

Machine Learning Algorithms for Automatic English Essay Scoring

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***Abstract:** Due to the advancement of natural technology for language processing (NLP) generation as well as device learning, the task for English electronic notation (AES) research has become clearer, while the research issues are posed because of the interconnected constraints of research methodologies and the annotated information. What is the best way to build a comprehensive solid and reliable scoring system is now the main goal of modern research. In this article we have developed the English AES machine, and verified its effectiveness RF in the English scoring model by analyzing the predictive effect of RF on textual and non-textual characteristics, then we evaluated with the Pearson correlation coefficients (PCC) of RF (RF), GBD and XG Boost The study found that the general effectiveness for the RF set of rules is better than those of the two other scoring methods for composition.*

KEY WORDS- machine learning, random forest algorithm, automatic scoring of English essays, Pears on correlation coefficient

I. INTRODUCTION

In English classes, the teacher typically teaches several lessons simultaneously which includes many college students which means that if every student has a challenge in grading it will cause the teacher's work load to increase, as will it is

expected that the variety of reclassifications, revisions and revisions are expected to grow. Teachers' work load. Automated grading can be quick and eco-friendly, which can significantly decrease the time spent by teachers in grading their essays. It allows teachers to devote

more time in giving instruction to students and English novices to improve their writing skills in English with a method which affects their grades.

The research into automated grading of English essays has produced amazing results. As an example, some researchers successfully designed the PEG device, which is the first essay grader computerized. PEG believes that the overall good of an essay could be considered in a variety of basic linguistic tasks, such as comprise essay length, sentence length, preposition count as well as the number of pronouns and on.

The trial intensity function is later extracted and utilized to predict trial fluency and vocabulary, sentence structure complexity, other such things. After that, regression coefficients are calculated by employing a multiple regression technique to identify the regression coefficients as well as composition score. It is predicted using regression calculations, which do utilize techniques for natural language processing. Examine the content and the bankruptcy structure of the composition. Also not forget about the theme of the composition. . . . Certain

researchers incorporate the composition's name along with the content materials of the composition in the model of neural communities, so that it learns how to relate the content of the composition in relation to the theme of the composition.

There are a variety of different forms of AES for English and each has been successful in their scoring. In this essay we begin by introducing the methodologies that are associated with AES and suggest the evaluation method for English grade. We then develop the scope of an English AES machine, incorporating four modules that are useful and an apparatus that gains knowledge of predictions models, and eventually or later verify the efficiency that the RF model to rate predictability in the ASAP review the set. Also, review of the data for degree 4 and the crucial net dissertation data.

II LITERATURE REVIEW

1) Word advice for English work of art using big quantity data dispensation.

AUTHORS: Keon-Myung Lee, Chan Sik Han, Kwang-II Kim, Sang Ho Lee

Making technical and essay-length documents are a challenge for many people who are not native speakers. Ideas and positive statistics are essential to writing. However, well-written and concise phrases that convey the meaning of these thoughts to readers are essential for writing effectively. A lot of writers find it difficult to choose the right phrases for their writing. The most appropriate words could be commonly utilized words, which are used like they belong in the same context. It is possible to determine this through an analysis of the statistics in the corpus, which comprises an enormous variety of sentences. This paper offers a strategy which will endorse the appropriate words that are primarily based upon syntactic inquiries which are expressed as phrase combination and parts-of-speech (POS) indicators, as well as contemporary phrases. These are composed of "1:1. 2". . . . Notion.' The approach proposed can suggest potential phrases to be used as queries POS tags, along with their growing popularity, and also instances sentences from the corpus. In order to facilitate query processing it first applies POS mark-up on all sentences

within the corpus. For punctuated sentences it creates grams from 2 to 5 grams that are composed of words that include POS tags, phrases and an image that is a distinct common phrase "*". This is then a reversed form of records, like an archive that contains relevant information for every capacity word in the grams n-grams. Because of the huge number of sentences and phrases Map Reduce algorithms are utilized to analyze the approach and H Base can be used to deal with the data structure that is inverted as an actual document.

