A study on various techniques of image mining

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ABSTRACT

Image mining is a field of data mining that deals with the extraction of knowledge from large collections of digital images. It is a relatively new field, but it has grown rapidly in recent years due to the increasing availability of image data.

The amount of image data being generated is growing exponentially. This data can be used to improve our understanding of the world around us, but it is also a challenge to manage and analyze. There is a lot of knowledge that can be extracted from image data, but this knowledge is often hidden and difficult to discover. Image mining can be used to extract this knowledge and make it available to researchers and decision-makers. The manual analysis of image data is time-consuming and labor-intensive. Image mining can be used to automate the analysis of image data, freeing up human resources for other tasks.

The choice of features and data mining techniques depends on the specific application. For example, medical image analysis may use features such as the intensity of pixels, the shape of objects, and the texture of tissues. Security applications may use features such as the location of objects, the direction of movement, and the time of day.

Image mining is a rapidly growing field with a wide range of applications. As the amount of image data continues to grow, image mining will become increasingly important for extracting knowledge and insights from this data.

KEYWORDS: Image, Mining, Data

I. INTRODUCTION

There are many reasons why we need image mining. First, the amount of image data is growing exponentially. Every day, billions of new images are created and stored. This data is often unstructured and difficult to search. Image mining can help us to make sense of this data and extract valuable information. (Carter, 2018)

Second, image mining can be used to improve the accuracy of image retrieval. Traditional image retrieval systems rely on keywords to search for images. However, this can be inefficient and inaccurate, as keywords do not always accurately describe the content of an image. Image mining can be used to extract features from images that are more discriminative than keywords. This can lead to more accurate and efficient image retrieval.

Third, image mining can be used to detect objects and events in images. This can be used for a variety of applications, such as surveillance, medical diagnosis, and traffic monitoring. For example, image mining can be used to detect faces in a crowd, or to identify tumors in medical images.

Fourth, image mining can be used to classify images. This can be used to organize images into meaningful categories, such as "nature" or "cityscapes." Image classification can also be used to identify the content of an image, such as the type of object or the scene depicted.

Fifth, image mining can be used to find patterns in images. This can be used to identify trends, anomalies, or correlations in image data. For example, image mining can be used to find patterns in satellite images that can be used to track deforestation or climate change. (Seidel, 2019)

These are just a few of the many ways that image mining can be used. As the amount of image data continues to grow, the need for image mining will only increase.

Image mining is a powerful tool that can be used to extract knowledge from large collections of images. It has the potential to revolutionize the way we interact with images and to make more informed decisions.

In the automotive industry, image mining is being used to inspect car parts for defects. This can help to improve the quality of cars and to reduce the number of recalls.

In the food industry, image mining is being used to detect foodborne contaminants. This can help to keep food safe and to prevent outbreaks of illness.

In the pharmaceutical industry, image mining is being used to develop new drugs. This can help to speed up the drug discovery process and to improve the success rate of new drugs.

In the mining industry, image mining is being used to locate and extract minerals. This can help to improve the efficiency of mining operations and to reduce costs.

In the oil and gas industry, image mining is being used to monitor pipelines and other infrastructure. This can help to prevent leaks and spills and to improve safety.

These are just a few examples of the many ways that image mining is being used in industry today. As the technology continues to develop, we can expect to see even more innovative applications for image mining in the future. (Nagthane, 2019)

Various techniques of image mining

There are many different techniques of image mining, but some of the most common include:

• Object recognition: This is the task of identifying objects in an image. It is often used in applications such as facial recognition, medical image analysis, and video surveillance.

• Image classification: This is the task of assigning a label to an image, such as "cat" or "dog". It is often used in applications such as image search and content tagging.

• Image clustering: This is the task of grouping images together based on their similarity. It is often used in applications such as image retrieval and image segmentation.

• Image retrieval: This is the task of finding images that are similar to a given query image. It is often used in applications such as web search and e-commerce.

• Image segmentation: This is the task of partitioning an image into its constituent parts. It is often used in applications such as object recognition and image understanding. (Zhang , 2019)

These are just a few of the many techniques of image mining. The choice of technique depends on the specific application.

The techniques used in image mining can be broadly classified into two categories:

• Feature extraction: This involves extracting features from images that can be used to represent them. These features can be low-level, such as the intensity of pixels, or high-level, such as the presence of objects or scenes.

• Pattern recognition: This involves using machine learning algorithms to identify patterns in images. These patterns can be used to classify images, cluster images, or identify objects in images.

Some of the most common techniques used in image mining include:

• Principal component analysis (PCA): PCA is a statistical technique that can be used to reduce the dimensionality of images. This can be helpful when working with large datasets of images.

• Wavelet transform: The wavelet transform is a mathematical tool that can be used to decompose images into different frequency bands. This can be helpful for extracting features from images that are invariant to changes in scale or rotation.

• Support vector machines (SVMs): SVMs are a type of machine learning algorithm that can be used for classification and regression tasks. They have been shown to be effective for image mining tasks such as object recognition and image classification.

• Deep learning: Deep learning is a type of machine learning that uses artificial neural networks to learn from data. Deep learning has been shown to be very effective for image mining tasks such as object recognition and image classification. (Kreinovich, 2018)

Image mining is a field of data mining that deals with the extraction of knowledge from large collections of images. It is a relatively new field, but it has a wide range of applications, including:

• Medical image analysis: Image mining can be used to analyze medical images, such as X-rays, MRI scans, and CT scans, to detect diseases, identify abnormalities, and plan treatments.

