Classification of Blood Clot Origins in Ischemic Strokes 🩸

This notebook explores the task of classifying the etiology of blood clots in whole-slide digital pathology images, specifically identifying whether they are of Cardioembolic (CE) or Large Artery Atherosclerosis (LAA) origin. Through an extensive exploratory data analysis (EDA), we describe the dataset, analyze missing and duplicate values, examine the distribution of image sizes, classify variables, and review the label distribution for the training set, along with plenty of other analysis. This EDA provides a comprehensive understanding of the data and helps identify potential preprocessing steps for optimal model performance.

Following the EDA, we preprocess the images to standardize them for model input. The preprocessing involves resizing, converting images to grayscale, normalizing pixel values, and applying Gaussian blur to reduce noise. These steps ensure that the images are suitable for a Convolutional Neural Network (CNN) by preparing them with a consistent size, format, and reduced noise, enabling more efficient training and improved classification accuracy.

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(1) Import Libraries ⬇️

In [ ]:
```
# Data manipulation and visualization
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
import seaborn as sns
from sklearn.model_selection import train_test_split
import subprocess
from PIL import Image
import tifffile as tifi
import cv2

# Set the maximum allowable pixels to a higher number.
Image.MAX_IMAGE_PIXELS = None

# Standard libraries
```

```
import warnings
warnings.filterwarnings('ignore')

# ===== ===== Reproducibility Seed ===== =====
# Set a fixed seed for the random number generator for reproducibility
random_state = 42

# Set matplotlib inline
%matplotlib inline

# Set default figure size
plt.rcParams['figure.figsize'] = (6, 4)

# Define custom color palette
palette = sns.color_palette("viridis", 12)

# Set the style of seaborn
sns.set(style="whitegrid")
```

(2) Data Upload 📄

In [ ]:

```
# Note:
# -----
# For this specific task, we are only using the training data (`train.csv`)
# The test dataset (`test.csv`) will not be used until the very end when mak
# We will train, validate, and tune your model using the training data only.
#
# The reason the test data is only used at the very end is to prevent any bi
# We want the test data to remain completely unseen until after we've finali
# as a true evaluation of our model's performance.
# ---------------------------------------------

df = pd.read_csv('../input/mayo-clinic-strip-ai/train.csv')  # Load the trai
df.head()  # Display the first 5 rows of the DataFrame for a quick inspectic
```

Out[ ]:

	image_id	center_id	patient_id	image_num	label
0	006388_0	11	006388	0	CE
1	008e5c_0	11	008e5c	0	CE
2	00c058_0	11	00c058	0	LAA
3	01adc5_0	11	01adc5	0	LAA
4	026c97_0	4	026c97	0	CE

(3) Exploratory Analysis 🔎

(1) Descripción de los Datos

In [ ]:
```
# Print the number of records in the DataFrame
print("The given dataset has", df.shape[0], "registers and", df.shape[1], "c
```

```
The given dataset has 754 registers and 5 columns.
```

Observaciones 💡 -->

> - El conjunto de datos contiene más de mil imágenes de patología digital de alta resolución de diapositivas completas. Cada diapositiva representa un coágulo de sangre de un paciente que sufrió de un accidente cerebrovascular isquémico agudo.

> - En el conjunto `train.csv` , con el que se trabajará por el momento, se cuenta con 754 registros y 5 columnas, lo que indica que tiene una dimensión relativamente pequeña. Cada uno de los 754 registros representa una anotación única con relación a una de las imágenes dentro del directorio `train/` .

Fuente: Página oficial de Kaggle

In [ ]:
```
# Basic information about the dataset
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 754 entries, 0 to 753
Data columns (total 5 columns):
 #   Column      Non-Null Count  Dtype 
---  ------      --------------  ----- 
 0   image_id    754 non-null    object
 1   center_id   754 non-null    int64 
 2   patient_id  754 non-null    object
 3   image_num   754 non-null    int64 
 4   label       754 non-null    object
dtypes: int64(2), object(3)
memory usage: 29.6+ KB
```

- `image_id` : Un identificador único para la instancia, con el formato {patient_id}_{image_num}. Corresponde a la imagen {image_id}.tif.
- `center_id` : Identifica el centro médico donde se obtuvo la diapositiva.
- `patient_id` : Identifica al paciente del que se obtuvo la diapositiva.
- `image_num` : Enumera las imágenes de coágulos obtenidas del mismo paciente.
- `label` : La etiología del coágulo, que puede ser CE (embolia cardioembólica) o LAA (ataque isquémico). Este campo es el objetivo de clasificación.

Nota importante: Las diapositivas que conforman los conjuntos de entrenamiento y prueba representan coágulos con una etiología (es decir, origen) que se conoce como CE (cardioembólica) o LAA (aterosclerosis de grandes arterias).

Fuente: Página oficial de Kaggle

(2) Clasificación de Variables

Nombre	Descripción	Tipo
image_id	Identificador único de la imagen.	Cualitativa (Nominal)
center_id	Identificador del centro médico donde se tomó la diapositiva.	Cualitativa (Nominal)
patient_id	Identificación del paciente de la diapositiva.	Cualitativa (Nominal)
image_num	Número que indica la secuencia de imágenes de un mismo paciente.	Cuantitativa (Discreta)
label	Clasificación del coágulo: CE (cardioembólica) o LAA (aterosclerosis).	Cualitativa (Nominal)

Observaciones 💡

> - En nuestro conjunto de datos, la mayoría de las variables son de tipo **cualitativo nominal**.

> - Por otro lado, solo una variable es de tipo **cuantitativo discreto**, que corresponde al número de imagen.

(3) Exploración y Limpieza Inicial de los Datos

(1) Análisis de Data Faltante

In [ ]:
```
df.isnull().sum()
```

Out[ ]:
```
image_id      0
center_id     0
patient_id    0
image_num     0
label         0
dtype: int64
```

Observaciones 💡

> - A partir de este breve análisis de los datos faltantes, podemos observar que no hay columnas con valores nulos. Esto significa que no es

> necesario realizar ningún tipo de imputación en el conjunto de datos. Todas las variables están completamente pobladas, lo que garantiza la integridad de los datos para su análisis posterior.