2) Automatic scoring of Arabic essay over three linguistic levels

AUTHORS: Waleed Alsanie, Mohamed I. Alkanhal, Mohammed Alhamadi, Abdulaziz O. Al-Qabbany

The significance of open-ended questions that require logical responses for college students to assess their abilities, and the increasing number of students who are applying to college, has led to the need the creation of systems that grade regularly writing essays. A better system will result in the most demanding of circumstances. First, most of the time the scoring of loose-reactions is undefined and subjective. Second, scoring free-shape

response requires thorough knowledge of the spoken language. This paper will provide an automated grading device to Arabic that takes these two issues into study. The essays we recall are primarily from Arabic as a second language college student at the intermediate and beginner levels. Essays from higher-degree College students because of that they'll be an implication of problems that require more knowledge of the languages. Essays are graded by creating specific elements at three level of language: lexical semantic and syntactic. Syntax-level notation is principally dependent on the shape of sentences. Every degree is evaluated independently. The score of the final essay is the sum of the scores. The authors present remarkable experiments using mixed methods that are nonlinear and linear with a collection of data. Our tests show that models trained using a human rather can achieve accurate and weighted quadratic coefficients consistent with the consensus with humans who rate. We can see from our results that, with certain practical assumption that the Arabic scoring device for selection aids can be constructed.

III System Analysis

EXISTING SYSTEM:

Automated English scoring systems for essays mostly based on the gadget's understanding of involves gathering and processing a set of information of scores and essays using the help of human. Aspects like word count numbers sentences length, phrase count, and word spacing are recorded within the memory. The models that machine learning learn from as well as advanced models, or linear regression made up of neural networks have been trained to anticipate the scores of essays to be based predominantly upon these features. The performance of the model is assessed as delicate and tense, and then used for scoring assays using computers additionally, with interpretability capabilities. Continuous improvement involves collecting more data, and keeping up-to-date with the advancements with NLP as well as system mastering, which provides greater quality.

PROPOSED SYSTEM:

The suggested machine to automate the scoring English

essays, based on the device mastering involves series and processing of various data sets composed of essays. It also includes the extraction of the structural and linguistic abilities as well as the selection and training of the right system while that can be trained to understand the model and the development of an interpreter for user feedback. The process should include development of a scoring grid that is user-friendly, interface, the ability to scale and continuous advancement. Security, ethical concerns in compliance, customer assistance and continuous protection are essential aspects in designing and functioning. The reliability of the device is dependent on the accuracy of information provided as well as the decision to select the model, and the layout of the presumed functions, thus you can provide reliable and reliable tests.

IV SOFTWARE REQUIREMENTS SPECIFICATION

Purpose

Through the advancement of natural processing of language (NLP) generation as well as technology-based research, the scope in English

automatic notation (AES) research will be clearer. Studies that require a lot of effort are likely to be created because of the interconnectedness of research methods and the annotated data. . . . The process of creating a comprehensive reliable scoring system is an important task in the days' research. In this article we designed the English AES device that was tested the efficacy of RF for the English scoring model, by means by analyzing the effects on the impact of RF prediction on the non-textual capability of content and on textual capabilities.

Scope

The computer-generated English scoring rules examines the written English by analyzing various factors that affect the English language, including usage of the language, grammar as well as coherence and readability. Through the study of these aspects an algorithm gives an assessment score to the written text that indicates its standard of first class as well as its proficiency within its English language. The rules are designed to provide powerful and clear assessment, aid in language acquisition assessment of writing

capabilities, as well as automated systems for grading.

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1. <https://cs229.stanford.edu/proj2013/SongZhaoAutomatedEssayScoring>
2. <https://ieeexplore.ieee.org/iel7/10099473/10099480/10099945.pdf>

OVERALL DESCRIPTION

Product Perspective

It functions in a way that it can be used as a stand-by me gadget or it can be integrated in existing educational platforms for becoming familiar with management structures or evaluating software. The algorithm improves the existing scoring strategies for essays through providing brief, constant and objective examinations of responses to written questions. The goal is to improve educators' work and makes the assessment system simpler, which will allow educators to provide immediate feedback to college students.