• Security: Image mining can be used to detect objects and people in images, and to identify patterns of behavior. This can be used for security applications, such as surveillance, fraud detection, and border control.

• Retail: Image mining can be used to analyze product images to identify trends, improve product recommendations, and personalize the shopping experience.

• Social media: Image mining can be used to analyze social media images to understand user behavior, identify trends, and target advertising.

• Web search: Image mining can be used to improve the relevance of web search results by understanding the content of images.

The image mining process typically involves the following steps:

1. Image preprocessing: This step involves cleaning the images, removing noise, and converting them to a format that can be easily analyzed.

2. Feature extraction: This step involves extracting features from the images, such as color, texture, and shape.

3. Image mining: This step involves using data mining techniques to discover patterns in the extracted features. (Elmaghraby, 2017)

Here are some additional details about the image mining process:

• Image preprocessing: The goal of image preprocessing is to improve the quality of the images and make them easier to analyze. This may involve removing noise, correcting distortions, and resizing the images.

• Feature extraction: The goal of feature extraction is to identify the key features of the images that are relevant to the application. This may involve extracting features such as color, texture, shape, and location.

• Image mining: The goal of image mining is to discover patterns in the extracted features. This may involve using data mining techniques such as clustering, classification, and association rule mining.

Image mining is a challenging task, but it has the potential to yield valuable insights from large collections of image data. As the amount of image data continues to grow, image mining will become increasingly important for extracting knowledge and insights from this data.

Here are some of the challenges of image mining:

• Data volume: The amount of image data is growing rapidly, which makes it difficult to store and process.

• Data complexity: Image data is often complex and noisy, which makes it difficult to extract meaningful patterns.

• Heterogeneity: Image data can be heterogeneous, meaning that it can come in different formats and resolutions. This makes it difficult to develop a single image mining technique that can be used for all types of image data.

• Scalability: Image mining techniques need to be scalable to handle large volumes of image data.

Despite these challenges, image mining is a promising field with a wide range of applications. As the technology continues to develop, image mining will become more powerful and easier to use. This will make it possible to extract valuable insights from large collections of image data, which will have a significant impact on a variety of industries.

II. DISCUSSION

Image mining is a challenging task, but it has the potential to revolutionize the way we interact with images. By extracting knowledge from large collections of images, image mining can help us to better understand the world around us and to make more informed decisions. (Thiang, 2019) Here are some of the challenges in image mining:

• The high dimensionality of image data: Images are represented by a large number of features, such as the color of each pixel, the texture of the image, and the shape of the objects in the image. This makes it difficult to find patterns in image data.

• The noise in image data: Images are often noisy, which can make it difficult to extract meaningful information from them.

• The variability of image data: Images of the same object can vary greatly in appearance, depending on the lighting, the angle of view, and other factors. This makes it difficult to develop image mining techniques that are robust to variations in the data.

Despite these challenges, image mining is a rapidly growing field with a lot of potential. As the technology continues to improve, image mining is likely to become even more powerful and widespread.

Here are some of the applications of image mining:

• Medical imaging: Image mining can be used to diagnose diseases, track the progression of diseases, and design new treatments. For example, image mining has been used to identify early signs of cancer in medical images.

• Security: Image mining can be used to detect suspicious activity, such as people carrying weapons or entering restricted areas. For example, image mining is used by airports to detect potential terrorists.

• Retail: Image mining can be used to analyze customer behavior, personalize product recommendations, and improve product placement. For example, image mining is used by Amazon to recommend products to customers based on their past purchases.

• Social media: Image mining can be used to analyze social media posts, track trends, and identify influencers. For example, image mining is used by businesses to understand what their customers are talking about on social media.

• Entertainment: Image mining can be used to generate new content, such as movies and TV shows. For example, image mining is used by Hollywood studios to create special effects.

Image mining is a rapidly growing field with a wide range of applications. As the amount of digital images continues to grow, image mining is becoming increasingly important for extracting knowledge from these images.

Image mining can be used in a variety of industrial applications, including:

• Quality control: Image mining can be used to inspect products for defects. For example, it can be used to detect cracks in welds, blemishes on food products, or errors in printed circuit boards.

• Fraud detection: Image mining can be used to detect fraudulent activities, such as counterfeit products or insurance fraud. For example, it can be used to identify fake banknotes or to match images of people to known criminals.

• Logistics: Image mining can be used to track the movement of goods and materials. For example, it can be used to monitor inventory levels in warehouses or to track the delivery of packages.

• Safety: Image mining can be used to improve safety in industrial settings. For example, it can be used to detect hazards in the workplace or to monitor the behavior of workers.

• Maintenance: Image mining can be used to predict when equipment needs to be repaired or replaced. For example, it can be used to identify wear and tear on machinery or to detect cracks in structures.

These are just a few of the many ways that image mining is being used in industry. As the technology continues to develop, we can expect to see even more innovative applications for image mining in the future.

III. CONCLUSION

The future of image mining is bright. As the amount of image data continues to grow and the challenges in image mining are addressed, image mining will become an increasingly important tool for extracting knowledge from images.

Image mining has the potential to revolutionize the way we interact with and understand the world around us. It can be used to improve our health, safety, security, and quality of life.

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