

# How Computer Vision is Transforming Medical Imaging in Healthcare

As a physician and a specialist in health information technology, I have witnessed firsthand the transformative power of technology in healthcare. One such technology that has been making waves in recent years is computer vision. This powerful tool, which allows machines to 'see' and interpret visual data, is revolutionizing how we approach healthcare. This article aims to shed light on this topic, particularly for large providers, payers, and patients who stand to benefit immensely from these advancements.

## Use Cases in Healthcare

Computer vision, a field that falls under the broader umbrella of artificial intelligence, has been steadily gaining traction in the healthcare sector. It has the potential to drastically improve the way we diagnose diseases, monitor patient health, and even deliver treatments. As a physician, I have seen how this technology can enhance the accuracy of diagnoses, prevent medical errors, and improve patient outcomes.

1. **Automated Image Analysis** - using algorithms to automatically analyze medical images, identify patterns, and detect anomalies. Clinical researchers can train an algorithm to recognize the signs of a stroke in a brain scan. After feeding the algorithm a set of images for review, the analytics generated could include the number of suspected stroke cases properly identified, the accuracy of the algorithm, and the time saved by automating the process. To apply this, you would need to train a machine-learning model using a large dataset of brain scans, both with and without strokes.
2. **Predictive Analytics** - using historical data to predict future outcomes. By analyzing past mammograms and their associated outcomes, data scientists are building a predictive model to predict the likelihood of breast cancer in future patients. The analytics generated could include the accuracy of the predictions, the number of early detections made, and the impact on patient outcomes. To apply this, you would need a large dataset of mammograms and associated outcomes and a machine-learning model trained to identify patterns in this data.
3. **Precision Medicine** - analyzing medical images alongside other patient data to develop personalized treatment plans. By analyzing patients' MRI scans and their genetic data, oncologists could create a customized treatment plan for each diagnosed brain tumor. The analytics generated include the success rate of personalized treatment plans, patient outcomes improvement, and precision medicine's cost-effectiveness.
4. **Tumor Detection and Monitoring** - using analytics to detect tumors in medical images, determine their size and location, and



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Dr. Chaiken has over 25 years' experience in healthcare information technology, clinical transformation, and business intelligence. He provides thought leadership and strategic and analytics assessments in healthcare information technology, quality of care, clinical change management, and business development.

Chaiken has worked with the NIH, Tableau/Salesforce, Infor, McKesson, UK National Health Service, Boston University, and others.

## Navigating the Code

The healthcare industry, unlike many others, runs on time-tested ways to practice excellence in medicine. But does that mean adherence to practices and processes that are fifty, seventy, even a hundred years old?

Dr. Barry P. Chaiken thinks not. His 25+ years of experience as a physician and an informaticist, he believes information technology is healthcare's greatest problem-solving tool for resolving the greatest medical and business problems of the 21<sup>st</sup> century.

[Navigating the Code: How Revolutionary Transforms the Patient-Physician Journey—Available on Amazon \(Kindle and Audible\) and at \*navigatingthecode.com\*](#)

monitor changes over time. Researchers are creating algorithms to detect lung tumors in CT scans and monitor their growth over time. The analytics generated could include the number of tumors detected, the accuracy of the measurements, and the impact on treatment planning.

5. **Image Segmentation** – separating different structures in medical images, such as organs, tissues, and tumors. Algorithms can assist radiologists with segmenting a liver tumor from the surrounding healthy tissue in an MRI scan. Analytics can help improve the accuracy of the segmentation, reduce the time needed to review the scan, and enhance surgical planning.
6. **3D Reconstruction** – analyzing multiple 2D images to construct a 3D model of a patient's anatomy. For example, multiple X-ray images could build a 3D model of a patient's spine. The analytics generated include the accuracy of the 3D models, the improvement in surgical planning, and the impact on patient education.
7. **Comparative Analysis** – comparing a patient's medical images with a database of other images. To aid in the diagnosis, a physician can compare a patient's skin lesion to the images in a skin cancer and benign lesions database. The analytics generated could include the accuracy of the comparative analysis, the number of rare or complex conditions identified, and the impact on patient outcomes.
8. **Quality Control** – using analytics to ensure the quality of medical images. A properly designed algorithm can identify blurry or improperly aligned X-ray images. The analytics generated could include the number of poor-quality images identified, the improvement in image quality, and the impact on diagnosis accuracy.
9. **Radiomics** – involves extracting many quantitative features from medical images. A radiomics algorithm can identify features from a lung CT scan that predict the response to treatment in lung cancer, assisting the oncologist with treatment planning. The analytics generated for algorithm improvement include the accuracy of the predictions, the number of features extracted, and the impact on patient prognosis.
10. **Deep Learning** – using machine learning algorithms to analyze medical images and learn from the data. A deep learning model trained to identify signs of diabetic retinopathy in retinal images can assist with diabetic screening. The analytics generated could include the accuracy of the model, the number of cases identified, and the improvement in patient outcomes.

## Conclusion

The advent of computer vision in healthcare is a testament to the transformative power of technology. As we explore its potential, we can look forward to a future where healthcare is more accurate, effi-

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cient, and accessible. If you are part of a provider, payer, pharmaceutical, or life science company and are interested in exploring the potential of computer vision in your operations, feel free to reach out. Let us revolutionize healthcare together.

*[Author Note:](#) I wrote this article using ChatGPT (4.0). By requesting several “regenerations” of the responses, I constructed a more informative article from pieces of each version. This is the finished document.*