Product Features

The main characteristics of this system are:
Accuracy of Scoring: Uses modern methods of machine learning to assess the quality in English essay writing, based on

the syntax, vocabulary, and uniformity, as well as general ability to write.

Efficiency: Allows for quick and automatic grade-making. This significantly cuts down on the time it takes to complete manual tests and is ensuring the highest quality and accuracy.

User Classes and Characteristics

User classes

Students are individuals who compose essays to be evaluated as well as grades.

Teacher's Instructional teachers, instructors as well as educational institutions using an algorithm that automatism the evaluation of essays.

Administrators: The people who make the decisions who are responsible for the use and implementation of algorithms within educational institutions.

User Characteristics

Student's different proficiency levels in English written.

The student must be able to provide honest and reliable reviews of the essays.

Teachers need an effective tool for grading the vast amount of essays. Prefer a customizable scoring system that will match the standards of education.

Administrators: Evaluate the privacy of information and make sure that it is in compliance to Privacy regulations. Seek verification of efficacy of the algorithm for improving learning outcomes.

Operating Environment

The device is installed through an Internet server with the following configurations:

1. Operating System: Linux, Windows, or Mac OS
2. Web Server: Apache or Nginx
3. Database: Postgre SQL or MySQL

V DATA SET DESCRIPTION

❖ Essay-id: An essay is a written composition that offers and argues a particular topic, idea, or angle, commonly prepared into an advent, frame paragraphs, and a end.

❖ Essay Set: An essay set is a group of essays written by using humans on the same subject matter or spark off, which can be used as input.

❖ Essay: An essay is a dependent text that provides thoughts or arguments on a specific topic, typically such as an introduction, frame paragraphs, and a end, that's used as an opening.

❖ Endpoint: A numerical representation of the excellent or mastery of essay

content and language use, based totally on device studying fashions based on annotated essay statistics.

❖ Clean essay: “Clean essay” inside the context of refers to an essay this is properly established, coherent, grammatically correct, and free from spelling errors and irrelevant content material.

❖ Marking: Marking refers to a quantitative degree of the full number of marks, along with spaces and punctuation, in a given English essay record.

❖ Word depend: The total wide variety of words in a given text or essay, that is used as a component in automated English essay scoring algorithms to assess textual content duration and complexity.

❖ Sent-matter: Sent-remember is a feature utilized in automated English essay scoring algorithms, which represents the full number of sentences in an essay.

❖ Word Average: The definition of “word average” inside the automatic English essay scoring set of rules usually manner the score of essays primarily based on their average complexity.

❖ Number of spelling errors: This metric shows the number of spelling errors found in an essay.

❖ Number of nouns: Number of nouns refers back to the counting of the number of nouns gift inside the textual content of

an essay as a measure of complexity and linguistic structure.

◊ Adj-be counted: Refers to the quantity of adjectives utilized in an English essay, which can be used as a feature of computerized scoring algorithms aimed toward assessing the richness and descriptive great of the textual content.

◊ Number of verbs: It represents the full variety of verbs within the essay textual content.

Size ID datasets = (633, 15)

essay_id	essay_text	final_score	clear_sentence_count	clear_sentence_avg	word_count	avg_word_count	avg_sentence_count	verb_count	adj_count
1	Dear father	1765	413	17.273077	21	98	84	20	19
2	Dear father	1441	344	16.020352	21	76	75	18	24
3	Dear local	2084	490	20.4612245	31	142	96	39	29
4	Dear local	2032	460	19.4342663	31	130	95	35	26
5	Dear think	956	238	14.0180937	11	46	48	10	17
6	Dear think	2386	500	23.8272	38	120	79	21	28
7	Dear people	2094	499	20.0262028	31	124	112	32	28
8	Dear people	1662	447	16.0261	31	100	87	21	21
9	Dear technology	2024	512	20.393225	23	102	111	42	26
10	Dear local	1310	320	13.79062	24	60	70	20	20
11	Dear local	1706	393	15.4300668	35	81	86	39	29
12	Dear local	790	203	6.5497476	24	42	42	16	8
13	Dear three	1286	307	10.809251	40	81	57	15	12
14	Dear three	790	176	6.8499254	35	52	35	15	8
15	Dear the	2479	523	24.4709538	35	157	102	39	23
16	Dear local	1406	345	10.001294	31	66	74	19	23
17	Dear local	1528	372	11.4072389	7	76	77	24	28
18	Dear local	291	66	7.4008266	33	20	14	10	10

