Could you guess an interesting movie from the posters?:
An evaluation of vision-based features on movie poster database

Yuta Matsuzaki† Kazushige Okayasu† Takaaki Imanari† Naomichi Kobayashi†
Yoshihiro Kanehara† Ryousuke Takasawa† Akio Nakamura†
Hirokatsu Kataoka‡

†Department of Robotics and Mechatronics Tokyo Denki University
5 Asahi-cho, Adachi-ku, Tokyo 120-8551, JAPAN
{matsuzaki.y, okayasu.k}@is.fr.dendai.ac.jp
nkmr-a@cck.dendai.ac.jp

‡National Institute of Advanced Industrial Science and Technology (AIST)
1-1-1 Umezono, Tsukuba Ibaraki 305-8568, JAPAN
hirokatsu.kataoka@aist.go.jp

Abstract

In this paper, we aim to estimate the Winner of world-wide film festival from the exhibited movie poster. The task is an extremely challenging because the estimation must be done with only an exhibited movie poster, without any film ratings and box-office takings. In order to tackle this problem, we have created a new database which is consist of all movie posters included in the four biggest film festivals. The movie poster database (MPDB) contains historic movies over 80 years which are nominated a movie award at each year. We apply a couple of feature types, namely hand-craft, mid-level and deep feature to extract various information from a movie poster. Our experiments showed suggestive knowledge, for example, the Academy award estimation can be better rate with a color feature and a facial emotion feature generally performs good rate on the MPDB. The paper may suggest a possibility of modeling human taste for a movie recommendation.

1 Introduction

Imagine the situation when you see a movie at a theater. (see Figure 1).

Q. What if you decide the movie that you will watch?
A. Many movie posters are displayed in the theater, we will go to see the movie posters!

A good movie have a visually interesting movie poster in advertisement. The benefit of movie poster is greater than as we expected. Elberse et al. have placed that “advertising has a positive and statistically significant impact on market-wide expectations prior to release” and “the impact of advertising is lower for movies of lower quality” in the paper. According to the statistics, a good movie poster seems to be instresting that has interesting contents in the movie.

One one hand, the Academy award is one of the worldwide film festival which is explained as follows:

• The Academy Awards, or “Oscars”, is an annual American awards ceremony hosted by the AMPAS

Figure 1. This is an example of an exhibited movie poster. Question: Could you identify which movie from this movie poster is the film festival Winner? We will address this issue.

to recognize excellence in cinematic achievements in the United States film industry as assessed by the Academy’s voting membership.—(wikipedia)

The Academy awards have a long history as much as the 89th in 2016. The successive movies remain in the memory and record. A good movie like nominated for the Academy awards focus a lot of efforts into advertisement. As for the poster, it effects a great contribution to the impact of the movie and its sales. Furthermore, the expert’s eyes are refined in the long history, and it is excellent in a sensation to pick a good movie such a nominated one. Therefore, if we can learn a relationship between a winner and the poster, we enable to quantify the goodness of the movie from the poster. Here, we introduce an award prediction in film festival and similar research in computer vision field.

Award prediction In the Academy award prediction, Krauss et al. made a prediction based on a correlation value by utilizing a change of box-office takings, social situations and ratings on the IMDb. Although the accurate prediction is clear since the system enables to estimate an Academy award from various information, the approach is not helpful for choosing a good movie at a very moment. If we can model an expert’s judg-
ment with a movie poster, we instantly select a good movie at the movie theater.

*Related works in CV* Joo et al. succeeded to estimate election winning/losing from a facial image [3]. They extracted a low-level, mid-level feature from a face attribute and parts appearance of candidates and evaluated them. Based on the evaluation, it is possible to estimate that election winning/losing with 70% accuracy. Furthermore, even for the task of estimating camera-man from given pictures, Thomas et al. have significantly succeeded an estimation with 75% rate for 41 categories [4].

Above these lines, we believe that the vision-based technology allows us to predict a winner in the worldwide film festival with movie posters.