(2) Previsualización de Imágenes

In [ ]:
```python
# Assuming your DataFrame is named df
base_path = "../input/mayo-clinic-strip-ai/train/"

# Add the full path to the df
df['image_path'] = base_path + df['image_id'] + '.tif'

# Preview the DataFrame to ensure the new column is added correctly
df.head()
```

Out[ ]:

	image_id	center_id	patient_id	image_num	label	image_path
0	006388_0	11	006388	0	CE	../input/mayo-clinic-strip-ai/train/006388_0.tif
1	008e5c_0	11	008e5c	0	CE	../input/mayo-clinic-strip-ai/train/008e5c_0.tif
2	00c058_0	11	00c058	0	LAA	../input/mayo-clinic-strip-ai/train/00c058_0.tif
3	01adc5_0	11	01adc5	0	LAA	../input/mayo-clinic-strip-ai/train/01adc5_0.tif
4	026c97_0	4	026c97	0	CE	../input/mayo-clinic-strip-ai/train/026c97_0.tif

Observaciones 💡

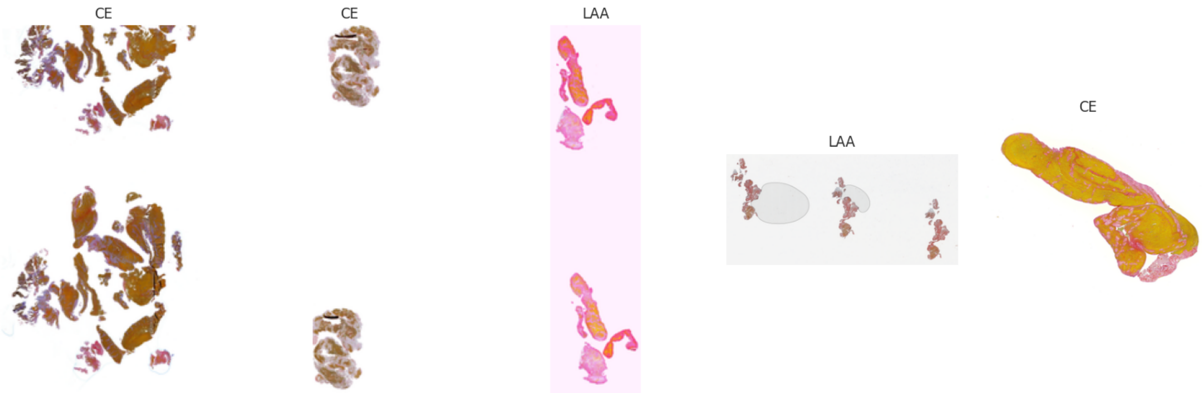
> - Una de las principales técnicas de preprocesamiento será **incluir el path o la dirección de cada imagen** dentro del DataFrame. Esto permitirá acceder a las imágenes de manera más sencilla, ya que, como se ha mencionado anteriormente, todas las imágenes se encuentran almacenadas en el directorio `train`.

In [ ]:
```python
def plot_images(df, num_images=5):
    # Select first n images
    sample_df = df.head(num_images)

    # Create subplots
    fig, axes = plt.subplots(1, num_images, figsize=(15, 5))

    # Flatten axes to make it easier to iterate
    axes = axes.flatten()

    for i, (img_path, label) in enumerate(zip(sample_df['image_path'], sampl
        img = Image.open(img_path)
        img.thumbnail((300,300), Image.Resampling.LANCZOS)
```

```python
        axes[i].imshow(img)
        axes[i].set_title(label)
        axes[i].axis('off')  # Hide axes

    plt.tight_layout()
    plt.show()
```

In [ ]:

```python
# Show 5 images
plot_images(df)
```



Observaciones 💡

> - Una de las primeras observaciones clave es la gran variabilidad en los tamaños de las imágenes presentes en el conjunto de datos. Estas imágenes muestran diferentes valores de relación de aspecto, lo cual puede presentar desafíos al momento de alimentar estos datos en un modelo predictivo.

> - También hemos detectado una variación considerable en las características cromáticas de las imágenes. Por ejemplo, algunas de ellas presentan fondos blancos, mientras que otras tienen fondos grises. Esto puede introducir ruido en el modelo si no se trata adecuadamente, ya que los modelos de visión por computadora son sensibles a las variaciones en las condiciones de iluminación y color de las imágenes.

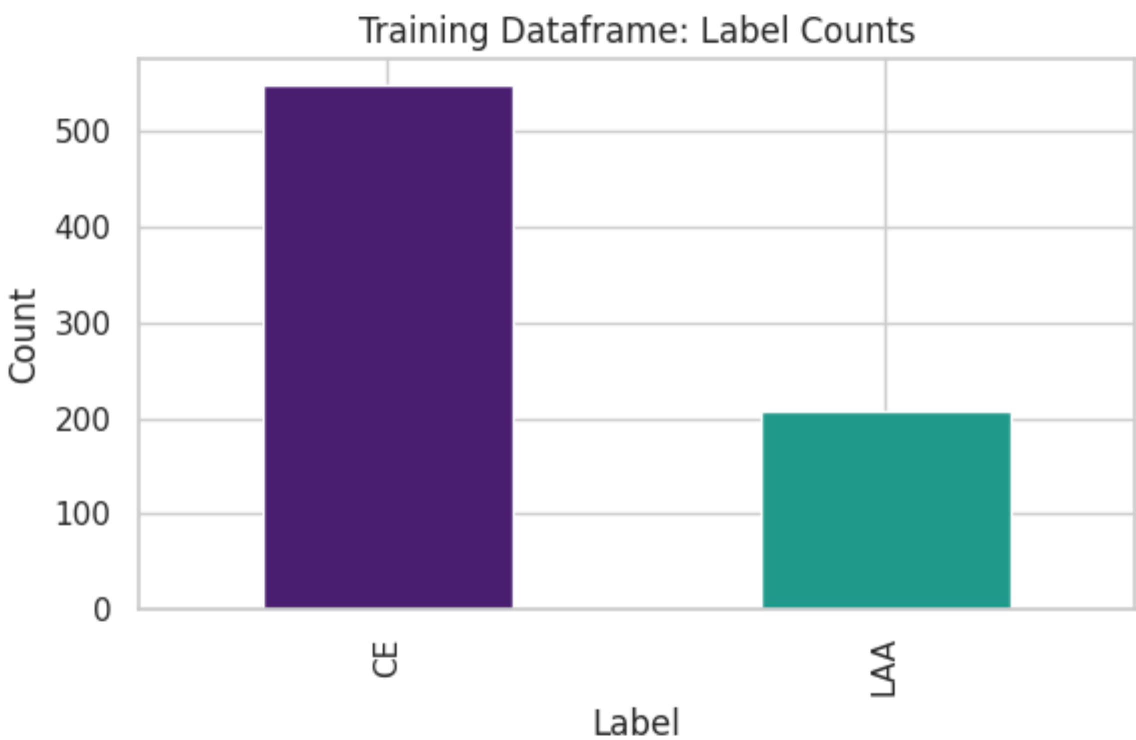
> - Adicionalmente, las imágenes presentan una mezcla de elementos que pueden afectar la precisión del modelo. Algunos coágulos están claramente definidos, mientras que otros están parcialmente oscurecidos o presentan sombras y brillos que pueden confundir al modelo.

Todas extas observaciones reflejan los primeros desafíos que enfrentaremos en el preprocesamiento de imágenes, un paso que se realizará más adelante.