essay_id	essay_text	final_score	clear_sentence_count	clear_sentence_avg	word_count	avg_word_count	avg_sentence_count	verb_count	adj_count
22	Dear daily	1624	364	16.013285	36	79	79	28	38
23	Dear local	2143	508	21.139259	23	100	94	34	23
24	Dear local	2415	537	20.4327251	39	140	103	47	34
25	Dear local	1319	291	10.412646	30	74	59	19	13
26	Dear think	1465	345	10.404478	35	85	73	19	12
27	Dear computers	484	115	6.208927	33	25	24	10	8
28	Dear computers	1620	382	16.2672164	30	94	74	21	20
29	Dear local	1560	369	15.28424	20	88	73	23	22
30	Dear local	1159	282	10.470924	11	61	70	19	13
31	Dear that	1952	469	19.232689	30	103	92	38	39
32	Dear that	2043	477	20.478265	16	128	102	28	32
33	Dear that	686	162	11.245679	5	43	30	16	9
34	Dear that	1465	345	10.404478	30	94	74	21	20
35	Dear think	1786	417	15.282975	8	92	94	31	21
36	Dear that	2455	565	24.510586	32	109	69	25	21
37	Dear that	2036	483	20.211209	32	109	102	45	25
38	Dear that	1484	361	12.47026	11	77	73	23	18
39	Dear local	1883	441	18.268643	20	114	80	29	29
40	Dear local	780	190	10.252624	11	52	42	10	10

essay_id	essay_text	final_score	clear_sentence_count	clear_sentence_avg	word_count	avg_word_count	avg_sentence_count	verb_count	adj_count
43	Dear Comp	1156	281	14.113879	13	59	51	21	18
44	Dear people	1166	303	11.435067	5	65	28	18	5
45	Dear people	1216	303	12.0132013	17	81	58	19	19
46	Dear local	1200	315	12.0825214	20	82	47	22	24
47	Dear local	1297	310	11.458871	14	75	75	17	7
48	Dear think	1291	317	10.400686	35	86	54	20	27
49	Dear think	2156	554	21.009173	8	122	100	32	47
50	Dear local	1433	338	14.002024	17	80	55	24	22
51	Dear computers	186	47	4.000224	7	21	13	6	8
52	Dear local	2148	509	21.770469	11	85	123	30	46
53	Dear that	1681	237	16.02051	34	61	49	17	13
54	Dear that	1992	425	19.2611265	9	84	97	34	31
55	Dear that	1774	417	16.454266	30	112	80	29	20
56	Dear YOU	1094	238	10.066937	31	56	40	15	24
57	Dear YOU	2379	565	23.527778	30	151	89	33	24
58	Dear YOU	1337	307	13.478223	35	75	64	30	13
59	Dear YOU	1426	338	14.4372172	5	42	42	12	13
60	Dear YOU	1889	430	18.3302233	11	100	80	34	19
61	Dear you	136	48	6.4761126	3	22	14	7	5
62	Dear you	1345	305	13.4008361	35	75	68	24	28

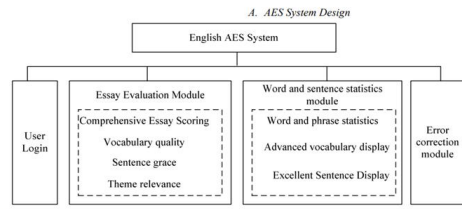


Fig. 1. Overall framework of the system

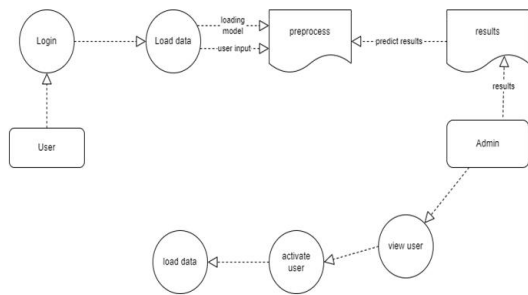
DATA FLOW DIAGRAM:

