

Artificial Intelligence-Powered State-of-the-Art Visual Effects for Computer Vision in Animated Films

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Abstract: A new generation of AI-based media has emerged as a result of the introduction of AI-based technology into the film and television industries, which has had a significant impact on TV show programming and production. Artificial intelligence (AI) driven solutions are becoming more accessible, enhancing the production quality of movies and television programs. However, in recent years, the animation business has been expanding rapidly, and more and more animated features are being produced every year. It is becoming more difficult for users to find their preferred animated films amid the mountain of data associated with such films. With this context in mind, this essay will examine the use of AI and ML for groundbreaking visual effects in the animation film industry. Using computer vision and machine learning technology as a foundation, this essay proposes key concepts to support the inventive creation of cinema visual effects at the strategic level. This essay uses the informatization, intelligent development, and upgrading of the film business as a springboard to explore new methods and paradigms for creating VFX for moving images. The new visual effects of the popular Hollywood animation "Coco" are analyzed in this article by means of convolutional neural algorithms. The results of the study show that the accuracy of the test set barely changed after the parameters of the model were determined, remaining stable at around 59%; this is of critical importance for enhancing the audiovisual quality and creative standards of film works and fostering the industry's long-term, sustainable growth.

Keywords: Animation Movies, Film Visual Effects, Artificial Intelligence, Machine Learning, Computer Vision.

1. INTRODUCTION

Computer animation is the most popular form of multimedia expression right now [1]. This achievement owes a great deal to the accessibility of computers and electronic information resources. When it comes to advancements in computer animation technology, character modeling and painting are at the very top. Adding intricacy to character design is an effort to improve the presentation of animated figures and boost animation's overall impact. [2] This is done so that cartoon figures seem better. As a result, animators and academics have shifted their effort to discovering a foolproof method of painting and modeling cartoon figures. People's expectations of what they can purchase and how much they can enjoy it increase along with the economy. There are certain systemic issues mirrored in the inventiveness and aesthetics of the contemporary new media animation business. Animation's use of visual effects and color palettes as a symbolic language for conveying information is essential, yet each element of the medium has a unique impact on the final product. Performances that fuse several visual languages into a single, cohesive whole have recently seen a surge in popularity. More and more hybrid graphics will be created as digital media production and consumption capabilities grow. In recent years, the phenomena of picture special effects and graphic color mixing has been more pervasive and its area of application in the field of current visual communication design has expanded and become more widespread as a consequence of the development and fast expansion of new media. This is due of the widespread availability of and rapid improvement in new media. The preceding photos' animation was made using a wide range of basic color schemes and geometric forms. However, this approach is ubiquitous. As a result, coming up with original facial expressions of emotion becomes challenging.

However, today's visuals are a consequence of combining the two, and the results are mutually beneficial. There is a deluge of fresh visual stimuli available nowadays.

The question of how successful a film is aesthetically is one of the most perennial in the medium. It's intrinsically tied to progressive social and economic changes and the advent of innovative technology [3]. An effective and compelling illustration of the process of film conception. It is important for films to employ visual elements to express the emotions of their characters and to communicate the experiences of their characters in order to increase the worldview, life perspective, and value of cinematic works [4]. [Cameras] Maintaining the visual impact of cinematic works requires a visual performance that evolves with the times [5, 6].

2. RELATED WORK

Experts in the area agree that the term "artificial intelligence" refers to the process of "training" computer systems to "act" like humans and "complete" activities that people can only achieve when using their brains[7-19]. Please understand that this in no way defines artificial intelligence. Make sure the machine passes all the necessary tests for operation. Researchers in the academic community are making great efforts in developing more effective artificial intelligence that may be used in a wide variety of contexts. Reading [20] helps users model, while navigating the design space helps shape future modeling tools.

An animation picture innovation design model [21] based on evolutionary computation has already been built. This model was made using cutting-edge computer-generated imagery (CGI) design techniques. Component reconstruction, evolutionary design, component management, and autonomous assembly are among the primary system capabilities. It enables not just automated component creation and assembly, but also visual model change by humans. This may be done by a user. Within a limited range, it has been demonstrated that 3D model simplification based on the identification of symmetrical features may enhance model simplification efficiency [22]. There is less need for storage space and less distortion when simplifying a three-dimensional model that keeps its original features intact. The model's flexibility also allows it to be tuned to use as little resources as possible. Its practical use is rapidly becoming apparent, particularly in simulation studies involving a large number of models.

In the literature [23], the issues with developing 3dsmax expressions using the current technology are analyzed. Then, the concept of a character face expression animation system built on the 3dsmax engine is conceived. Evidenced through study of methods and techniques for giving computer-generated characters lifelike expressions. In the academic literature [24, 25], a method is provided for reducing 3D models by locating and eliminating symmetrical elements. This article will examine three methods—diversity modeling of 3D models, degree simplification, and recognizing symmetrical features—that have been shown to increase the simplification efficiency of 3D models in simulation applications. The author of literature [26] compiled a database of cartoon personalities after researching everyone in existence. The modeled characters may be made more interesting and shared via user testing and feedback.