(3) Distribución de Labels

In [ ]:
```
def label_distribution(df, name='Training Dataframe'):
  label_counts = df['label'].value_counts().sort_index()
  total_counts = label_counts.sum()

  # Calculate percentages
  label_percentages = (label_counts / total_counts) * 100

  # Plot counts
  plt.plot(1, 2, 1)
  label_counts.plot(kind='bar', color=[palette[0], palette[6]])
  plt.title(f'{name}: Label Counts')
  plt.xlabel('Label')
  plt.ylabel('Count')
  plt.tight_layout()
  plt.show()
```

In [ ]:
```
label_distribution(df)
```



Observaciones 💡

> - Nuestro conjunto de datos presenta un **fuerte desbalance**, ya que se cuenta con aproximadamente 550 imágenes de coágulos de **etiología cardioembólica (CE)**, lo que representa más del 70.0% del total de muestras.

> - En contraste, solo se disponen de 200 imágenes de coágulos de **etiología aterosclerótica de grandes arterias (LAA)**, lo que corresponde a **menos del 30.0%** del conjunto de datos.

> - Este desequilibrio entre las clases puede tener un impacto negativo en

> el rendimiento del modelo predictivo, ya que probablemente este se incline a **predecir mayormente la clase CE** debido a la falta de representatividad de la clase LAA. Un modelo sesgado de esta manera podría no capturar adecuadamente las características distintivas de los coágulos LAA, reduciendo su capacidad para hacer predicciones precisas.

> - Para mitigar este problema, se tendrá que explorar técnicas como **submuestreo de la clase mayoritaria**, o el uso de algoritmos tales como **data augmentation**. Aunque esto forma parte de los siguientes pasos a tomar.

(4) Análisis de Tamaños de Imagen

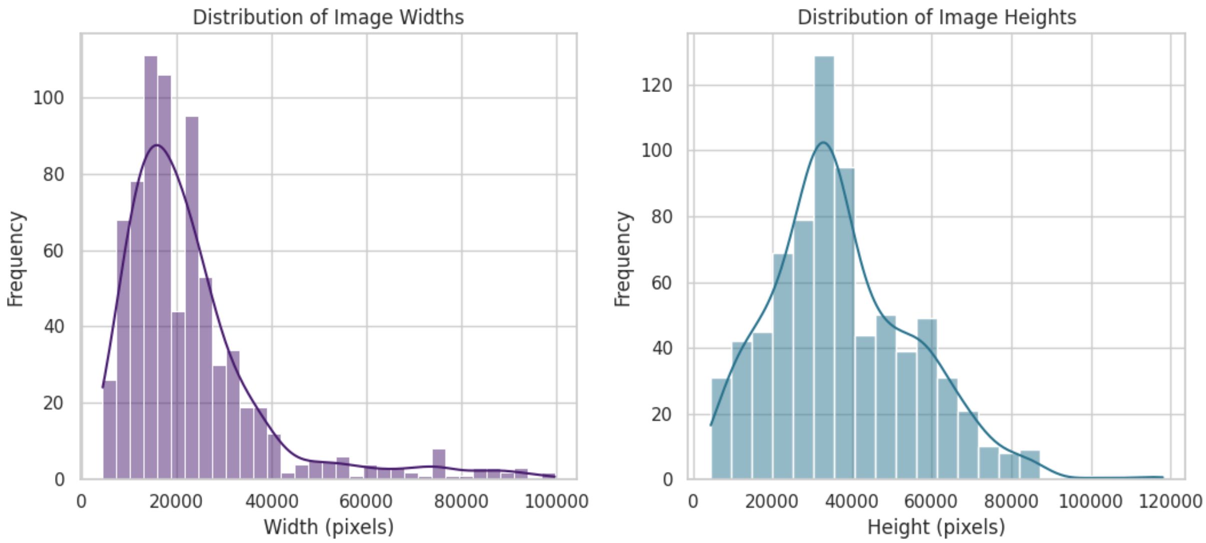
In [ ]:
```
def get_image_size(image_path):
    try:
        with Image.open(image_path) as img:
            return img.size  # Returns (width, height)
    except Exception as e:
        print(f"Error loading {image_path}: {e}")
        return None

# Applying the function to the DataFrame
df['image_size'] = df['image_path'].apply(get_image_size)

# Separate width and height if needed
df['width'], df['height'] = zip(*df['image_size'].dropna())
```

Observaciones 💡

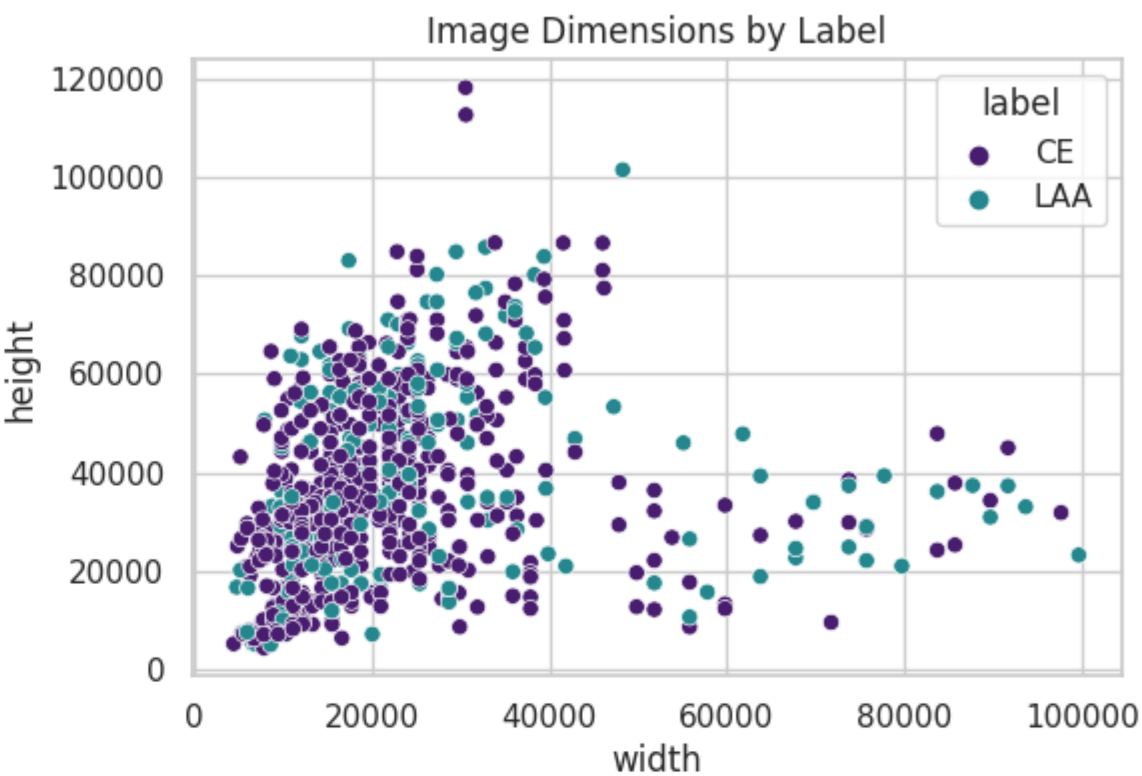
> - Como parte del proceso de preprocesamiento de datos, se extraerán las dimensiones (ancho y alto) de cada imagen del DataFrame. Estas dimensiones se almacenarán en dos columnas separadas (width y height) para facilitar el acceso y análisis de esta información. Este paso es importante ya que nos permitirá analizar la distribución de los tamaños de las imágenes, lo cual puede ser útil para determinar de qué manera se deberá redimensionar estas en los próximos pasos.

