



# MEDICAL IMAGING PROCESSING

## FOURTH STAGE



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## Feature Extraction in Medical Imaging

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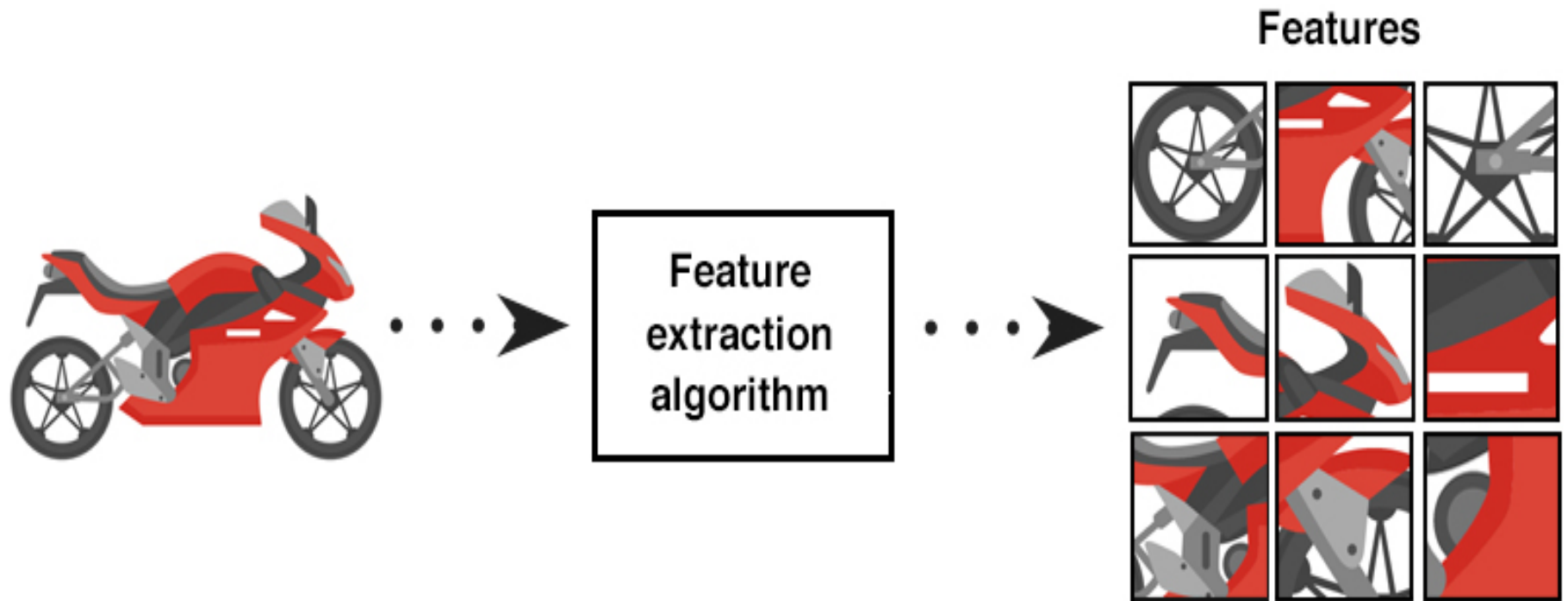
# Outline

- Introduction
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# Introduction

Feature extraction is one of the most important steps in medical image analysis. It aims to identify meaningful characteristics or patterns in images that can be used for diagnosis, classification, or segmentation.

In medical imaging, extracted features help describe tissues, organs, and abnormalities in a quantitative way, supporting computer-aided diagnosis (CAD) systems.



## 2. Importance of Feature Extraction

Feature extraction reduces the complexity of image data while keeping the essential information needed for analysis.