The project at [27] combines the concepts of computer-aided design (CAD) with biological genetic evolution (BGE) to provide designers with a tool to combat their natural predisposition toward sluggish reasoning. A technique was created for face emotion animation in the literature [28] and implemented in 3dsmax. The approach was designed with the typical behavior of the muscles responsible for facial expression in mind. Research done in close contact to human people is highlighted in the literature [29] because it is useful for achieving the study's primary objective. The ability to properly reduce the number of components while maintaining their ability to perform many tasks is made possible by classification and analysis. Animation production software, as described in the existing literature [30], is a sophisticated system requiring the participation of computer scientists, artists, and producers from a wide range of fields. It is crucial to pool the knowledge of these individuals while developing such programs. This is an essential tool for every professional computer animator. The authors of the research [31] provide a novel approach to creating visual effects for 3D animation characters, as well as a CAD system for creating pictures of animated characters, that is smart and based on deep neural networks. Each of these state-of-the-art techniques uses evolutionary computing to create completely new animation pictures. All of these advancements (DNN) were very helpful to the area of animation image design. This presentation will examine the new visual effects that have lately

surfaced in animated films as a consequence of advances in AI and ML technologies. [32] This article provides fresh perspectives that may aid in the strategic development of VFX in motion pictures. The article lays out a plan of action for how computer vision and machine learning might be used to investigate alternative methods and models for animating VFX in the movie. In this article, we look at how the current informatization, intelligent development, and upgrading in the film industry might inspire new strategies for the actual use of visual effects.

3. PROPOSED METHODOLOGY

New visual effects in animated films are a primary target of the suggested technique in this section, which places a heavy emphasis on artificial intelligence and machine learning technology. This means the following technologies will emerge:

Machine Learning and the Future of Movies Visual Effects

Additionally, AI is being used to improve visual effects in films. Machine learning algorithms may be taught to parse a scene and name its constituent parts. This makes adding VFX capabilities more simpler and quicker to install. Machine learning culminates in updating the training or learning model to take use of the new information. In conclusion, generalization is a powerful method for preventing both over fitting and under fitting. Over fitting is a problem that occurs when a model can properly forecast the known data but cannot accurately predict the unknown data. The student's ability to generalize weakens as a consequence. This is currently machine learning's greatest obstacle and must be overcome. If you don't have enough information to learn from, then you don't have enough training examples. Not devoting enough time to studying is a major cause of poor academic performance. The fields of computer graphics, movie virtual asset analysis, and the development of 3D animation have all benefited greatly from the use of machine learning. In the realm of 3D animated filmmaking, for instance, the concept of reusing 3D motion data has recently been a hot issue of debate. This issue has gained considerable attention in recent years.

This is the launching pad for data-driven action. By applying techniques like subspace analysis, statistical learning, principal component analysis, and multiple body learning, mechanical learning technology can realize the reuse of 3D motion data from a massive amount of previously collected 3D motion data and expression data. The process of learning from several bodies may help us get there. You may do this by using both sets of data. Learn more about it and supervise the gathering of updated mobility data. Also essential to the growth of the 3D animation film industry is research into automating and intelligently creating action animation. It is accomplished by creating and animating the motions of digital figures and programming agents to make their own judgments. The development of a default pattern of behavior is one area where machine learning technologies may be put to use. Creating a dialogue-based virtual world and implementing a dynamic action plan are both made possible by determining the virtual character's action and consciousness mode. By opening up these lines of communication, the agent may learn how to have meaningful conversations with a broad variety of individuals quickly and securely.

The field of computer-generated film is booming, and semi-supervised learning technology is a major reason why. In semi-supervised learning, students utilize unlabeled data in an automated form rather than the actions of their peers to increase their knowledge. Different from pure semi-supervised learning is directly directed learning. The latter assumes that the anticipated data is reflected in the unlabeled samples used for assessment throughout the learning process, and that the purpose of the learning process is to maximize generalized performance with regard to the unlabeled instances. During the course of gathering and analyzing a film's digital assets, it is standard practice to acquire unlabeled samples. Getting labels produced requires a lot of work and time. Labeled and unlabeled samples work together to speed up the completion of learning tasks, improve the accuracy of data models, and cut down on labeling time. Figure 1 depicts the finalized data model for semi-supervised learning.

Humans are able to put together the vast, unfathomable universe via the three processes of perception, learning, and cognition. Topics for research, however, must adhere to the established norms and characteristics of the human species. The ineffability of the human intellect and the progress of science in the future are also factors in these concepts.

