of the contract document, one or more parties to the contract document, characteristics of one or more entities associated with the contract document, and characteristics of one or more users associated with the contract document. Examples of characteristics of an entity associated with the document include a legal type of the entity, an industry of the entity, and a jurisdiction associated with the entity. In some embodiments, the training set **310** further includes a level of risk associated with each text portion in each of the historical contract documents **125**. The document management system **110** automatically generates the training set **310** by labeling portions of the historical contract documents **125**.

[0037] The document management system **110** may generate the training set **310** with user feedback. For example, users of the document management system **110** may manually label the initial text **330**, the completed text **340**, and the characteristics **350** of each of the historical contract documents **125**. In other embodiments, the document management system **110** identifies a first set of sentences (and/or clauses) within a subset of the historical contract documents **125** that include a threshold number of the terms initially input by the user (e.g., in the initial text **330**). The document management system **110** identifies a second set of sentences (and/or clauses) from a second subset of the historical contract documents **125** that are similar to the first set of sentences. The user provides input on whether to add each of the sentences in the second set to the training set **310**. Accordingly, the training set **310** may be dynamically updated with user input. This embodiment is further described with respect to FIGS. **6-7**.

[0038] The model generator **220** uses the training set **310** to train the machine learned model **300**. The machine learned model **300** learns to draw conclusions from relationships between the data in the training set **310**. Specifically, the machine learned model **300** learns to relate initial portions of text to completed portions of text in the historical contract documents **125**. For example, the machine learned model **300** may learn that the word "intellectual" is always followed by the word "property" in agreements, or that the initial text "trade" has a higher likelihood of being completed as "trademark" than "trade secret." In another example, the machine learned model **300** may predict that an intellectual property provision in a service agreement always includes "patents, copyrights, trademarks, and trade secrets."

[0039] The machine learned model **300** learns to make connections between portions of text in the historical contract documents **125** and the documents' authors and associated entities. For example, the machine learned model **300** may learn that a high percentage of service agreements created by an entity associated with the document management system **110** include indemnity clauses. The

machine learned model **300** may learn that when an entity authors a licensing agreement for use outside of the United States, the entity prefers a licensing term of less than **5** years. In another example, the machine learned model **300** may recognize that a particular user consistently prefers agreements to be governed by the laws of California, as stated in choice of law provisions in agreements authored by the user.

[0040] The model generator **220** may use different versions of supervised or unsupervised machine learning, or another training technique to generate and update the machine learned model **300**. In some embodiments, other training techniques may be linear support vector machines (linear SVM), boosting for other algorithms (e.g., AdaBoost), neural networks, logistic regression, naïve Bayes, memory based learning, random forests, bagged trees, decision trees, boosted trees, boosted stumps, and so on. After training, the machine learned model **300** is applied to the characteristics **360** of the target contract document **120** and the candidate set of text suggestions **305**. The machine learned model **300** outputs a set of ranked text suggestions **380** that are likely to be selected by the user to complete the initial text from the user.

[0041] In determining its output, the machine learned model **300** may additionally be trained on and factor in characteristics of the user, risk levels of the text suggestions, and a type of the initial text from the user. Characteristics of the user may include a position of the user within an entity associated with the document management system **110**, types of documents the user has worked on in the past, and so on. Risk levels of each of the text suggestions may be determined based on prior litigation history associated with a particular clause or portion of text surrounding the text suggestion. In some embodiments, risk levels may be determined by how frequently the text suggestion appears in the historical contract documents **125** or based on input of a level of risk from the user and/or the entity associated with the user. In other embodiments, the greater the monetary value associated with or included in a text suggestion, the greater the level of risk. The type of the initial text from the user may impact the relevance of certain predictive text suggestions.

[0042] For example, if the initial text is a heading (e.g., a heading for a choice of law provision), the machine learned model may determine that a paragraph corresponding to the heading (e.g., the completed choice of law clause) is more likely to be selected by the user than a sentence corresponding to the heading (e.g., the first sentence of a choice of law clause).