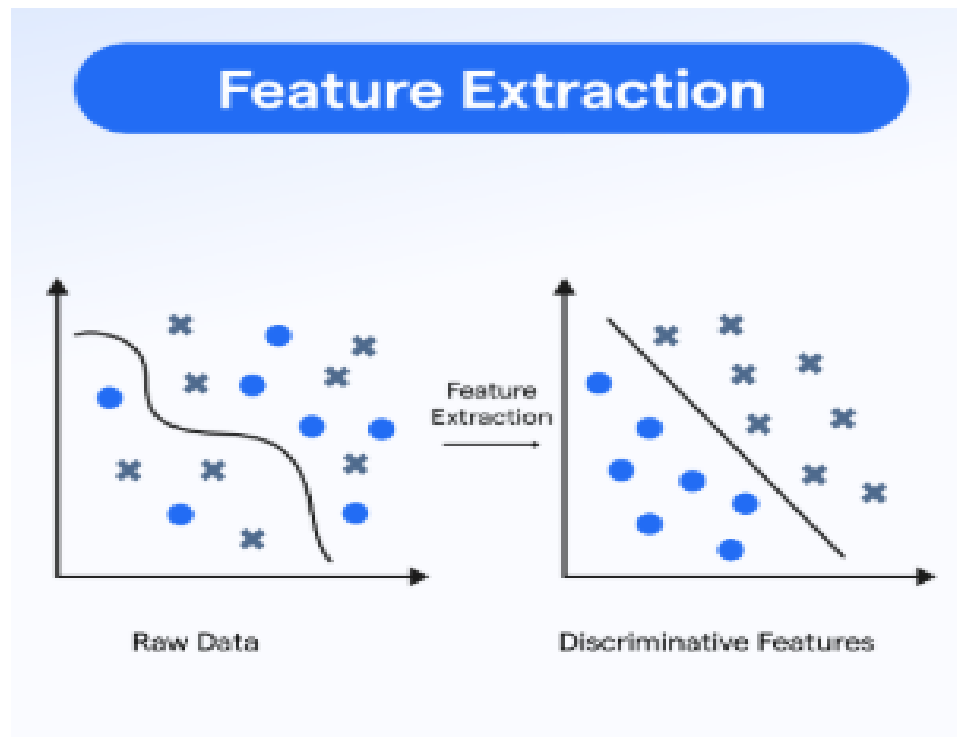
It helps in:

- ❖ Identifying abnormal regions (tumors, fractures, lesions).
- ❖ Improving model accuracy in machine learning.
- ❖ Speeding up processing by removing redundant data.
- ❖ Providing interpretable parameters for medical experts.

# 3. Types of Features

Medical image features can be divided into several categories:

- ❖ **Intensity-based features:-** pixel brightness values.
- ❖ **Shape-based features:-** object geometry, area, and perimeter.
- ❖ **Texture-based features:-** describe the smoothness or roughness of regions.
- ❖ **Frequency-based features:-** obtained using transforms like Fourier or Wavelet.