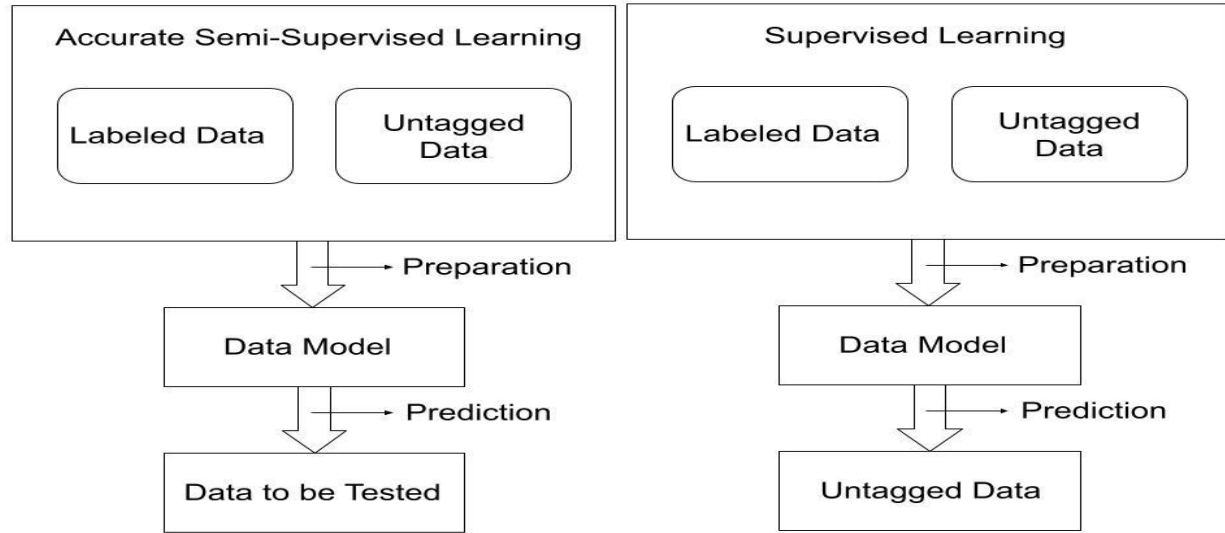


Figure 1. Architecture of semi supervised and supervised learning

4. EXPERIMENTAL STUDY ON ANIMATION FILMS

a. Modeling Process

Primary parts of modeling, such as animation scene creation, animation model development, and animation prop development, are shown in Figures 2 and 3. Modeling is the first step in the animation process. Points, lines, and evenly spaced gaps are often utilized to construct models for usage in film and television projects. There also has to be a parallel relationship between the lines. In addition, the size of the replica must be accurate to that of the original. Using the standard model's points, lines, and faces, the desired look may be achieved, enabling for the production of even extreme designs. The human or animal character model must have precise anatomical and physiological information. Appropriate balance between the upper and bottom halves of the body is crucial. The eyes, nose, ears, and mouth of a person or animal should all be in their correct anatomical places and proportions. After finishing the model, the next step is to break it down into its constituent parts.



Figure 2. The process of making animation scenes

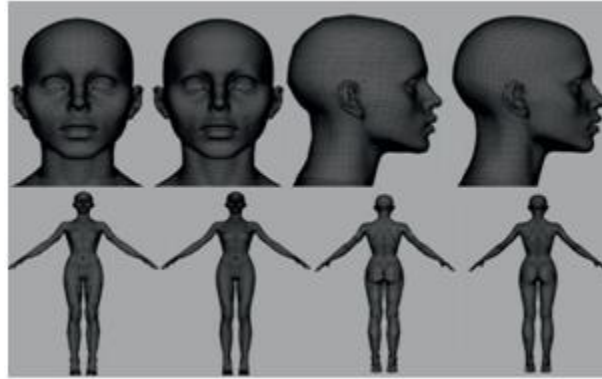


Figure 3. Modeling in animation

b. Animation

Data from the new domain of computer vision in anime is analyzed here using a convolutional neural network approach. Iteratively training the network using gradient descent to minimise the loss function and then feeding the results of the modifications back up through the layers is the basis of the training approach for convolutional neural networks. Determine the best settings for each layer's parameters to achieve a high degree of similarity between the model and the training data while maintaining a stable and well-balanced weight offset. Only by accurately modeling the training data can such parameters be acquired. In order to implement a convolutional neural network, the following steps must be taken:

Algorithm: Convolutional Neural Network

Step 1: Initialization of the Neural Networks. // how many times to practice, how fast to learn, and what value to assign to the loss function //

Step 2: Produce a random initial value for the weights and offsets by utilising a sequence of random numbers.

Step 3: Selects n input data: $X(n) = (x_1(n), x_2(n), \dots, x_k(n))$

$D(n) = (d_1(n), d_2(n), \dots, d_q(n))$ //To train the model, randomly select n input and output pairs from the training set. //

Step 4: $F_{in}(n) = \sum_{i=1}^n (p_i, w_i + b_i)$

$F_{out}(n) = \text{Max}(0, F_{in}(n))$ //You may calculate the output by adding the information from each layer of neurons until you reach the one in the end. //

Step 5: Define error function: $\text{error} = \frac{1}{2} \sum_{o=1}^q (d_o(k) - f_{out}(n))^2$

Step 6: Weights and offsets should be constantly refined through iterative learning to ensure precision.