In this paper, we estimate a winner in the world-wide film festival from nominated movie posters. We have collected a database which contains a large amount of movie posters in the 4 biggest film festival (Academy, Berlin, Cannes, Venice). The movie poster database (MPDB) has historic movies over 80 years which are nominated a movie award at each year. We employ various types of features such as hand-craft (e.g. HOG, SIFT), mid-level (e.g. Face) and deep (e.g. AlexNet, VGGNet) features on the MPDB. In the experiments, we estimate a winner at each film festival with various features.

The contributions of the paper are as follows.

- The estimation of winner in the 4 biggest film festival with movie posters. To tackle the problem, we have evaluated various types of features.
- The collection of movie poster database (MPDB) which has 3,500+ nominates and 290+ winners over 80 years.

2 Movie poster database (MPDB)

Posters of this dataset are collected from Internet Movie Database (IMDb). The MPDB is consist of the nominations and winner films in the 4 biggest film festivals (Academy Awards, Berlin International Film Festival, Cannes International Film Festival, Venice International Film Festival). The image database contains 3,844 nominates, 295 winners in the film festivals. The Berlin International Film Festival, the Cannes International Film Festival, and the Venice international Film Festival are the 3 notable film festivals in the world. The Academy award is an authoritative award with a history that is older than the three major film festivals, and its influence on the box-office takings of the awards and the economy. The structure and detail of the MPDB are shown in Table 1. The poster images are annotated the winners and nominates. The database has been collected on the IMDb which is a prestigious movie rating service. However, we did not collect any ratings and other captions.

In the settings of training and testing, we assign a leave-one-year-out-cross-validation at each film festival. When a certain year is tested for winner prediction, others for the training. In the winner prediction, we calculate a posterior probability of a movie poster to be identified, then let the highest probability is selected as a winner. In addition to the original database, the training data is also subjected to data augmentation with 4 divisions (see Figure 2). In the Cannes and Venice, since there are two or more winner’s works, in that case winner acquired works in order from the posterior probability top. Moreover, there is a year that there are 0 winner works, so Winner estimation is not done for that year.

3 Features

Our goal is to train and estimate an award winner from movie posters. We have evaluated various features for the task. We employed (i) hand-craft feature, (ii) mid-level feature (iii) deep feature in order to extract various information from the posters in the MPDB. We itemize the features as below.

**Handcraft feature**

- SIFT + BoF [5][6]
- HOG [7]
- CoHOG [8]
- ECoHOG [9][10]
- L*a*b* [11]
- LBP [12]
- GIST [13]
- Combined handcraft feature

**Mid-level feature**

- FlickrStyle [14]
- PlaceNet [15]
- EmotionNet [16]
- Combined mid-level feature

Figure 2. Data augmentation

Figure 3. Comparison of various features in (i) handcraft (ii) mid-level (iii) deep features
### Deep feature

- AlexNet (DeCAF)\(^{[17]}\)
- VGGNet (DeCAF)\(^{[18]}\)
- Combined deep feature

Lab expresses each expression of L,a,b as 30 bins. Emotion expresses by inputting the face image acquired with Haar-like feature to EmotionNet and voting on the histogram to which the expression bin is allocated (Each histograms normalized). Also, rough face position information is included in the histogram (upper-left, upper-right, lower-left, lower-right). Therefore, the dimension of the face feature acquired in this research is 32. 

Classifier. The extracted feature quantity is identified by SVM. The parameters for identification were set to \(C = 5.0 \times 10^4\), \(\text{gamma} = 1.0 \times 10^{-5}\), \(\text{kernel} = "rbf"\). Combine. We combine features extracted by each feature extraction method. As a method of combination, “late fusion” is adopted.