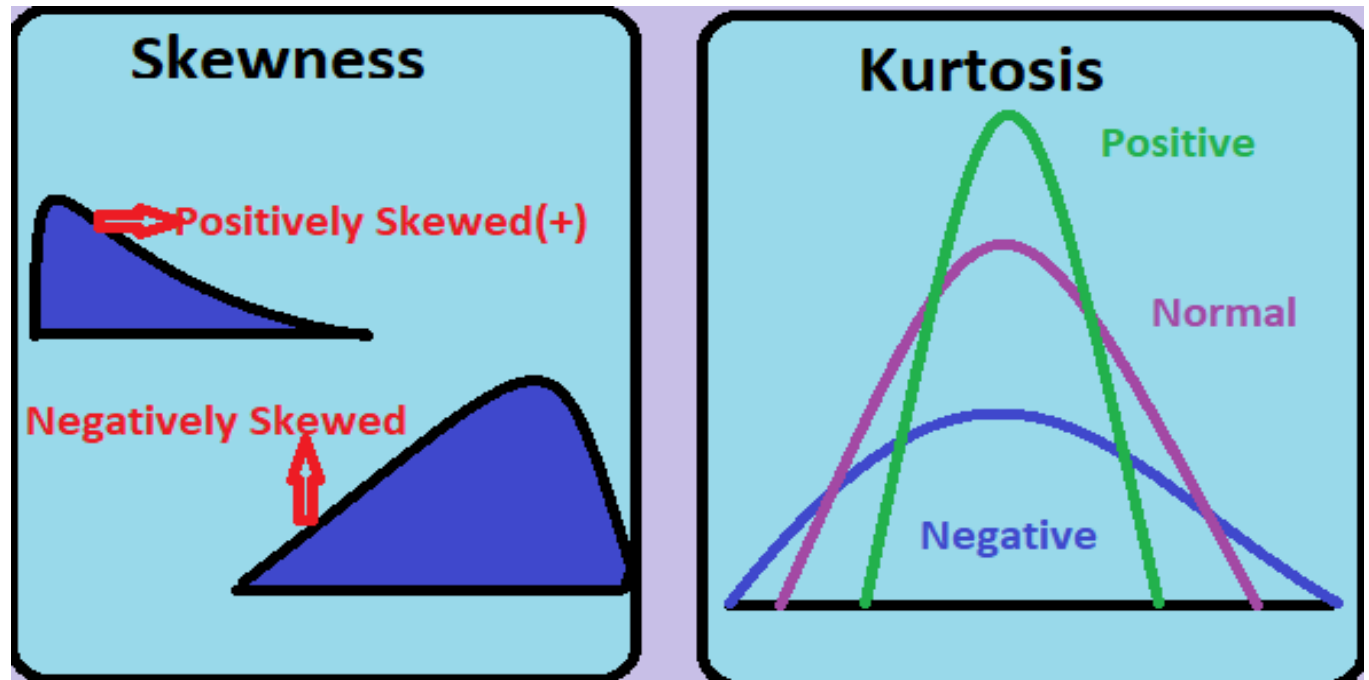


## 4. Statistical Features

Statistical features are simple yet powerful descriptors used in medical image classification. They include:

- ❑ Mean and Standard Deviation
- ❑ Skewness and Kurtosis
- ❑ Entropy (measure of randomness)

These features can represent tissue uniformity and texture differences in modalities like MRI or CT.



## 5. Texture Features

Texture features describe the spatial variation of pixel intensities in an image. They are often extracted using:

- **Gray Level Co-occurrence Matrix (GLCM)**
- **Local Binary Pattern (LBP)**
- **Gabor Filters**

These methods help distinguish between soft tissue, bone, and pathological regions.

# Gray Level Co-occurrence Matrix (GLCM)

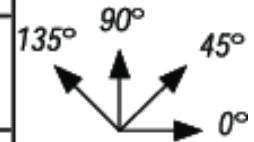
A Gray Level Co-occurrence Matrix (GLCM) is a statistical method for analyzing image texture by quantifying the spatial relationships between pixels. It works by counting how often pairs of pixels with specific gray level values appear together at a given distance and orientation, creating a square matrix that represents this joint probability. This matrix can then be used to derive second-order statistical features like energy, contrast, and entropy, which describe the texture of an image.

**a**

3	1	1	0	1
0	1	2	2	1
1	0	1	2	1
3	1	3	1	3
0	1	1	3	1

**b**

Pixel pair ↓ →	0	1	2	3
0	0	4	0	0
1	0	2	2	3
2	0	2	1	0
3	0	4	0	0





## 6. Shape Features

Shape-based features are useful in identifying anatomical structures or lesions.

Common examples include:

- **Area, Perimeter, and Compactness**
- **Aspect Ratio and Eccentricity**
- **Contour and Boundary Descriptors**

They are often used in tumor segmentation or bone fracture detection.

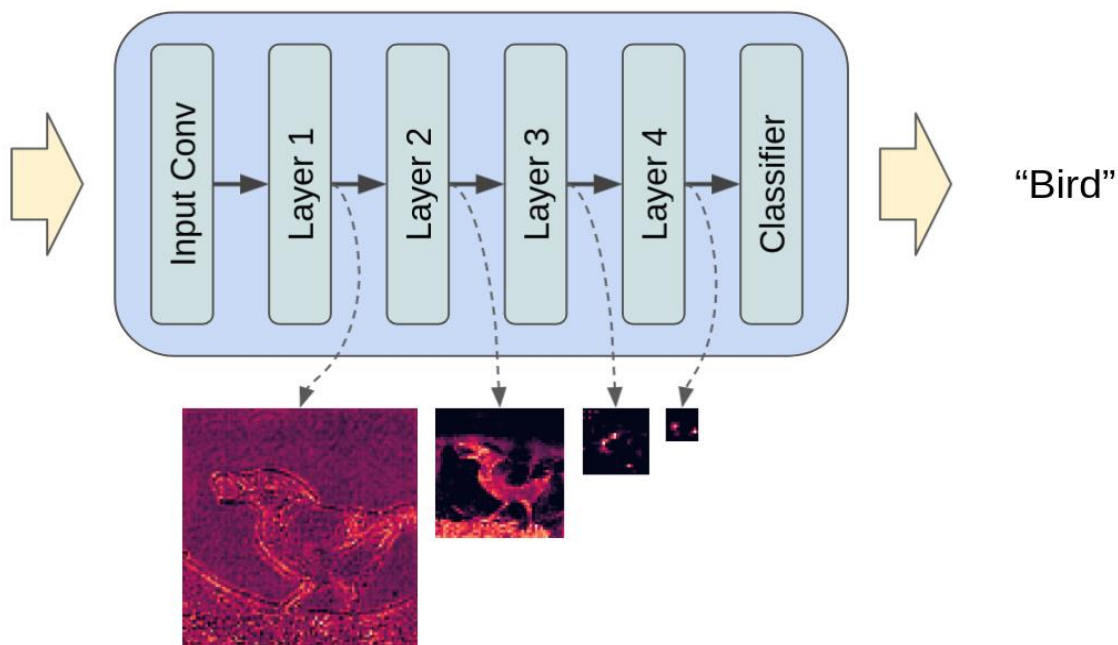
# 7. Transform-Based Features

Transform-based methods extract features from the frequency domain of an image.