Step 7: Employ the error function to determine if the requirements have been met. If the error value is within the allowed range, or if the number of iterative training sessions hits a threshold, the training is complete.

Step 8: To continue with the training if this is not the case, continue to Step 3.

c. Selection of Training Data

For this, we utilize the Movie lens Movielens-100 K dataset. There are 400,000 ratings for 1882 anime films in this data collection, supplied by 943 people. Contents of data set include user ID, age, profession, gender, and region, as

well as anime movie title, genre, and rating. Figure 4 depicts the collection of feature vectors that will be analyzed. The convolutional neural network can only take jobs as input if they are represented numerically. In the report, the research team chooses 21 jobs at random. Various occupational aspects and the user's interest value percentage are taken into consideration during the forward adjustment error procedure, which modifies the original amount given. This will happen after the initial value has been determined.

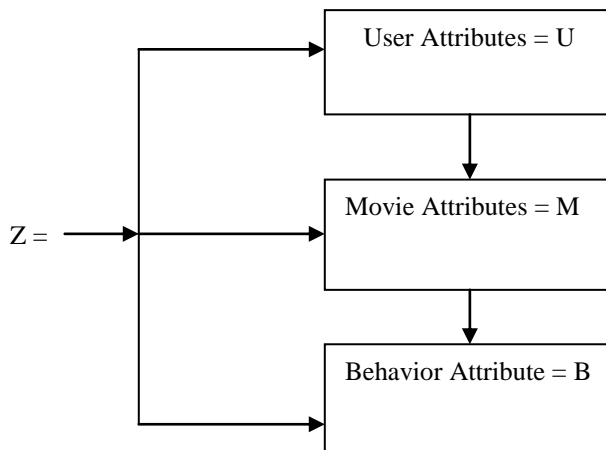


Figure 4. Information based on feature vectors

d. Verification Method of Training Data

i. K-Fold Cross-validation Method: When evaluating machine learning models on small datasets, cross-validation is a resampling approach that may help. The number of subsets into which the data will be divided is controlled by a single parameter in the procedure, denoted by "k." This method has been given the label k-fold cross-validation to reflect its popularity. In machine learning, cross-validation is used to evaluate a model's ability to correctly predict the results of trials using incomplete or missing data. Figure 5 shows that its rising popularity may be attributed to the fact that it is simple to understand, yields reliable findings even with a small data sample, and provides a judgment that is less prone to prejudice.

This technique is known as "K-fold Cross Validation" because the data sample is divided into a set number of subsamples. You may also hear the terms "fourfold cross validation" or "tenfold cross validation," both of which mean that the data set is being divided into four or ten subsamples for the purpose of analysis. k's value must be carefully considered. If k is not carefully selected, an inaccurate assessment of the model's performance may result. Depending on the data used for training, it may either drastically alter the assessed capabilities of the model or cause them to be greatly underestimated (high bias).

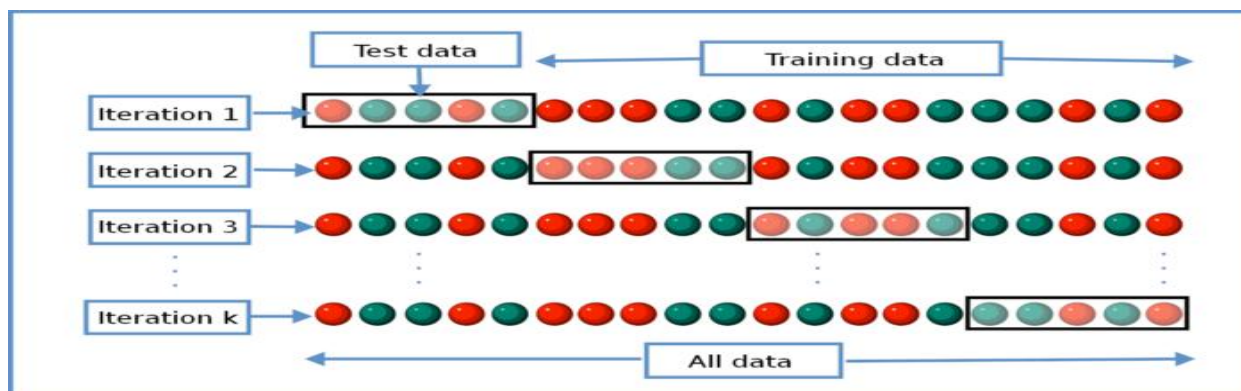


Figure 5. K-Fold Cross-Validation Method

ii. **2K-Fold Cross-Validation Method:** Based on the K-fold method, the 2K-fold cross-validation method halves each subgroup. When referring to the datasets, S0 and S1 are used for training and S1 and S2 for testing, respectively. S0 is the sample used for instruction, whereas S1 is the test sample. After running the experiment on Set S1, we will put it through its paces using Set S0 as the test set. One advantage of the 2K-fold cross-validation approach is that both the test set and the training set are sufficiently sizable to be subjected to repeated testing in line with the quantity of data. This method is far better than the others. When doing a 2K-fold cross-validation, the number 10 is often used for the parameter K. Based on their average accuracy, the five models shown in Figure 6 were selected to represent the false negative.