In [ ]:
```
# Set up the plot size and style
plt.figure(figsize=(16, 5))

# Plot the distribution of the width
plt.subplot(1, 3, 1)
sns.histplot(df['width'], kde=True, color=palette[0])
plt.title('Distribution of Image Widths')
plt.xlabel('Width (pixels)')
plt.ylabel('Frequency')

# Plot the distribution of the height
```

```
plt.subplot(1, 3, 2)
sns.histplot(df['height'], kde=True, color= palette[4])
plt.title('Distribution of Image Heights')
plt.xlabel('Height (pixels)')
plt.ylabel('Frequency')

# Show the plots
plt.tight_layout()
plt.show()
```



In [ ]:
```
sns.scatterplot(data=df, x='width', y='height', hue='label', palette=[palett
plt.title('Image Dimensions by Label')
plt.show()
```



In [ ]:
```
# Display descriptive statistics for image dimensions
print("Training Set: Image Width Statistics")
print(df['width'].describe())

print("\nTraining Set: Image Height Statistics")
print(df['height'].describe())
```

```
Training Set: Image Width Statistics
count      754.000000
mean     22988.594164
std      15653.642619
min       4417.000000
25%      13215.250000
50%      18700.000000
75%      26376.750000
max      99699.000000
Name: width, dtype: float64

Training Set: Image Height Statistics
count       754.000000
mean      37622.196286
std       18058.750676
min        4470.000000
25%       25402.500000
50%       34981.500000
75%       48919.750000
max      118076.000000
Name: height, dtype: float64
```

Observaciones 💡

> - En estos nuevos gráficos se confirma nuevamente la presencia de múltiples valores tanto en altura como en ancho de las imágenes. La mayoría de las imágenes presentan un ancho concentrado alrededor de los 20,000 píxeles, mientras que la altura se encuentra predominantemente alrededor de los 40,000 píxeles, como lo muestra claramente el histograma.

> - Además, al continuar con este análisis, podemos observar en el diagrama de dispersión que existe una relación predominante entre el tamaño de las imágenes y el tipo de etiología del coágulo del paciente.

> - Se han identificado **datos atípicos**, específicamente en relación con las dimensiones de las imágenes. Sin embargo, estos valores no son de gran relevancia, ya que se normalizarán al redimensionar las imágenes para su inclusión en el modelo.

A continuación veremos cuál es la imagen que posee las dimensiones más grandes y sus detalles.

In [ ]:
```
# Find the largest image without creating a new column for area
largest_image_index = (df['width'] * df['height']).idxmax()
```

```python
largest_image = df.loc[largest_image_index]

# Print the details neatly
print("Largest Image Details:")
print("-" * 30)
for column in df.columns:
    print(f"{column}: {largest_image[column]}")

# Open and display the image using Matplotlib
image_path = largest_image['image_path']
img = Image.open(image_path)
img.thumbnail((400,400), Image.Resampling.LANCZOS)

plt.imshow(img)
plt.axis('off')  # Hide axes
plt.title(f"Largest Image: {largest_image['image_id']}")
plt.show()

del img
```

```
Largest Image Details:
------------------------------
image_id: 6baf51_0
center_id: 11
patient_id: 6baf51
image_num: 0
label: LAA
image_path: ../input/mayo-clinic-strip-ai/train/6baf51_0.tif
image_size: (48282, 101406)
width: 48282
height: 101406
```

Largest Image: 6baf51_0

(5) Cross-Tab Analysis

In [ ]:
```python
# Create a cross-tabulation of 'center_id' and 'label'
```

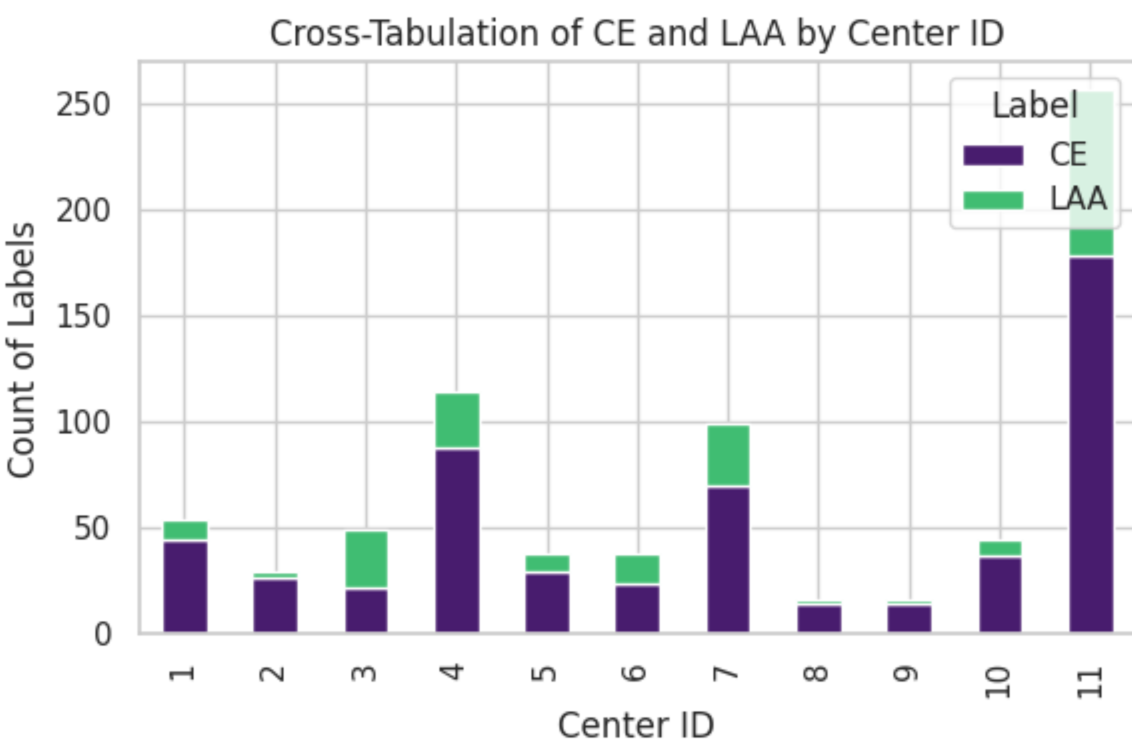
```
cross_tab = pd.crosstab(df['center_id'], df['label'])

# Display the cross-tabulation table
cross_tab.head()
```

Out[ ]:

label center_id	CE	LAA
1	44	10
2	26	3
3	22	27
4	88	26
5	29	9

In [ ]:

```
# Plot a stacked bar plot for better visualization
cross_tab.plot(kind='bar', stacked=True, color=[palette[0], palette[8]])

# Adding titles and labels
plt.title('Cross-Tabulation of CE and LAA by Center ID')
plt.xlabel('Center ID')
plt.ylabel('Count of Labels')
plt.legend(title='Label', loc='upper right')

# Show plot
plt.tight_layout()
plt.show()
```



Observaciones 💡

> - El Centro 11 tiene la mayor cantidad de imágenes, predominantemente etiquetadas como CE (179) en comparación con LAA (7). Este desequilibrio significativo sugiere un posible enfoque o especialización en casos cardioembólicos en este centro.