Popular techniques include:

- **Fourier Transform:** analyzes frequency components.
- **Wavelet Transform:** captures both frequency and spatial details.

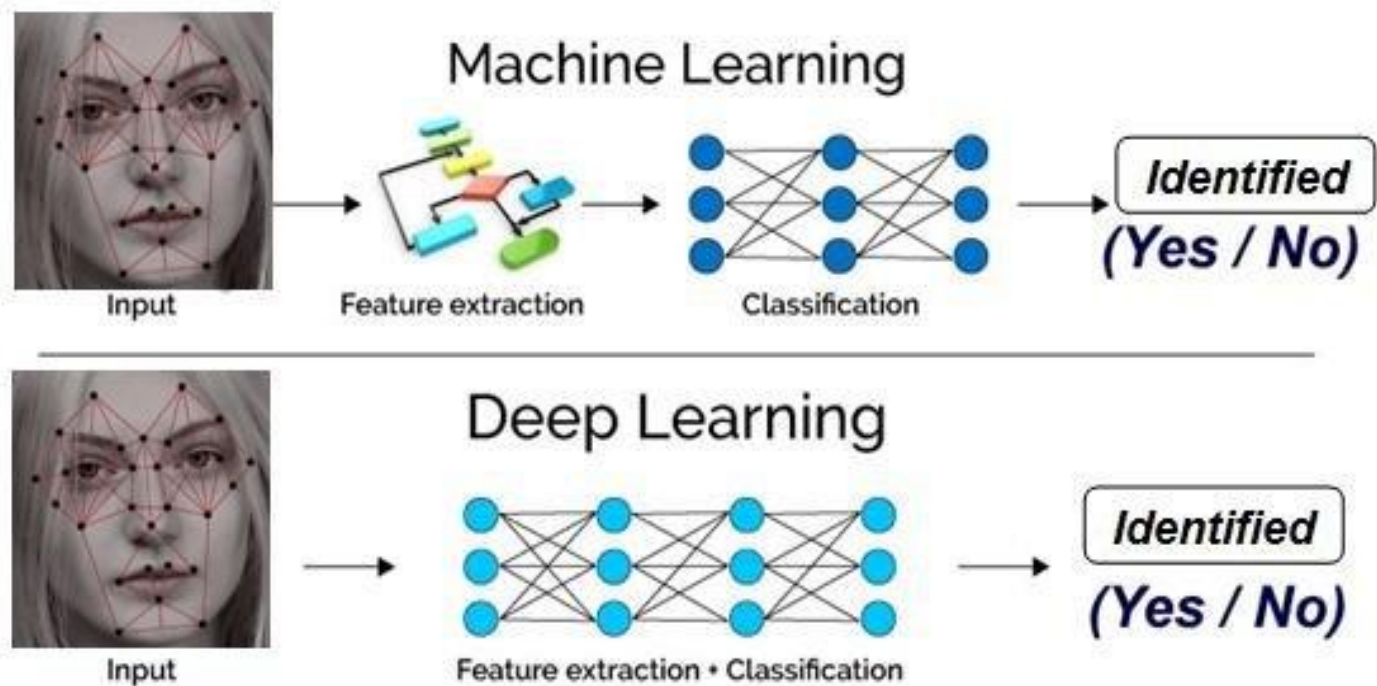
These features are useful for detecting subtle patterns that may not appear clearly in the spatial domain.



## 8. Deep Learning Features

Deep learning models such as Convolutional Neural Networks (CNNs) can automatically extract high-level features from medical images.

These features represent complex visual patterns that may correspond to disease markers or anatomical structures, eliminating the need for manual feature engineering.



## 9. Applications and Conclusion

Feature extraction plays a critical role in medical image processing, enabling accurate analysis in:

- **Disease diagnosis (e.g., cancer, fractures)**
- **Organ segmentation**
- **Tissue classification**
- **Computer-aided detection (CAD)**

In conclusion, effective feature extraction transforms raw medical images into structured information that supports both automated and human decision-making in healthcare.

GOOD LUCK  
EVERYONE