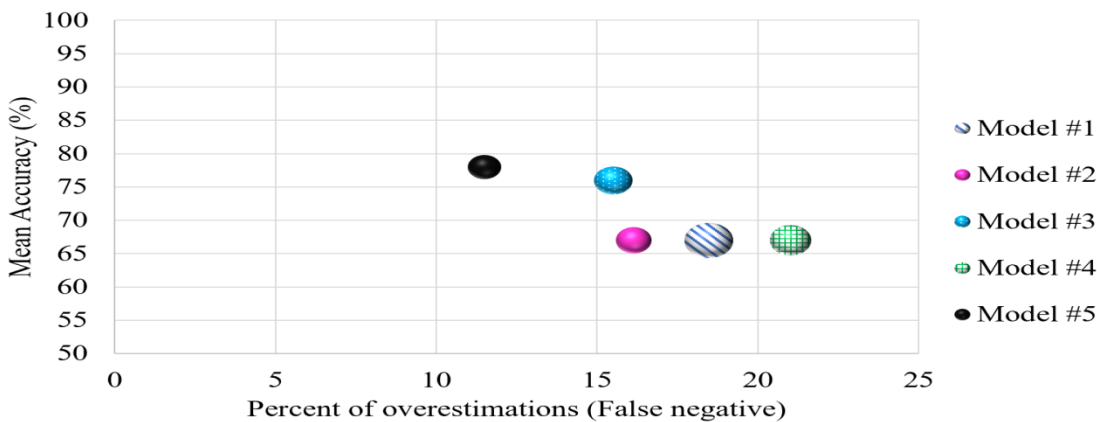


Figure 6. 2K-Fold Cross-Validation Method

5. RESULT ANALYSIS

i. **Training Analysis of Convolutional Neural Network:** Convolutional neural networks (CNNs) are a kind of deep learning model that can be programmed to learn spatial feature hierarchies automatically and adaptively, from the most fundamental to the most complex. CNN, inspired by the organization of the visual cortex in animals [13, 14], works well with information that has a grid pattern, such as photographs. This includes the number of convolution layers, the number of convolution kernels, the number of pooling layers, and the number of iterations. The stair's height is shown as well. Accuracy ratings for the various hyper parameter combinations for the convolutional neural network are shown in Table 1 and Figure 7.

| No. of Replications | Cluster Size | CNN Layers | Learning Rate | Dropout Layer | Accuracy |
|---------------------|--------------|------------|---------------|---------------|----------|
| 300 | 300 | 5 | 0.0016 | 0.60 | 38.55% |
| 300 | 400 | 5 | 0.0012 | 0.80 | 19.55% |
| 300 | 400 | 5 | 0.0012 | 0.80 | 20.56% |
| 300 | 300 | 5 | 0.0012 | 0.80 | 20% |
| 300 | 300 | 7 | 0.0016 | 0.60 | 32.54% |
| 300 | 400 | 7 | 0.0012 | 0.80 | 45.56% |
| 300 | 400 | 7 | 0.0012 | 0.80 | 43.98% |
| 300 | 400 | 7 | 0.0012 | 0.80 | 9.55% |
| 300 | 300 | 11 | 0.0016 | 0.60 | 34.55% |
| 300 | 400 | 11 | 0.0012 | 0.80 | 54.67% |
| 300 | 300 | 11 | 0.0012 | 0.80 | 18.56% |
| 300 | 400 | 11 | 0.0012 | 0.80 | 59.56% |

Table 1. The effectiveness of various hyper parameter combinations for convolutional neural networks.

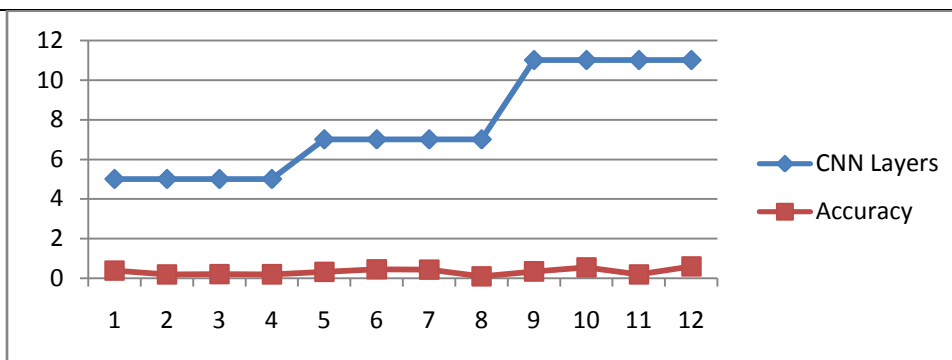


Figure 7: Accuracy graphs for various hyper parameter combinations in a partial convolutional neural network.

