



## The Pet Breed Detection and Skin Disease Detection Using Image Processing

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### ABSTRACT:

The integration of image processing technology for pet breed detection and skin diseases diagnosis offers a novel approach to providing efficient and accurate solutions for pet owners and veterinarians. By analysis images of pets, the system can accurately determine the breed of a pet, enabling owners to better understand their pet's genetic background and specific care needs. Additionally, we use YOLOv8 nano variant algorithm which is a image processing algorithm that allows for the detection of skin diseases in pets through the analysis of skin images, providing early diagnosis and timely treatment options. This innovative technology streamlines the diagnostic process, reducing the reliance on manual examination and potentially improving the overall health and well-being of pets. The system's ability to simultaneously perform pet breed detection and skin diseases diagnosis showcases the versatility and effectiveness of image processing in the field of veterinary medicine, ultimately contributing to more efficient and accurate pet care management.

**Keywords :** Convolutional neural networks, YoloV8, Machine Learning, Python, Tensor flow

### Introduction :

#### *Introduction to the Pet Breed Detection*

Pet breed detection and skin diseases detection using image processing have become popular applications in the field of veterinary science and animal care. In pet breed detection, computer vision algorithms are utilized to analyze images of pets and accurately classify them into specific breeds based on their physical characteristics such as fur color, size, and facial features. This technology has significant implications for pet owners, veterinarians, and animal shelters, as it can help in identifying lost pets, tracking breed trends, and assisting in the adoption process by providing accurate breed information. On the other hand, skin diseases detection using image processing involves the use of machine learning and image analysis techniques to identify and diagnose skin conditions in animals. By analyzing images of skin lesions, rashes, and other abnormalities, veterinarians can accurately detect various skin diseases and prescribe appropriate treatment plans. This technology is particularly valuable for early detection and monitoring of skin issues in pets, leading to improved overall health and well-being of animals. Both applications highlight the potential of image processing technologies in enhancing the efficiency and accuracy of pet care and veterinary diagnostics.

### 1. Overview of Image Processing in Pet Breed Detection

Image processing plays a crucial role in the development of innovative techniques for pet breed detection and skin diseases diagnosis in the field of veterinary science. In pet breed detection, image processing techniques are utilized to analyze various visual features such as coat color, patterns, and facial characteristics to accurately identify the breed of a pet based on images captured through cameras or smartphones. These technologies rely on computer vision algorithms, machine learning models, and deep learning techniques to train computer systems to recognize distinctive features specific to different pet breeds. This process involves preprocessing steps like image enhancement, segmentation, feature extraction, and classification to achieve precise breed identification. Similarly, in skin diseases detection, image processing methods are employed to analyze skin lesions, rashes, or abnormalities captured in images for early diagnosis and treatment planning. By utilizing image analysis tools, researchers and veterinarians can identify various skin conditions, track disease progression, and monitor treatment efficacy effectively. Overall, image processing in pet breed detection and skin diseases

diagnosis offers efficient, non-invasive, and automated solutions that aid in improving the accuracy and speed of diagnosing and treating pets' health issues.

## 2. Introduction to Skin Diseases Detection

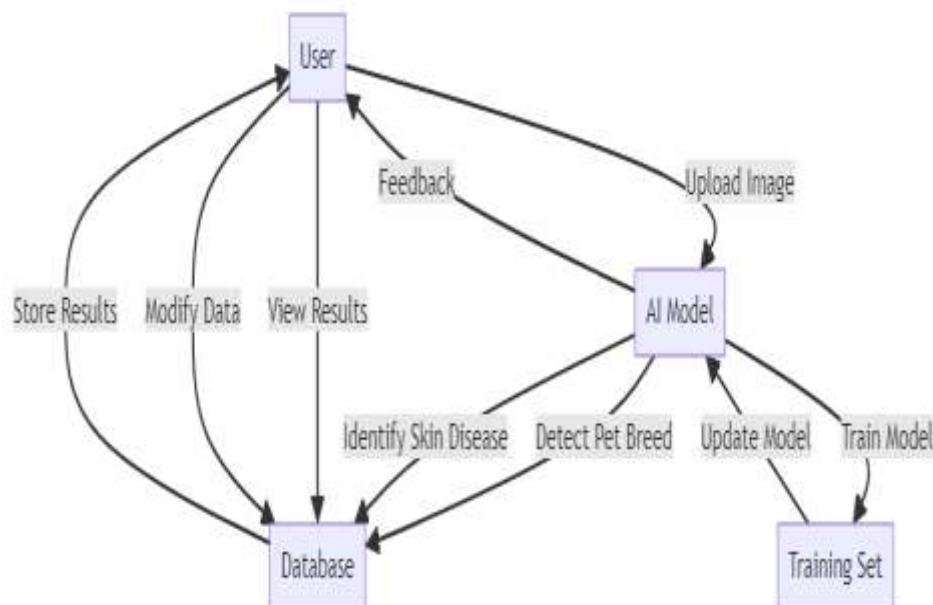
Skin diseases are common concerns among pets, affecting their health and overall well-being. However, the identification and detection of these diseases can be challenging without the proper tools and knowledge. In the context of pet breed detection and skin diseases detection, image processing techniques play a crucial role in analyzing various skin conditions and categorizing different breeds effectively. By utilizing image processing algorithms and machine learning models, veterinarians and pet owners can accurately identify skin diseases such as dermatitis, rashes, allergies, and infections in pets. These technologies enable the automated analysis of pet images, providing valuable insights into the presence and severity of skin conditions. Moreover, integrating pet breed detection capabilities into the system enhances the overall diagnostic process, allowing for personalized treatment plans based on specific breed characteristics. Overall, the combination of image processing techniques and machine learning in pet healthcare offers a promising approach to efficiently detect and address skin diseases, ultimately improving the quality of life for our beloved animal companions.