> - El Centro 4 también muestra una fuerte prevalencia de CE con 88 casos frente a 26 casos de LAA.

> - Los Centros 3 y 7 demuestran una distribución más equilibrada, con el Centro 3 teniendo 22 CE y 27 LAA, y el Centro 7 reportando 70 CE frente a 29 LAA.

> - Otros centros, como el Centro 2 y el Centro 5, presentan un menor número de casos en general, con el Centro 2 mostrando 26 CE y 3 LAA, lo que indica una posible subrepresentación de casos de aterosclerosis de grandes arterias.

Vemos que en general, los datos sugieren que los centros están más especializados en etiologías de accidente cerebrovascular, particularmente en casos **cardioembólicos**.

(6) Tablas de Frecuencia

In [ ]:
```
# Define a function for formatted printing
def print_frequent_values(df):
    # Iterate over each column in the DataFrame
    for column in df.columns:
        frequency_values = df[column].value_counts().head(10)
        print(f"\n{'='*50}\nTop 10 most frequent values for column '{column}
        print(f"{'Index':<5} {'Value':<30} {'Frequency':<10}")
        print('-' * 50)  # Separator line

        for index, (value, frequency) in enumerate(frequency_values.items(),
            print(f"{index:<5} {str(value)[:30]:<30} {frequency:<10}")

        print(f"{'='*50}")
```

In [ ]:
```
# Call the function
print_frequent_values(df)
```

```
==================================================
Top 10 most frequent values for column 'image_id':
Index Value                          Frequency
--------------------------------------------------
1     ffec5c_1                       1
2     006388_0                       1
3     008e5c_0                       1
4     00c058_0                       1
5     01adc5_0                       1
6     026c97_0                       1
7     028989_0                       1
8     029c68_0                       1
9     032f10_0                       1
10    0372b0_0                       1
==================================================

==================================================
Top 10 most frequent values for column 'center_id':
Index Value                          Frequency
--------------------------------------------------
1     11                             257
2     4                              114
3     7                              99
4     1                              54
5     3                              49
6     10                             44
7     5                              38
8     6                              38
9     2                              29
10    8                              16
==================================================

==================================================
Top 10 most frequent values for column 'patient_id':
Index Value                          Frequency
--------------------------------------------------
1     91b9d3                         5
2     56d177                         5
3     09644e                         5
4     3d10be                         5
5     a4c877                         4
6     f40c69                         4
7     4f6fb1                         4
8     5987c0                         4
9     a26055                         3
10    abc4a3                         3
==================================================

==================================================
Top 10 most frequent values for column 'image_num':
Index Value                          Frequency
--------------------------------------------------
1     0                              632
2     1                              89
3     2                              21
4     3                              8
```

```
5     4                              4
==================================================

==================================================
Top 10 most frequent values for column 'label':
Index Value                          Frequency
--------------------------------------------------
1     CE                             547
2     LAA                            207
==================================================

==================================================
Top 10 most frequent values for column 'image_path':
Index Value                          Frequency
--------------------------------------------------
1     ../input/mayo-clinic-strip-ai/ 1
2     ../input/mayo-clinic-strip-ai/ 1
3     ../input/mayo-clinic-strip-ai/ 1
4     ../input/mayo-clinic-strip-ai/ 1
5     ../input/mayo-clinic-strip-ai/ 1
6     ../input/mayo-clinic-strip-ai/ 1
7     ../input/mayo-clinic-strip-ai/ 1
8     ../input/mayo-clinic-strip-ai/ 1
9     ../input/mayo-clinic-strip-ai/ 1
10    ../input/mayo-clinic-strip-ai/ 1
==================================================

==================================================
Top 10 most frequent values for column 'image_size':
Index Value                          Frequency
--------------------------------------------------
1     (15496, 12017)                 2
2     (15410, 29532)                 2
3     (18772, 17529)                 2
4     (14355, 53736)                 1
5     (14368, 16628)                 1
6     (23078, 40562)                 1
7     (51843, 36423)                 1
8     (17545, 42426)                 1
9     (24121, 52541)                 1
10    (25276, 40548)                 1
==================================================

==================================================
Top 10 most frequent values for column 'width':
Index Value                          Frequency
--------------------------------------------------
1     37885                          6
2     51843                          5
3     10533                          5
4     12062                          4
5     75771                          4
6     73777                          4
7     55831                          4
8     7814                           4
9     18660                          3
```

```
10    83747                         3
==================================================

==================================================
Top 10 most frequent values for column 'height':
Index Value                         Frequency
--------------------------------------------------
1     7339                          3
2     29538                         3
3     12017                         2
4     36924                         2
5     33182                         2
6     17529                         2
7     35972                         2
8     35086                         2
9     31366                         2
10    38737                         2
==================================================
```

Observaciones 💡

> - Tal como se espera, el valor más frecuente en la columna `center_id` es 11, que aparece 257 veces, lo que indica que es el centro médico principal involucrado en la mayoría de los casos. Otros centros notables incluyen el 4, con 114 ocurrencias, y el 7, con 99.

> - En la columna `patient_id`, los IDs más frecuentes (por ejemplo, 91b9d3, 3d10be y 09644e) aparecen cada uno 5 veces, lo que indica que estos pacientes tienen múltiples imágenes tomadas para análisis. Esta repetición sugiere un posible enfoque en pacientes específicos con imágenes recurrentes, posiblemente debido a un seguimiento continuo de condiciones particulares.

> - Finalmente, la columna `image_num` muestra que 0 es el valor más común, con 632 ocurrencias, lo que sugiere que el conjunto de datos consiste principalmente en las primeras imágenes tomadas de varios pacientes. Los valores posteriores, con 1 apareciendo 89 veces y 2 solo 21 veces, indican que hay menos imágenes subsiguientes por paciente. Este patrón podría sugerir que la mayoría de los casos implican evaluaciones iniciales en lugar de imágenes de seguimiento.

(7) Análisis de Pacientes

In [ ]:
```python
# Count unique patients
unique_patient_count = df['patient_id'].nunique()

# Print the result neatly
print(f"Unique Patients Count: {unique_patient_count}")
```

```
Unique Patients Count: 632
```

Observaciones 💡

> - Como se ha mencionado anteriormente, solamente algunos pacientes tienen más de una imagen asociada a su persona, por lo que en nuestro conjunto de datos hay imágenes de 630 pacientes **únicos**.