Using the results of the continuous training verification, an eleven-layer convolutional neural network was shown to have the highest chance of achieving the best training outcome in terms of both training time and accuracy. In light of this, this system shall henceforth be used as the yardstick by which all other systems are evaluated. This research determined the optimum size of the first layer of the convolution kernel by considering both the user data and movie data sizes. The first layer of the convolution kernel is relatively large since each record includes six valid features. Table 2 and Figure 8 provide sample values for the network parameters used to make score predictions using a convolutional neural network with eleven layers. In this experiment, we don't retrain on the same data again, no matter how many times we retrain. Instead, 300 batches are used, with 300 pieces of data for each training input and iteration. When there is too much training for a dataset that does not exist, this is known as over fitting. For each batch, testing included 400 separate data sets. The findings revealed that the percentage of correct answers to the training set grew with the number of training cycles. The accuracy of the test set has not deviated from about 59% ever since the parameters of the model were defined.

| Name of the Parameters | Parameter Values |
|--|------------------|
| Kernel size for convolution | [1, 6] |
| Amount of kernels used for convolution | 760 |
| Dropout layer ratio | 0.73 |
| Cluster size | 300 |
| Convolutional layers | 11 |
| steps | 1 |
| Number of replications | 300 |
| Activation function for a convolutional kernel | ReLU |

Table 2. Convolutional neural network setting diagram with 11 layers

a. The Experimental Evaluation of Competing Algorithms

Comparing the method discussed here with existing algorithms is crucial for making an informed decision about it. The efficiency of the score prediction system increases as its predictions get more accurate. Several ways for doing comparisons are shown below, along with a table of potential parameter values for each.

- (1) The term "Support Vector Machine" is often used to describe this kind of device. We use a radial Gaussian kernel function with a regularization parameter of $C = 200$ and a Gamma of 0.005.
- (2) When making a decision tree, entropy is used as the criterion and "best" is the setting for the Splitter.
- (3) Thirdly, by using the classification strategy of naive Bayes. Both the data dimension and the feature vector dimension sum to 6.

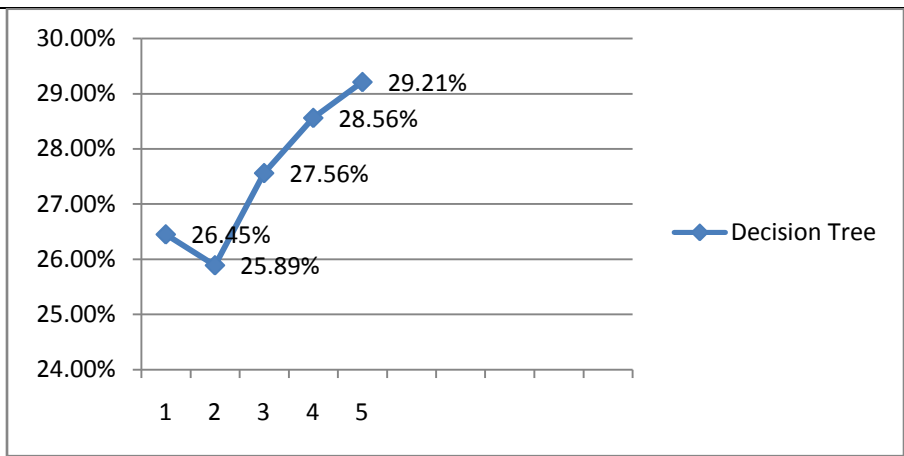


Figure 8. Relationship between number of iterations and accuracy

The outcomes of these tests, as well as those of the eleven-layer convolutional neural network, are shown in Table 3 and Figure 9, respectively. Compared to older methods and older machine learning algorithms, recent convolutional neural network-based visual expression research methodologies for animated films offer several benefits. You can see these advantages right away if you compare the two approaches. The eleven-layer convolutional neural net method described in this research also achieves better accuracy than previous methods of machine learning. Table 3 and Figure 9 display the results.

| No. of Experiments | Decision Tree | Naïve Bayes Classifier | SVM | 11 layers of CNN |
|--------------------|---------------|------------------------|--------|------------------|
| 1 | 26.45% | 29.65% | 35.76% | 56.89% |
| 2 | 25.89% | 28.97% | 34.78% | 55.97% |
| 3 | 27.56% | 31.45% | 35.73% | 57.34% |
| 4 | 28.56% | 32.13% | 36.94% | 58.63% |
| 5 | 29.21% | 34.12% | 41.76% | 59.56% |