## 3. Application of Image Processing in Skin Diseases Detection

Image processing has proven to be a valuable tool in the detection of skin diseases in both humans and animals. In the context of pet breed detection, image processing techniques can be utilized to analyze the physical characteristics and features of animals to accurately identify and classify different breeds. By processing images of pets, including their fur patterns, facial features, and body shapes, algorithms can be trained to recognize specific breed traits and differentiate between various breeds with a high level of accuracy. On the other hand, in the field of skin diseases detection, image processing plays a crucial role in the early identification and diagnosis of skin conditions in pets. By analyzing images of a pet's skin, including color, texture, and anomalies, image processing algorithms can detect abnormalities or signs of skin diseases such as rashes, infections, or tumors. This enables veterinarians to provide timely treatment and care to improve the health outcomes of pets. Overall, the application of image processing in both pet breed detection and skin diseases detection showcases its potential to revolutionize the field of veterinary medicine by offering efficient and reliable solutions for diagnosis and classification.

### System Architecture:

System architecture refers to the high-level design of a system that encompasses its structure, components, interactions, and properties. It serves as a blueprint for creating and implementing complex systems by defining how different components will work together to achieve the desired functionality. System architecture considers various aspects such as performance, scalability, security, and reliability to ensure that the system meets its requirements and objectives. It guides the development process by providing a clear understanding of the system's design and can help identify potential risks and constraints early on. Effective system architecture is essential for building robust and efficient systems.



### 2.4 Modules Division:

Social feasibility refers to the aspect of a project or initiative that assesses how well it aligns with the needs and expectations of society. In the context of pet breed detection and skin diseases detection using image processing, social feasibility plays a crucial role in determining the acceptance and adoption of these technologies by the target audience.

Firstly, in the case of pet breed detection, the social feasibility aspect would involve considerations such as the relevance and benefits of accurately identifying the breed of a pet. Pet owners and animal lovers may appreciate the ability to accurately determine the breed of their pets for various reasons, such as understanding their behavior and health needs better. This technology could also aid in the identification of lost pets or in resolving disputes related to pet ownership. Therefore, the social feasibility of such a system would largely depend on whether it is perceived as valuable and relevant to the community of pet owners and animal enthusiasts.