(8) Análisis de Valores Duplicados

In [ ]:
```
# Check duplicate rows in dataset
df = df.drop_duplicates()
# Print the number of records in the DataFrame
print("The given dataset has", df.shape[0], "registers and", df.shape[1], "c
```

```
The given dataset has 754 registers and 9 columns.
```

Observaciones 💡

> - Vemos que no existen valores duplicados en el conjunto de datos.

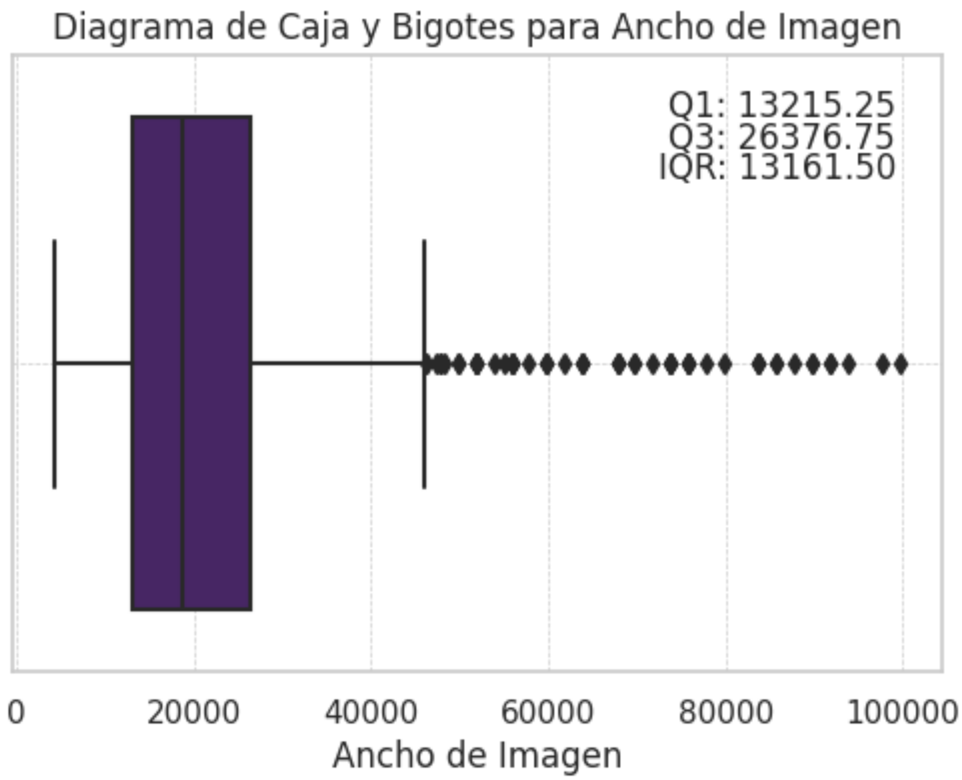
(9) Análisis de Datos Atípicos

In [ ]:
```
# Calculate quartiles
# Calculate the first quartile (Q1)
Q1 = df["width"].quantile(0.25)
# Calculate the third quartile (Q3)
Q3 = df["width"].quantile(0.75)
# Calculate the interquartile range (IQR)
IQR = Q3 - Q1

# Create the box and whisker plot with quartiles
sns.boxplot(x="width", data=df, palette=palette)
plt.title("Diagrama de Caja y Bigotes para Ancho de Imagen")
plt.xlabel("Ancho de Imagen")

# Add quartile information as text annotations
plt.text(0.95, 0.9, f"Q1: {Q1:.2f}", transform=plt.gca().transAxes, ha="righ
plt.text(0.95, 0.8, f"IQR: {IQR:.2f}", transform=plt.gca().transAxes, ha="ri
plt.text(0.95, 0.85, f"Q3: {Q3:.2f}", transform=plt.gca().transAxes, ha="rig

# Adding grid with custom style
plt.grid(True, linestyle='--', linewidth=0.5)  # Adding grid with dashed lin
plt.show()
```



Observaciones 💡

> - Se observa en las estadísticas del conjunto de datos para el ancho de las imágenes que la mediana (50%) es de 18,700 píxeles, lo que indica que la mitad de las imágenes tienen un ancho igual o menor a este valor.

> - El primer cuartil (25%) se encuentra en 13,215 píxeles y el tercer cuartil (75%) en 26,376.75 píxeles, lo que sugiere que la mayoría de las imágenes tienen un tamaño moderado, con un rango intercuartílico (IQR) de 13,161.5 píxeles.

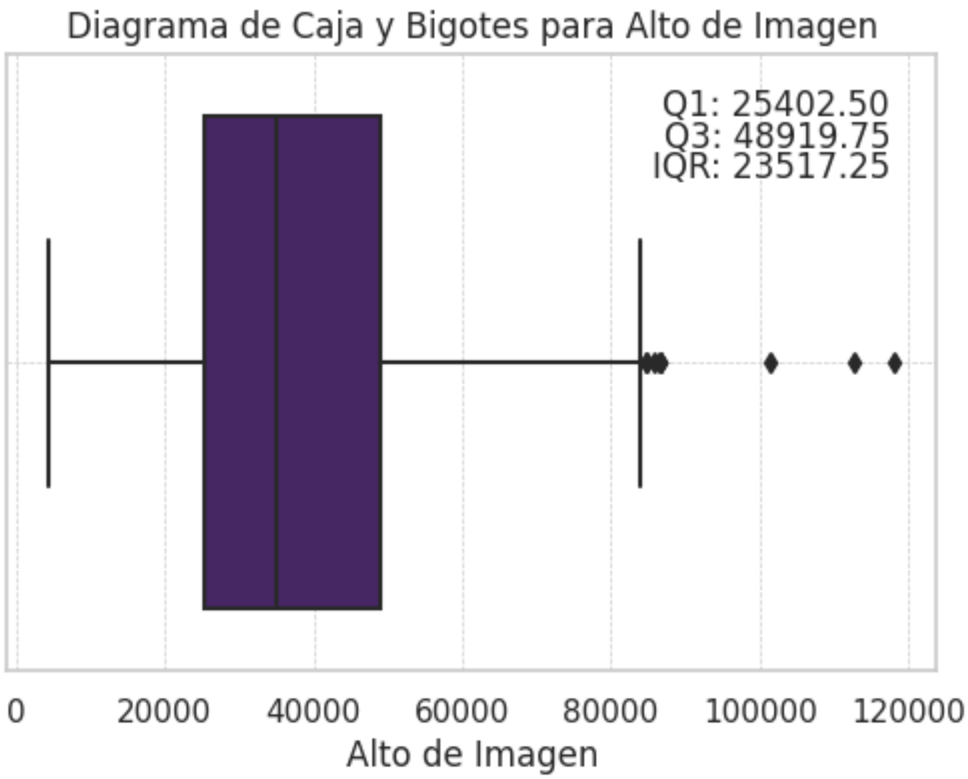
> - El valor máximo de 99,699 píxeles, así como la desviación estándar de 15,653.64 píxeles indica sugieren que existe una variabilidad notable entre los tamaños de las imágenes y revelan la existencia de imágenes significativamente más anchas en el conjunto