Example Document Management System Interface

[0043] FIGS. **4**A-B illustrate an interface **400** of the document management system **110** with ranked predictive text suggestions for contract generation, in accordance with an example embodiment. A first portion of the interface **400** shows a target contract document **405**, a Master Services Agreement, with initial text from a user. The initial text portion **410** is an incomplete paragraph corresponding to an intellectual property provision of the agreement. The initial text portion **420** is an incomplete sentence corresponding to a choice of law provision of the agreement. A second portion of the interface **400** shows ranked text suggestions **430** and **440**. The ranked text suggestions **430** are suggested to complete the paragraph in the initial text portion **410**; the ranked text suggestions **440** are suggested to complete the clause in the initial text portion **420**. As shown in FIG. **4**B, the document management system **110** will modify the interface **400** to include the user's selections of the ranked text suggestions **430** and **440**, completing the initial text portions **410** and **420**, respectively. In some embodiments, the document management system **110** designates one or more provisions in the target contract document **405** as the initial text portion **450**. Accordingly, the document management system **110** presents clause suggestions **460** on the interface **400** based on their relevance to the target contract document **405**. The example herein shows that the document management system **110** suggests adding an indemnity clause, force majeure clause, or a severability clause following the initial text portion **450**, the choice of law clause, based on the likelihood that the user would select each clause suggestion to follow the initial text portion **450**. In some embodiments, the document management system **110** may suggest clauses within a particular clause category (e.g., types of indemnity clauses).

Example Document Generation Workflow

[0044] FIG. **5** illustrates an example process for ranking predictive text suggestions for contract generation in the document management system, in accordance with an example embodiment. The document management system (e.g., the document management system **110**) generates **500** a database of contract text portions from historical contract documents (e.g., the historical contract documents **125**).

[0045] The document management system generates **510** a training set (e.g., the training set **310**) of data from the historical contract documents. For each historical contract document, the training set includes one or more initial text portions (e.g., initial text **330**) within the historical contract document and for each initial text portion, a corresponding completed text portion (e.g., completed text **340**) that includes the initial text portion. Additionally, the training set includes characteristics associated with each historical contract document and characteristics of entities associated with the historical contract document (e.g., characteristics **350**).

[0046] The document management system trains **520** a machine learned model (e.g., the machine learned model **300**) using the training set of data. The machine learned model is configured to rank a set of text portion suggestions based on a likelihood that each text portion suggestion will be selected as a completed text portion for an initial text portion received in the creation of a contract document. The machine learned model's output may also be based on the characteristics of the contract document.

[0047] The document management system receives **530** target initial text from a user seeking to create a new document (e.g., the target contract document **120**). The initial text may range from one or more words to a heading. The document management system searches **540** the database of contract text portions for text portion suggestions that are relevant to the initial text. The result is a candidate set of text portion suggestions (e.g., the candidate set of text suggestions **305**).

[0048] The document management system applies **550** the machine learned model to the candidate set of text portion suggestions and to characteristics of the target contract document (e.g., characteristics **360**) to identify a set of top ranked text portion suggestions.

[0049] The document management system modifies **560** a contract creation interface (e.g., the interface **400**) to include the identified set of top-ranked text portion suggestions (e.g., ranked text suggestions **380**). After the user selects the most relevant text portion suggestion, the target contract document is modified to include the selected text portion suggestion.

Using Machine Learning to Generate a Training Language Set

[0050] FIGS. **6**A-B illustrate an interface **600** of the document management system **110** used to generate a training set for the machine learned model **300**, in accordance with an example embodiment. As described with respect to FIG. **3**, the machine learned model **300** is trained with the training set **310**. In one embodiment, the training set **310** is dynamically updated using input from a user and the output of the machine learned model **300** itself.

[0051] The document management system **110** receives a set of clause terms **610** from the user via the interface **600**. The clause terms may be part of and/or separate from the initial text portion (e.g., the initial text **330**) input by the user. For example, as illustrated in FIG. **6**A, the user may input the clause terms "force," "majeure," and "pandemic" to identify force majeure provisions in agreement documents that include the word "pandemic."

[0052] The document management system **110** searches a first set of agreement documents using the set of clause terms **610** to identify a set of relevant matches **620** (e.g., agreement documents that include a threshold number of the set of clause terms). The first set of agreement documents comprises a subset of the historical contract documents **125**. For example, the document management system **110** may identify a threshold number of relevant matches **620**. This threshold number may be set by the user, and/or based on the type and/or number of clause terms input by the user. In another example, the document management system **110** may identify relevant matches **620** from historical contract documents **125** associated with the user (e.g., who input the clause terms),

an entity associated with the user, agreement documents in the public domain, or some combination thereof.

[0053] The document management system **110** identifies a set of sentences, clauses, phrases, paragraphs, terms, or any other text portions ("sentences" hereafter) **630** that include a threshold number of the clause terms input by the user. For example, as demonstrated in FIG. **6**A, the document management system **110** may identify clauses from the first set of agreement documents that include all of the clause terms input by the user, while in other embodiments, the document management system may identify clauses that include most or just some of the inputted clause terms.