1. DFD is also known as bubble chart. It is a simple graphical formalism that can be used to symbolize the machine in terms of the input information into the machine, the diverse processing implemented to those records, and the output records carried out to that machine.

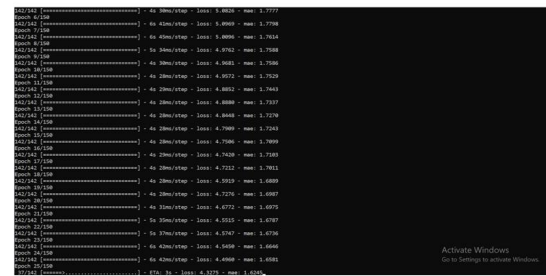
2. A data float diagram (DFD) is one of the most important modelling tools. Used to model device additives. These additives represent the device procedure. DFD indicates how facts actions through the gadget and the way it is modified through a series of differences. It is a graphical technique that suggests the go with the flow of information and alterations that occur as statistics movements from input to output.

3. DFD is also referred to as bubble table. A DFD can be used to represent a device at any level of abstraction. DFD may be divided into levels representing growing information waft and functional information.

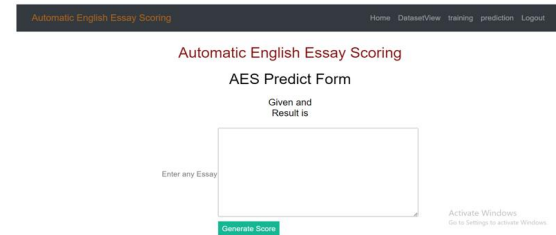
VI System Design SYSTEM ARCHITECTURE



Training:



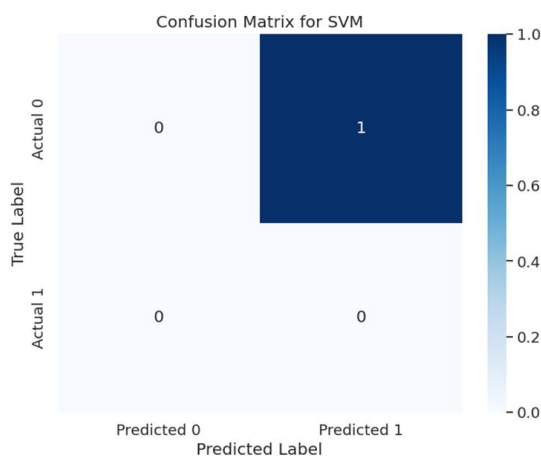
Prediction Result:



VII ACCURACY TECHNIQUES

Confusion Matrix:

The confusion matrix is a matrix used to decide performance of the category fashions for a given set of test information. It can simplest be determined if the authentic values for take a look at records are recognized. The matrix itself can be easily understood, but the related terminologies may be complicated. Since it shows the errors within the version overall performance in the shape of a matrix, therefore also referred to as an errors matrix.



VIII CONCLUSION

In traditional offline teaching, instructors examine students' English ability through extraordinary dimensions including mastery of words, grammar, lengthy and hard sentences and full-text expression capability within their English compositions, however due to the extreme mismatch between school room teaching time and the wide variety of college students within the school room, it is hard for instructors to do an extreme and cautious all-spherical take a look at on every student, and teachers' power and subjective factors also affect the judgment of college students' The trainer's energy and subjective elements also have an effect on the judgment of students' composition stage. Therefore, this paper constructs an AES gadget to recognize the efficiency

and fairness of grading, which introduces the RF set of rules and can get better machine grading results.

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