In [ ]:
```
# Calculate quartiles
# Calculate the first quartile (Q1)
Q1 = df["height"].quantile(0.25)
# Calculate the third quartile (Q3)
Q3 = df["height"].quantile(0.75)
# Calculate the interquartile range (IQR)
IQR = Q3 - Q1

# Create the box and whisker plot with quartiles
sns.boxplot(x="height", data=df, palette=palette)
plt.title("Diagrama de Caja y Bigotes para Alto de Imagen")
plt.xlabel("Alto de Imagen")
```

```
# Add quartile information as text annotations
plt.text(0.95, 0.9, f"Q1: {Q1:.2f}", transform=plt.gca().transAxes, ha="righ
plt.text(0.95, 0.8, f"IQR: {IQR:.2f}", transform=plt.gca().transAxes, ha="ri
plt.text(0.95, 0.85, f"Q3: {Q3:.2f}", transform=plt.gca().transAxes, ha="rig

# Adding grid with custom style
plt.grid(True, linestyle='--', linewidth=0.5)  # Adding grid with dashed lir
plt.show()
```



Observaciones 💡

> - Se observa que la mediana (50%) es de 34,981.5 píxeles, lo que indica que la mitad de las imágenes tiene una altura igual o menor a este valor.

> - El primer cuartil (25%) se sitúa en 25,402.5 píxeles y el tercer cuartil (75%) en 48,919.75 píxeles, sugiriendo que la mayoría de las imágenes poseen una altura moderada, con un rango intercuartílico (IQR) de 23,517.25 píxeles.

> - La media de 37,622.20 píxeles refleja la altura promedio de las imágenes, mientras que la desviación estándar de 18,058.75 píxeles indica una considerable dispersión en las alturas.

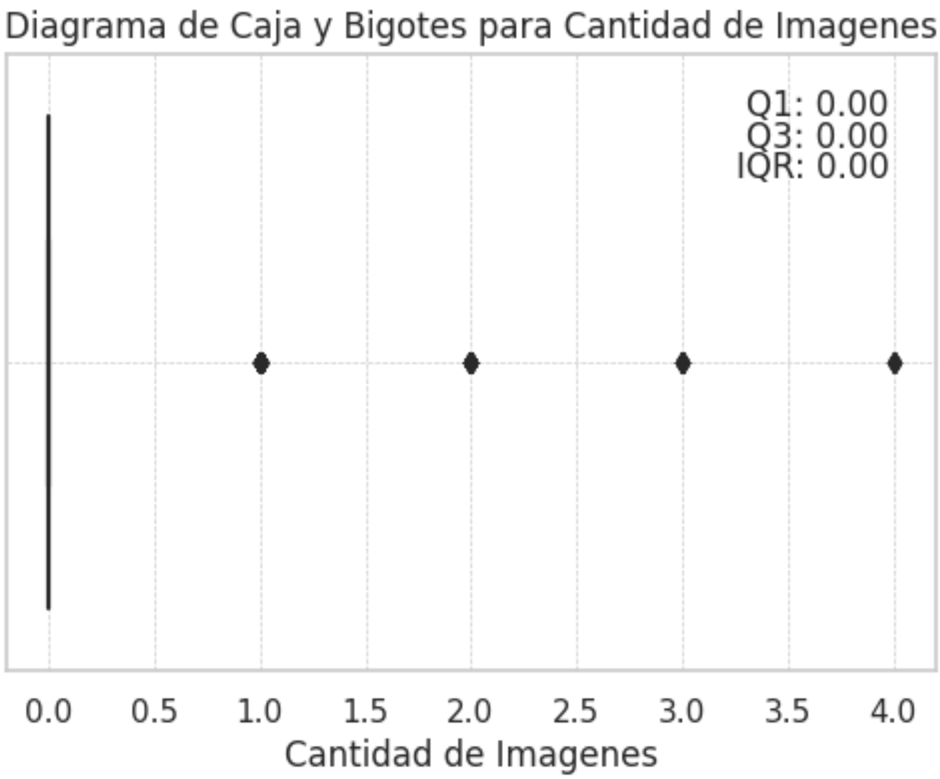
In [ ]:
```
# Calculate quartiles
# Calculate the first quartile (Q1)
Q1 = df["image_num"].quantile(0.25)
# Calculate the third quartile (Q3)
Q3 = df["image_num"].quantile(0.75)
```

```
# Calculate the interquartile range (IQR)
IQR = Q3 - Q1

# Create the box and whisker plot with quartiles
sns.boxplot(x="image_num", data=df, palette=palette)
plt.title("Diagrama de Caja y Bigotes para Cantidad de Imagenes")
plt.xlabel("Cantidad de Imagenes")

# Add quartile information as text annotations
plt.text(0.95, 0.9, f"Q1: {Q1:.2f}", transform=plt.gca().transAxes, ha="righ
plt.text(0.95, 0.8, f"IQR: {IQR:.2f}", transform=plt.gca().transAxes, ha="ri
plt.text(0.95, 0.85, f"Q3: {Q3:.2f}", transform=plt.gca().transAxes, ha="rig

# Adding grid with custom style
plt.grid(True, linestyle='--', linewidth=0.5)  # Adding grid with dashed lin
plt.show()
```



In [ ]:

```
# Define the limits to determine outliers
lower_limit = Q1 - 1.5 * IQR
upper_limit = Q3 + 1.5 * IQR

# Count the number of outliers
outliers = df[(df["image_num"] < lower_limit) | (df["image_num"] > upper_lim
num_outliers = len(outliers)

# Calculate the percentage of outliers
percentage_outliers = (num_outliers / len(df)) * 100

print("Outliers Percentage in 'image_num':", round(percentage_outliers, 2))
```

```
Outliers Percentage in 'image_num': 16.18
```

In [ ]:
```
print("\nTraining Set: 'image_num' Statistics")
print(df['image_num'].describe())
```

```
Training Set: 'image_num' Statistics
count    754.000000
mean       0.226790
std        0.599046
min        0.000000
25%        0.000000
50%        0.000000
75%        0.000000
max        4.000000
Name: image_num, dtype: float64
```

In [ ]:
```
# Create a cross-tabulation of 'center_id' and 'label'
cross_tab = pd.crosstab(df['image_num'], df['label'])

# Display the cross-tabulation table
cross_tab.head()
```

