Predict Fraud from Financial Text by Machine Learning

According to the survey by Ernst & Young (2010), the fraudulent activity has increased in the post financial crisis year. Therefore, auditors, regulators, and investors are responsible for detecting fraud by choosing both cost-effective and high-quality tools. Studies have shown that publicly listed companies' textual disclosures contain significant information about economic activities and intangibles related to unethical corporate behaviors. There are some scholars conducting several methods to detect the fraud by the language used in the management discussion and analysis. Thus, this project is going to explore the feasibility and accuracy of detecting fraud via several machine learning algorithms. Machine learning method can be a cost-effective fraud-detection tool for professionals in the financial industry. By comparing that effectiveness of our method to alternative fraud detection approaches across our data, we want to figure out the best predicted method to perform the fraud.

Data Source:

The text data is from the firm's annual 10-K filing which is public online, and we will mainly focus on the discussion part, management discussion and analysis (MD&A).

UNITED STATES SECURITIES AND EXCHANGE COMMISSION Washington, D.C. 20549	
FORM 10-K	
☑ ANNUAL REPORT PURSUANT TO SECTION 13 OR 15(d) OF THE SECURITIES EXCHANGE ACT OF 193 For the fiscal year ended September 30, 2017 or	i 4
☐ TRANSITION REPORT PURSUANT TO SECTION 13 OR 15(d) OF THE SECURITIES EXCHANGE ACT OF 19	934
For the transition period from to Commission File Number: 001-36743	
Apple Inc.	
(Exact name of Registrant as specified in its charter)	

Example: 10-K filing of Apple Inc.

Item 7. Management's Discussion and Analysis of Financial Condition and Results of Operation

This section and other parts of this Annual Report on Form 10-K ("Form 10-K") contain forward-looking statements, within the meaning of the "Private Securities Litigation Reform Act of 1995, that involve risks and uncertainties. Forward-looking statements are not statement that does not historical or current fact. Forward-looking statements can also be identified by words such as "future," rankcastes, "Pelevise," restmates," expects, "inlends," plans," "predicts," "will," "would," "could," "can," "may," and similar terms. Forward-looking statements are not guerantees of future performance and the Company's actual results may differ significantly from the results discussed in the forward-looking statements. Factors that might cause such differences include in a not limited to, those discussed in 18-fail, flem 1.0 of this Form 10-K. Under the heading "Risk-Factors," which are incorporated herein by reference. The following discussion should be read in conjunction with the consolidated financial statements and notes thereto included in Part II, Item 8 of this Form 10-K. All information presented herein is based on the Company's fiscal years. Teach of exhibition in Section Part of the section of the sectio

Overview and Highlights

The Company designs, manufactures and markets mobile communication and media devices and personal computers, and sells a variety of related software, services, accessories, networking solutions and third-party digital content and applications. The Company's products and services include Prihone, IPad, Mac, Apple Watch, Apple Wat

Fiscal Period

The Company's Sissal year is the 52 or 53-week period that ends on the last Saturday of September The Company's Sissal year of the second of t

Fiscal 2017 Highlights

Net sales increased 6% or \$13.6 billion during 2017 compared to 2016, primarily driven by growth in Sarvices, iPhone and Mac. The year-over-year increase in net sales reflected growth in each of the geographic operating segments, with the exception of Greater China. The weakness in foreign currencies relative to the U.S. dollar had an unfavorable impact on net sales during 2017 compared to 2016. In May 2017, the Company announced an increase to its capital return program to raising the expected total size of the program from \$250 billion to \$300 billion through March 2019. This included increasing its share repurchase authorization from \$157 billion to \$210 billion and raising its quarterly dividend from \$6.57 to \$0.63 per share beginning in May 2017. During 2017, the Company spent \$33.0 billion to repurchase shares of its common stock and paid dividend equivalents of \$12.8 billion. Additionally, the Company issued \$24.0 billion of \$2.5 billion of Canadidan follor-denominated term debt and contained between the during 2017.

Fiscal 2016 Highlights

Net sales declined 8% or \$18.1 billion during 2016 compared to 2015, primarily driven by a year-over-year decrease in Private 1 billion expenses in most foreign currencies relative to the U.S. dollar, partially offset by an increase sees seen to be program from \$2.00 billion to \$2.50 billion through Almach 2018. This included increasing its same program by raising the expense in April 2016, the Company point and raising its quarterly dividend from \$0.52 to \$0.57 per share beginning in May 2016. During 2016, the Company spent \$2.90 billion to 10.52 billion of 10.52 billion of \$1.00 billion to \$2.50 billion to \$1.00 billion to \$1

Example: MD&A of Apple Inc.

Since 1982, the U.S. Securities and Exchange Commission has issued Accounting and Auditing Enforcement Releases (AAERs) during or at the end of an investigation against a company, an auditor, or an officer for alleged accounting and/or auditing misconduct. Based on the AAERs, we can obtain our label: 1 - fraud; 0 - not fraud.

Specifically, our project is a classification problem and we need to build a classifier to tell labels.

Method

- 1. Data Clean: such as merging the information from AAERs and MD&A; removing punctuations and other basic work for next step
- 2. Data exploratory Analysis: create word cloud; plot hist gram of the number of sentences in each company's AAERs
 - 3. Word2Vect: convert the text information into numerical information
- 4. Modeling: Apply the dataset to several machine learning method, such as SVM, Neural Network, Random Forrest etc.
 - 5. Comparison of the machine learning algorithms by decision matrix