[0054] In some embodiments, the threshold number of clause terms used to identify the set of sentences **630** may be set by the user and/or an entity associated with the user. In other embodiments, the document management system **110** determines the threshold value based on the type of clause terms and/or agreement documents associated with the clause terms. In some embodiments, the document management system **110** identifies the set of sentences based on other factors. For example, the user may specify that the set of sentences must be associated with a particular entity, a monetary value, a jurisdiction, any other suitable characteristic, or some combination thereof. In this example, only sentences that are associated with (for instance) a particular entity that also include a threshold number of the inputted clause terms are included in the identified set of sentences.

[0055] The document management system **110** presents the relevant matches **620** and the identified set of sentences **630** to the user via the interface **600**. The user may provide feedback **640** on the identified set of sentences **630** via the interface **600**. For example, the user may approve or reject the identified set of sentences using corresponding interface elements. The user may also request that the document management system **110** present a new set of sentences, from either the same set of documents or from different documents that also include sentences associated with the inputted clause terms.

[0056] The document management system **110** generates a training set (e.g., the training set **310**) based on the matches from the first set of agreement documents and trains the machine learned model **300** using the generated training set. The machine learned model **300** outputs a second set of sentences and/or clauses **660** that are relevant and/or similar to the clause terms input by the user. The second set of sentences **660** may be a subset of the candidate text suggestions **305**, and the second set of sentences are identified from a second set of agreement documents associated with one or more subjects or document categories, such as legal paperwork, medical documents, medical reports, scientific papers, real estate agreements, employment agreements, court filings, business documents and/or agreements, and so on. In some embodiments, the user specifies the one or more subjects associated with the second set of agreement documents. In other embodiments, the document management system **110** selects the subjects based on the input clause terms and/or characteristics of the user.

[0057] The document management system **110** determines a measure of similarity and/or relevance between the set of sentences **630** and the second set of sentences **660**. The document management system **110** presents the second set of sentences **660**, as well as the measure of similarity **670**, to the user via the interface **600**. The user may request additional information about any agreement documents from which the second set of sentences **660** originate. The document management system **110** may accordingly present characteristics about these agreement documents on the interface **600**. For example, the document management system **110** may display a context and/or author of the agreement document, an entity associated with the agreement document, and/or an industry associated with the agreement document.

[0058] The user may provide input **680** on each of the second set of sentences **660**. The user input **680** may be positive or negative feedback about each of the sentences in the second set of sentences **660**. In response to positive feedback **680** from the user about a presented sentence, the document

management system **110** adds the presented sentence to the training set. The document management system **110** replaces the presented sentence with another sentence from the same agreement document (e.g., the agreement document from which the presented sentence originated).

[0059] In some embodiments, the document management system **110** replaces the presented sentence after receiving negative feedback **680** from the user, without adding the presented sentence to the training set. In response to presenting a threshold number of sentences from the same agreement document, or in response to receiving positive and/or negative feedback for a threshold number of sentences from the same agreement document, the document management system **110** may replace the second set of sentences **660** with sentences from another agreement document. In some embodiments, the user can manually request sentences from a different set of agreement documents. Accordingly, the training set is dynamically updated with user input identifying additional sentences to add to the training set from one or more documents. The document management system **110** the retrains the machine learned model **300** with updated training set.

[0060] FIG. **7** illustrates an example process for training the machine learned model, in accordance with an example embodiment. The document management system (e.g., the document management system **110**) identifies **710** a first set of agreement documents (e.g., a subset of the historical contract documents **125**) associated with a received set of clause terms. A user (e.g., one of the users **130**A-B) may input the clause terms to an interface (e.g., the interface **600**) of the document management system.

[0061] The document management system identifies **720** a first set of sentences (e.g., the set of sentences **630**) within the first set of agreement documents that include a threshold number of the received set of clause terms. The user may provide feedback (e.g., via user interface element **640**) on the first set of sentences, for instance either approving or rejected sentences. It should be noted that although this example is described in terms of sentences, in practice, clauses, terms, paragraphs, or any other document portion can be presented to a user for feedback and/or inclusion in a training set.

[0062] The document management system generates **730** a training language set (e.g., the training set **310**) based on the identified first set of sentences. In some embodiments, the training set includes each of the identified first set of sentences which have received positive feedback from a user, while in other embodiments, the training set includes all of the identified first set of sentences that didn't receive negative feedback from the user.