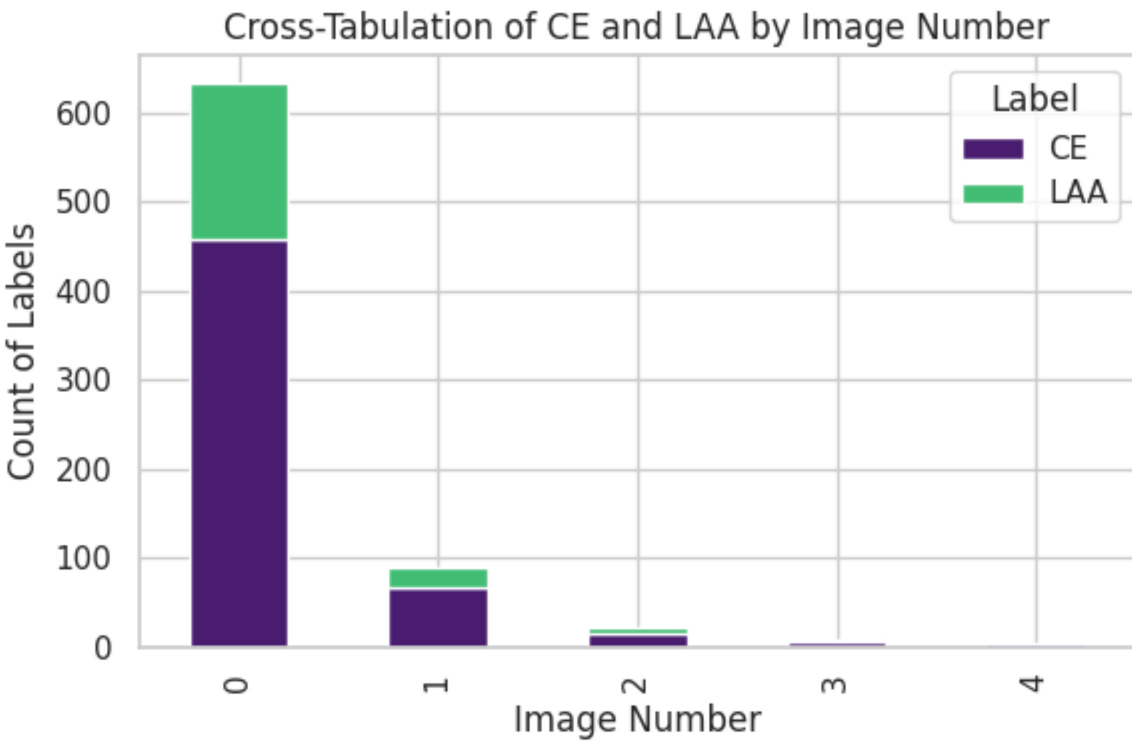
Out[ ]:

label image_num	CE	LAA
0	457	175
1	66	23
2	15	6
3	6	2
4	3	1

In [ ]:
```
# Plot a stacked bar plot for better visualization
cross_tab.plot(kind='bar', stacked=True, color=[palette[0], palette[8]])

# Adding titles and labels
plt.title('Cross-Tabulation of CE and LAA by Image Number')
plt.xlabel('Image Number')
plt.ylabel('Count of Labels')
plt.legend(title='Label', loc='upper right')

# Show plot
plt.tight_layout()
plt.show()
```



Observaciones 💡

> - El valor de todos los cuartiles es de 0, lo que indica que el conjunto de datos consiste principalmente en las primeras imágenes tomadas de varios pacientes. Sin embargo, el valor máximo de 4 indica que hay algunos pacientes que presentan hasta cuatro imágenes de coágulos, lo que introduce cierta variabilidad en el conjunto de datos.

> - Los puntos atípicos representan alrededor del 16.18% de la totalidad de los datos.

(4) Image Preprocessing 📷

In [ ]:
```
def preprocess_image(image_path, idx):
    print(f"Processing image at index {idx}")  # Print the current index
    # Read the image from the file
    img = tifi.imread(image_path)
    #img = cv2.imread(image_path)
    img = cv2.resize(img, (0,0), fx=0.05, fy=0.05)
    # Convert the image to grayscale (if required by your model)
    # If your CNN expects color images, skip this step
    img = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    # Reason: Convert to grayscale if the model is designed for single-chann
    # Normalize pixel values to the range [0, 1]
    img = img / 255.0
    # Reason: Normalization ensures that pixel values are in a consistent ra
    # Apply Gaussian blur to reduce noise (optional, depending on the noise
    img = cv2.GaussianBlur(img, (5, 5), 0)
```

```
    # Reason: Noise reduction can help the CNN by removing small details tha
    img = np.expand_dims(img, axis=-1)
    # Reason: Consistent image size is required for CNN input.
    return img
```

In [ ]:

```
# Apply the function
df['image'] = df.apply(lambda row: preprocess_image(row['image_path'], row.n
```

```
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Processing image at index 720
Processing image at index 721
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Observaciones 💡 -->

1. **Conversión a Escala de Grises**

   - **Código**: `img = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)`

- **Descripción**: Convierte la imagen de color (BGR) a una imagen en escala de grises.
- **Propósito**: Reducir la complejidad del modelo ya que se espera una entrada de un solo canal (grayscale). Permitirá que el entrenamiento sea más rápido.

2. **Normalización de Valores de Pixel**

   - **Código**: `img = img / 255.0`
   - **Descripción**: Normaliza los valores de los píxeles a un rango de [0, 1].
   - **Propósito**: Normalizar los valores de los píxeles asegura que estén en un rango consistente, lo que ayuda a mantener una iluminación y contraste uniformes en todas las imágenes. Esto facilita la comparación de patrones y características, haciendo que las imágenes sean más adecuadas para el análisis. Además, permite que los modelos de aprendizaje automático aprendan patrones de manera más efectiva, sin que las diferencias en las condiciones de iluminación afecten el rendimiento.

3. **Aplicación de Desenfoque Gaussiano**

   - **Código**: `img = cv2.GaussianBlur(img, (5, 5), 0)`
   - **Descripción**: Aplica un filtro de desenfoque gaussiano para reducir el ruido en la imagen.
   - **Propósito**: El desenfoque gaussiano reduce el detalle y el ruido en la imagen aplicando una función gaussiana a cada píxel y sus píxeles circundantes. Esto suaviza los bordes y elimina pequeños detalles que pueden no ser útiles para el aprendizaje, ayudando a la red neuronal a centrarse en las características más relevantes. La reducción de ruido mejora la capacidad de generalización del modelo y puede facilitar tareas como la detección de bordes o la segmentación al reducir las variaciones no deseadas en la imagen.

4. **Expansión de Dimensiones**

   - **Código**: `img = np.expand_dims(img, axis=-1)`
   - **Descripción**: Expande las dimensiones de la imagen para agregar un canal adicional.
   - **Propósito**: Asegura que la imagen tenga un formato consistente para la entrada de la red neuronal, especialmente si la red espera una entrada con un número específico de canales (por ejemplo, [alto, ancho, canales]).

5. Redimensionamiento de Imagen

   - **Código**: img = cv2.resize(img, (0,0), fx=0.05, fy=0.05)
   - **Descripción**: Redimensiona la imagen aplicando un factor de escala del 5% en los ejes horizontal (fx) y vertical (fy).
   - **Propósito**: Reduce el tamaño de la imagen para disminuir la cantidad de información que se procesa en la red neuronal convolucional (CNN), lo que puede ser útil para ahorrar recursos computacionales o ajustar la entrada a las dimensiones esperadas por el modelo.

Referencia

> - https://medium.com/@maahip1304/the-complete-guide-to-image-preprocessing-techniques-in-python-dca30804550c