

■ 연구논문 요약문

<p>논문제목</p>	<p>Box-office forecasting based on sentiments of movie reviews and Independent subspace method</p>																																																																				
<p>게재정보</p>	<p>Information Sciences</p>																																																																				
<p>개요</p>	<p>Box-office forecasting is a challenging but important task for movie distributors in their decision making process. Many previous studies have tried to determine a way to accurately predict the box-office, but the results reported have not been satisfactory for two main reasons: (1) lack of variable diversity and (2) simplicity of forecasting algorithms. Although the importance of word-of-mouth (WOM) has consistently emphasized in past studies, only summarized information, such as volume or valence of user ratings is commonly used. In forecasting algorithms, multiple linear regression is the most popular algorithm because it generates not only predicted values but also variable significances. In this study, new box-office forecasting models are presented to enhance the forecasting accuracy by utilizing review sentiments and employing non-linear machine learning algorithms. Viewer sentiments from review texts are used as input variables in addition to conventional predictors, whereas three machine learning-based algorithms, i.e., classification and regression tree (CART), artificial neural network (ANN), and support vector regression (SVR), are employed to capture non-linear relationship between the box-office and its predictors. In order to provide variable importance for machine learning-based forecasting algorithms, an independent subspace method (ISM) is applied. Forecasting results from six different forecasting periods show that the presented methods can make accurate and robust forecasts.</p>																																																																				
<p>연구결과</p>	<table border="1"> <caption>Forecasting Results (Approximate Values)</caption> <thead> <tr> <th>Training Period</th> <th>Model</th> <th>W01</th> <th>W02</th> <th>W03</th> <th>W12</th> <th>W13</th> <th>W23</th> </tr> </thead> <tbody> <tr> <td rowspan="2">Tm 60%</td> <td>RSM</td> <td>0.45</td> <td>0.50</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> <tr> <td>ISM</td> <td>0.48</td> <td>0.52</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> <tr> <td rowspan="2">Tm 70%</td> <td>RSM</td> <td>0.45</td> <td>0.50</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> <tr> <td>ISM</td> <td>0.48</td> <td>0.52</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> <tr> <td rowspan="2">Tm 80%</td> <td>RSM</td> <td>0.45</td> <td>0.50</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> <tr> <td>ISM</td> <td>0.48</td> <td>0.52</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> <tr> <td rowspan="2">Tm 90%</td> <td>RSM</td> <td>0.45</td> <td>0.50</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> <tr> <td>ISM</td> <td>0.48</td> <td>0.52</td> <td>0.55</td> <td>0.15</td> <td>0.18</td> <td>0.05</td> </tr> </tbody> </table>	Training Period	Model	W01	W02	W03	W12	W13	W23	Tm 60%	RSM	0.45	0.50	0.55	0.15	0.18	0.05	ISM	0.48	0.52	0.55	0.15	0.18	0.05	Tm 70%	RSM	0.45	0.50	0.55	0.15	0.18	0.05	ISM	0.48	0.52	0.55	0.15	0.18	0.05	Tm 80%	RSM	0.45	0.50	0.55	0.15	0.18	0.05	ISM	0.48	0.52	0.55	0.15	0.18	0.05	Tm 90%	RSM	0.45	0.50	0.55	0.15	0.18	0.05	ISM	0.48	0.52	0.55	0.15	0.18	0.05
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<p>활용분야 및 기대효과</p>	<p>In this study, we developed box-office forecasting models based on machine learning-based algorithms with ISM. Three different factors, motion picture, external, and audience were considered as the source of input variables. Particularly for the audience factors, user sentiments in a review text, not just a simple rating, were also considered by constructing a domain-specific sentiment dictionary based on movie review data. In addition, input variables in these factors were assessed for each forecasting model using ISM, which was proved to be robust to a correlated structure. For a prediction task, one linear model, the MLR, and three machine-learning based algorithms, the CART, ANN, and SVR, were employed. The forecasting results showed that machine learning-based algorithms could make more accurate forecasts when available data is insufficient or for long-term forecasting.</p>																																																																				