Table 3. Variations of the accuracy of 11 layers of CNN and rest of the Machine learning algorithms

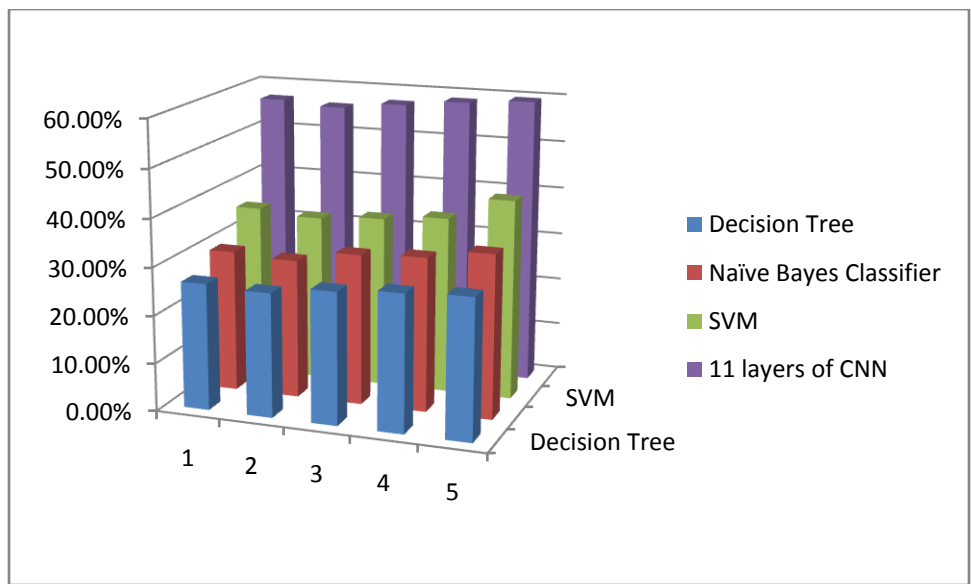


Figure 9. Comparisons of the accuracy of 11 layers of CNN and rest of the Machine learning algorithms

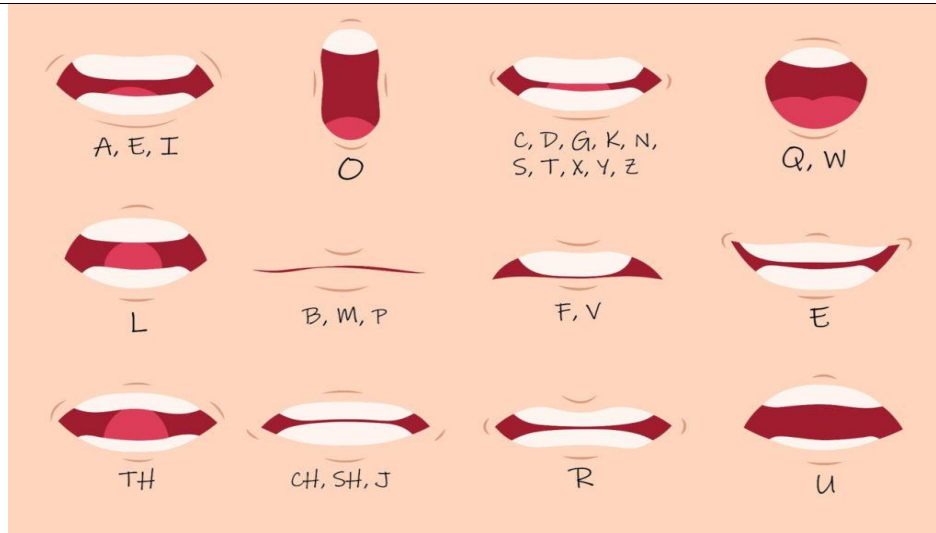


Figure 10. Illustrating the adobe character animator

Moreover, AI is playing a major part in the time-consuming voice over services performed during post-production. Adobe has released new animation software that utilizes artificial intelligence-powered lip-syncing to better synchronize the mouth motions of characters with their lines. If you're animating using Adobe Character Animator, you may use the software's built-in artificial intelligence engine, Adobe Sensei, to match lip movements to vocalizations. As can be seen in Figure 10, traditional frame-by-frame lip sync is used throughout the dictation process. This will allow the voice and character to mesh seamlessly.