### Experimtal Evaluation

The results of the winner prediction are shown in Figure 3. According to the Figure 3, it was confirmed that the identification results by the L*a*b* feature was the highest in the Academy award and relatively high with the EmotionNet. In addition, the Berlin and Venice also showed that the recognition rate using EmotionNet expression is the highest identification rate. We will consider focused on the feature whose identification rate is 30 % or more. In the Academy awards, there are many tendencies in which works selected as Winner tend to have certain colors. By analyzing L*a*b* histogram, we gained a knowledge that there are a couple of warm colors such as red, yellow, brown, etc., some of winners. We consider these colors contribute to improving the identification rate in the identification by the kernel SVM. In Figure 3, it can be seen that mode of facial expression is different in Berlin. In particularly Winner’s works are noticeable ‘Contempt’ in the upper left and upper right. In Venice (Figure 3), the remarkable difference is ‘Anger’ in the upper-left, ‘Sadness’ in the upper-right and lower-left. Further, it can be seen from figure 3 that the expression histogram of Winner’s works differs at each film festival. In Academy ‘Disgust’ appears in lower-right. In Berlin, ‘Contempt’ appears frequently in upper-right.In Cannes, ‘Disgust’ appears in upper-right and ‘Sadness’ appears in lower-right. In Venice, ‘Sadness’ appears in upper-right more. From this result, we suggested that it is possible to predict the movie festival by facial expression, because the facial expression drawn on the poster is different for each movie festival.

### 4 Exprimental Evaluation

| Film Festival | year | #winners | #nominates | Ave. #nominations at each year |
|---------------|------|----------|------------|------------------------------|
| Academy      | 1929-1932,1934-2016 | 88        | 440        | 16.6                        |
| Berlin       | 1951-2016          | 63        | 905        | 6.50                        |
| Cannes       | 1939,1946-1947,1949,1951-1968,1969-2016 | 91        | 1335       | 6.38                        |
| Venice       | 1932,1934-1942,1946-1972,1979-2016 | 53        | 869        | 5.75                        |

Table 1. Constitution of dataset

| Movie title     | Predicted score | Rank |
|-----------------|-----------------|------|
| Moonlight [Winner]| 0.167          | 1    |
| Lion            | 0.163          | 2    |
| Hell or High Water | 0.162        | 3    |
| Arrival         | 0.151          | 4    |
| Hacksaw Ridge   | 0.142          | 5    |
| Fences          | 0.138          | 6    |
| Hidden Figures  | 0.114          | 7    |
| Manchester by the Sea | 0.112     | 8    |
| La La Land      | 0.093          | 9    |

Table 2. Result of our prediction on Academy 2017 (sorted by predicted score)

In Cannes, accuracy was low in any feature, and there was no tendency of discrimination. We have considered that Cannes had chosen Winner based on box office income and the social situation at the time of movie release.

### 5 Application (further result in Academy 2017)

On February 26, 2017, the Academy Award 2017 was announced. We performed the winner prediction with our approach which assigns parameters from the experiments. The prediction results are shown in Table 2. As shown in Table 2, we have obtained interesting results at Academy Award 2017. Our system got a correct answer as the Academy Award 2017 “moonlight”.

### 6 Conclusion

In this work, we addressed a novel problem of estimating winner from a exhibited movie poster at film festival. In order to tackle this problem we have created a new dataset which is consist of winners and nominates in the 4 world-wide film festivals. The experiments on the self-collected MPDB (movie poster database) indicated a couple of suggestive realities such that the Academy award can be predictable with a color feature (L*a*b* corrects 34 % even though the random is around 15 %), also Berlin and Venice can be predictable with an emotional feature from face expression generally achieves well (around 25 % in the Academy, Berlin and Venice film festivals) on the MPDB. In Cannes, accuracy was low in any feature, and there was no tendency of winner.

This paper suggested the possibility of modeling human intuitions when selecting a movie. In the future, we would like to create a more sophisticated DCNN framework to estimate a winner in a movie festival.
Figure 4. Facial expression histogram between winner and nominate for each film festival (Bins of expression are arranged in order of upper left, upper right, lower left, lower right).

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