[0063] The document management system trains **740** a machine learned model (e.g., the machine learned model **300**) using the training language set. For example, the machine learned model can be a neural network trained to identify sentences or clauses within documents based on terms provided by a user, and/or based on a subject identified by a user or inferred from the terms provided by the user.

[0064] Applying the machine learned model to a document corpus, the document management system identifies **750** a second set of sentences (e.g., the candidate set of text suggestions **305**) from a second set of agreement documents within the document corpus. In some embodiments, the second set of agreement documents are each associated with a subject, for instance associated with key terms identified by a user.

[0065] The document management system presents **760** each of the plurality of the second set of sentences to the user within a corresponding portion of an interface. The user may provide feedback on each of the sentences in the second set of sentences using one or more interface elements, and can request additional sentences from a same document, or sentences from one or more different documents.

[0066] In response to receiving a selection of a presented sentence from the user, the document management system adds the selected sentence to the training language set. The document management system replaces the selected sentence from the interface with another sentence from

the same agreement document. Accordingly, based on the selections received from the user, the document management system updates **770** the training language set.

[0067] The document management system retrains **780** the machine learned model using the updated training language set. The machine learned model can be retrained each time the training data set is updated, after the passage of a threshold amount of time, periodically, or at the request of user. The updated machine learned model can then be used to identify sentences from additional documents for presenting to a user, for instance as suggestions when a user is creating an agreement document based on terms inputted by the user, as described above with regards to FIG. **5.**

Additional Configuration Considerations

[0068] The foregoing description of the embodiments has been presented for the purpose of illustration; it is not intended to be exhaustive or to limit the patent rights to the precise forms disclosed. Persons skilled in the relevant art can appreciate that many modifications and variations are possible in light of the above disclosure.

[0069] Some portions of this description describe the embodiments in terms of algorithms and symbolic representations of operations on information. These algorithmic descriptions and representations are commonly used by those skilled in the data processing arts to convey the substance of their work effectively to others skilled in the art. These operations, while described functionally, computationally, or logically, are understood to be implemented by computer programs or equivalent electrical circuits, microcode, or the like.

[0070] Furthermore, it has also proven convenient at times, to refer to these arrangements of operations as modules, without loss of generality. The described operations and their associated modules may be embodied in software, firmware, hardware, or any combinations thereof.

[0071] Any of the steps, operations, or processes described herein may be performed or implemented with one or more hardware or software modules, alone or in combination with other devices. In one embodiment, a software module is implemented with a computer program product comprising a computer-readable medium containing computer program code, which can be executed by a computer processor for performing any or all of the steps, operations, or processes described.

[0072] Embodiments may also relate to an apparatus for performing the operations herein. This apparatus may be specially constructed for the required purposes, or it may comprise a general-purpose computing device selectively activated or reconfigured by a computer program stored in the computer. Such a computer program may be stored in a non-transitory, tangible computer readable storage medium, or any type of media suitable for storing electronic instructions, which may be coupled to a computer system bus. Furthermore, any computing systems referred to in the specification may include a single processor or may be architectures employing multiple processor designs for increased computing capability.

[0073] Embodiments may also relate to a product that is produced by a computing process described herein. Such a product may comprise information resulting from a computing process, where the information is stored on a non-transitory, tangible computer readable storage medium and may include any embodiment of a computer program product or other data combination described herein.

[0074] Finally, the language used in the specification has been principally selected for readability and instructional purposes, and it may not have been selected to delineate or circumscribe the patent rights. It is therefore intended that the scope of the patent rights be limited not by this detailed description, but rather by any claims that issue on an application based hereon. Accordingly, the disclosure of the embodiments is intended to be illustrative, but not limiting, of the scope of the patent rights, which is set forth in the following claims.

Claims

1. A computer-implemented method, comprising: receiving, using at least one processor, a text for generation of an electronic document; querying, using the at least one processor, using one or more parameters associated with the text, a storage location storing one or more candidate text portion suggestions; applying, using the at least one processor, a machine-learned model to the one or more candidate text portion suggestions and the text to generate at least one document portion for the electronic document; and generating, using the at least one processor, using the machine-learned model, the electronic document using the at least one document portion.

2. The method of claim 1, wherein the one or more candidate text portion suggestions are configured to be ranked using the machine-learned model based on a likelihood that each candidate text portion suggestion in the one or more candidate text portion suggestions will be selected for generation of the at least one document portion.