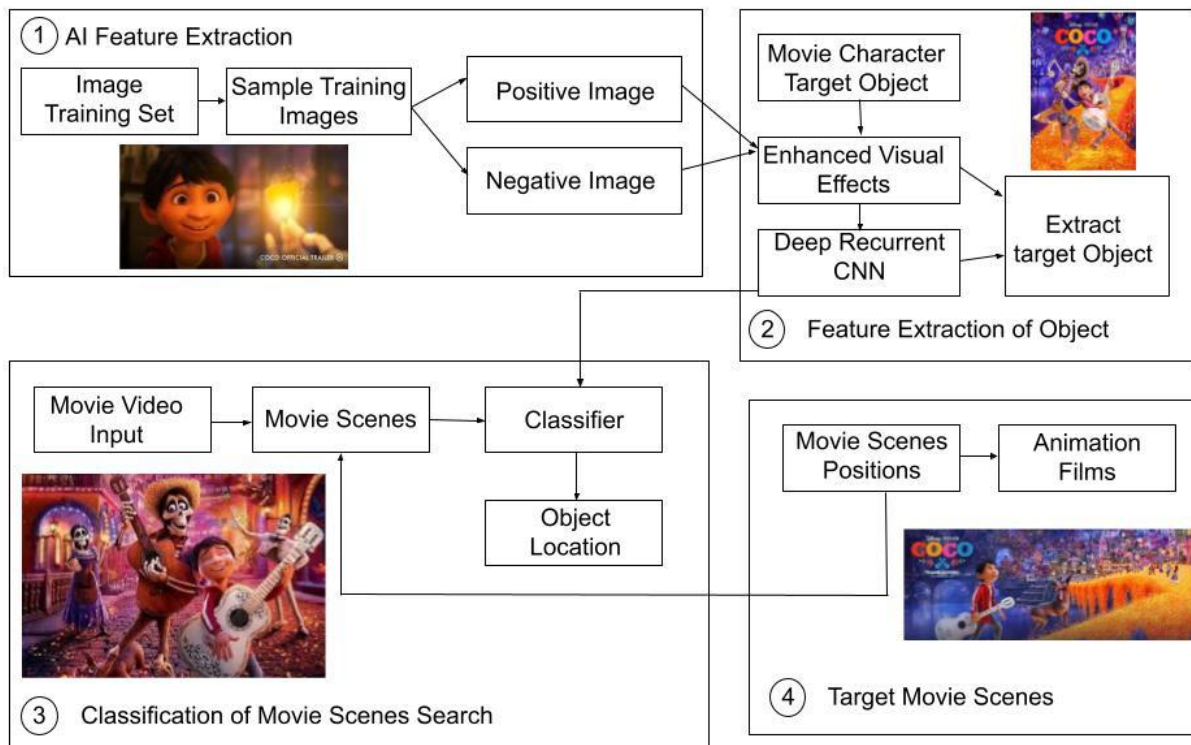


Figure 11. Recurrent Convolutional neural network for advanced visual effects of computer vision in coco animation films.

Several excellent techniques have emerged in response to applications in other disciplines; these algorithms are not reliant on the network topology and include the widely used SIFT features, Alex Net, RCNN, GoogLeNet, Faster RCNN, SOLO, and SSD. Many additional high-quality algorithms have been developed as well. In order to back up the current state of the art in artificial intelligence, which relies on the application of technology based on convolutional neural networks (CNNs), several algorithms have been devised. When implementing a revolutionary change, it is necessary to reevaluate all of the procedures that have come before. Input pictures are read in, analyzed, areas are extracted and labeled, features are extracted and classified, partial features are extracted, and classifiers are applied. Stochastic gradient descent may be used to determine the optimal configurations for your intermittent splicing panels and their related layers. Because of this, it is possible to get the data's most relevant qualities and classifications. A deep recursive convolutional neural network (shown in Figure 11) is used in this research's platform design, detection, and verification processes.

Due to the sheer volume of information, the digital picture we see is a compilation of many smaller images. In the film, a deep recursive convolutional neural network technique is used to quickly and accurately recognize and locate targets. This technology, which offers tools to enhance the accuracy of object detection, is used in the film. Modifying the LOSS function will allow us to do this.

6. CONCLUSION

It has the potential to both speed up the animation creation process and improve the quality of the final result. But it might aid the pursuit of animation accuracy, which would improve the experience for viewers and raise the bar for the medium as a whole. Implementing AI technology ethically requires studying the blunders of other nations. Similarly, the practicality of animation production for cinema and television has the potential to broaden the hybrid between animation and other kinds of creativity, therefore expanding new opportunities for both.

Following a brief overview of the research done in the subject and a scan of the literature, this article digs into the intricacies of the personalized recommendation algorithms that form the backbone of the recommendation system in a variety of sectors. This article presents a new methodology for studying anime visual effects and addresses the use of convolutional neural networks in this context. The new visual effects research model for animated films is developed, and the established algorithm is included into this model as part of the method for studying the development of these effects. As is customary when exploring new possibilities for visual effects in film, we do this test. The results of the further study on the principle of new vision animation films confirmed this.

We were unable to thoroughly review this content due to constraints on our time and resources. Cold-starting the algorithm, initializing and setting up the data, data structure, and striking a good balance between accuracy and recall are all challenges that require fixing. Tools and approaches for assessing the effects of AI on the film industry, including but not limited to filmmaking, film criticism, film promotion, and film distribution, are likely to become more popular as the area of artificial intelligence evolves.

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