3. The method of claim 2, wherein at least one candidate text portion suggestion in the one or more candidate text portion suggestions is selected for generation of the at least one document portion based on a rank associated with the at least one candidate text portion suggestion.

4. The method of claim 1, wherein the one or more parameters include at least one of: one or more characteristics of the electronic document, one or more words in the electronic document, one or more text portions in the electronic document, a type of the electronic document, one or more clause terms in the electronic document, and any combination thereof.

5. The method of claim 4, wherein the one or more characteristics of the electronic document include at least one of: a context of the electronic document, an author of the electronic document, an entity associated with the electronic document, an industry associated with the electronic document, or any combinations thereof.

6. The method of claim 1, wherein the machine-learned model has been trained using one or more historical candidate text portion suggestions selected for inclusion into one or more historical electronic documents.

7. The method of claim 6, wherein the machine-learned model is retrained based on at least one candidate text portion suggestion in the one or more candidate text portion suggestions selected for generation of the at least one document portion for the electronic document.

8. A non-transitory computer-readable storage medium storing executable instructions that, when executed by at least one processor, cause the at least one processor to: receive a text for generation of an electronic document; query using one or more parameters associated with the text, a storage location storing one or more candidate text portion suggestions; apply a machine-learned model to the one or more candidate text portion suggestions and the text to generate at least one document portion for the electronic document; and generate, using the machine-learned model, the electronic document using the at least one document portion.

9. The non-transitory computer-readable storage medium of claim 8, wherein the one or more candidate text portion suggestions are configured to be ranked using the machine-learned model based on a likelihood that each candidate text portion suggestion in the one or more candidate text portion suggestions will be selected for generation of the at least one document portion.

10. The non-transitory computer-readable storage medium of claim 9, wherein at least one candidate text portion suggestion in the one or more candidate text portion suggestions is selected for generation of the at least one document portion based on a rank associated with the at least one candidate text portion suggestion.

11. The non-transitory computer-readable storage medium of claim 8, wherein the one or more parameters include at least one of: one or more characteristics of the electronic document, one or more words in the electronic document, one or more text portions in the electronic document, a type of the electronic document, one or more clause terms in the electronic document, and any

combination thereof.

12. The non-transitory computer-readable storage medium of claim 11, wherein the one or more characteristics of the electronic document include at least one of: a context of the electronic document, an author of the electronic document, an entity associated with the electronic document, an industry associated with the electronic document, or any combinations thereof.

13. The non-transitory computer-readable storage medium of claim 8, wherein the machine-learned model has been trained using one or more historical candidate text portion suggestions selected for inclusion into one or more historical electronic documents.

14. The non-transitory computer-readable storage medium of claim 13, wherein the machine-learned model is retrained based on at least one candidate text portion suggestion in the one or more candidate text portion suggestions selected for generation of the at least one document portion for the electronic document.

15. A system, comprising: at least one processor; and a non-transitory computer-readable storage medium storing executable instructions that, when executed, cause the at least one processor to: receive a text for generation of an electronic document; query using one or more parameters associated with the text, a storage location storing one or more candidate text portion suggestions; apply a machine-learned model to the one or more candidate text portion suggestions and the text to generate at least one document portion for the electronic document; and generate, using the machine-learned model, the electronic document using the at least one document portion.

16. The system of claim 15, wherein the one or more candidate text portion suggestions are configured to be ranked using the machine-learned model based on a likelihood that each candidate text portion suggestion in the one or more candidate text portion suggestions will be selected for generation of the at least one document portion, wherein at least one candidate text portion suggestion in the one or more candidate text portion suggestions is selected for generation of the at least one document portion based on a rank associated with the at least one candidate text portion suggestion.

17. The system of claim 15, wherein the one or more parameters include at least one of: one or more characteristics of the electronic document, one or more words in the electronic document, one or more text portions in the electronic document, a type of the electronic document, one or more clause terms in the electronic document, and any combination thereof.

18. The system of claim 17, wherein the one or more characteristics of the electronic document include at least one of: a context of the electronic document, an author of the electronic document, an entity associated with the electronic document, an industry associated with the electronic document, or any combinations thereof.

19. The system of claim 15, wherein the machine-learned model has been trained using one or more historical candidate text portion suggestions selected for inclusion into one or more historical electronic documents.

20. The system of claim 19. wherein the machine-learned model is retrained based on at least one candidate text portion suggestion in the one or more candidate text portion suggestions selected for generation of the at least one document portion for the electronic document.