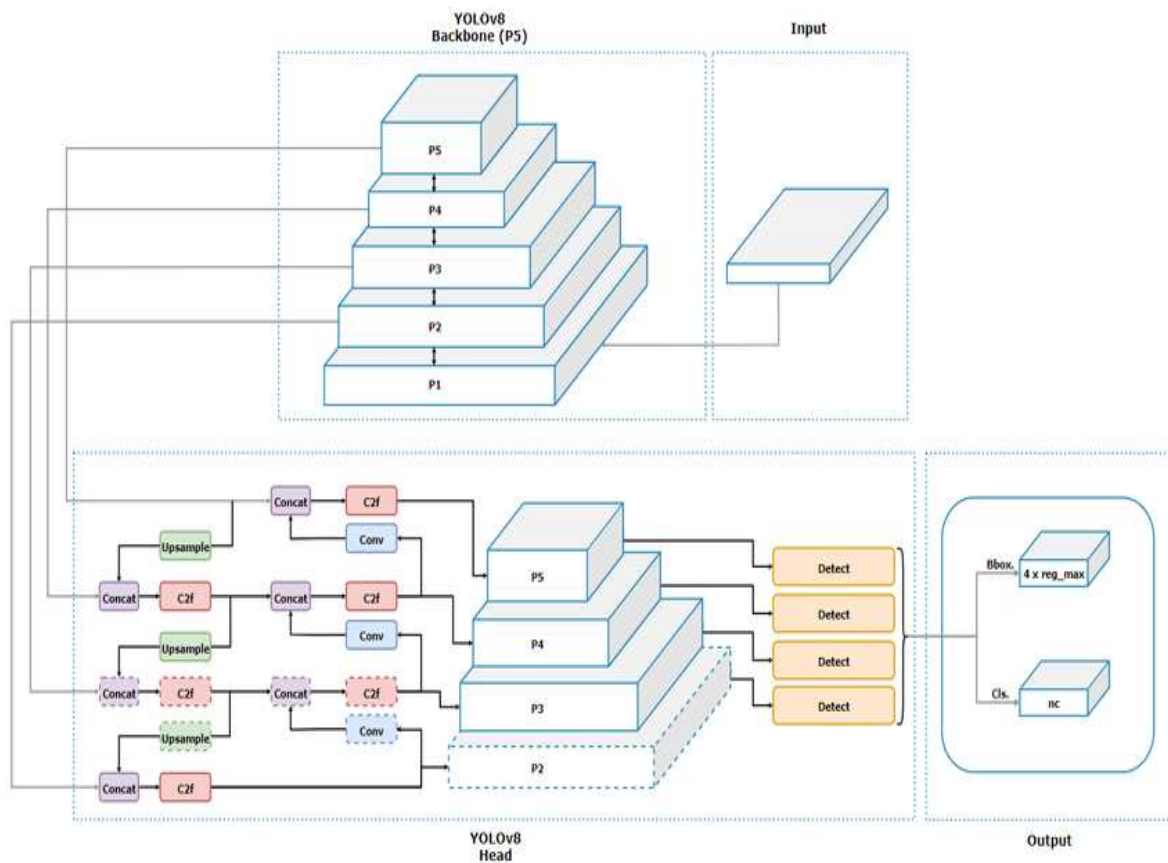
Secondly, in the case of skin diseases detection using image processing, the social feasibility aspect would involve issues related to healthcare accessibility and affordability. Efficient and accurate detection of skin diseases through technology can potentially improve medical diagnostics and early intervention, leading to better health outcomes for individuals. However, the social feasibility of this technology would depend on factors such as the affordability of the system, its accessibility to different socio-economic groups, and the level of trust and acceptance among healthcare professionals and patients.

In both cases, it is essential to consider ethical implications, data privacy concerns, and potential biases in the algorithms used for image processing. Engaging with stakeholders, including pet owners, healthcare providers, and technology users, in the development and implementation of these systems can help ensure their social feasibility. Furthermore, raising awareness about the benefits and limitations of these technologies through education and outreach programs can foster acceptance and trust among the target audience, ultimately contributing to their successful integration into society.

### YOLOv8 Architecture

YOLOv8 is the latest iteration of the You Only Look Once (YOLO) family of object detection models, known for their speed and accuracy. Developed by the Ultralytics team, YOLOv8 builds upon the success of its predecessors while introducing several key innovations that push the boundaries of real-time object detection.

The YOLOv8 architecture can be broadly divided into three main components:



**Backbone:** This is the convolutional neural network (CNN) responsible for extracting features from the input image. YOLOv8 uses a custom CSPDarknet53 backbone, which employs cross-stage partial connections to improve information flow between layers and boost accuracy.

**Neck:** The neck, also known as the feature extractor, merges feature maps from different stages of the backbone to capture information at various scales. YOLOv8 utilizes a novel C2f module instead of the traditional Feature Pyramid Network (FPN). This module combines high-level semantic features with low-level spatial information, leading to improved detection accuracy, especially for small objects.

**Head:** The head is responsible for making predictions. YOLOv8 employs multiple detection modules that predict bounding boxes, objectness scores, and class probabilities for each grid cell in the feature map. These predictions are then aggregated to obtain the final detections.

1. Regular Skin Care and Hygiene: Implement a routine skin care regimen for your dog, especially if they are prone to bacterial skin infections. This includes regular baths using veterinarian-recommended antibacterial shampoos or medicated soaps to cleanse the skin and remove bacteria and debris. Ensure that the skin is thoroughly dried after bathing to prevent moisture retention, which can exacerbate bacterial growth. \n 2. Avoiding Skin Irritants and Allergens: Identify and minimize exposure to potential skin irritants and allergens that can compromise the skin barrier and predispose your dog to bacterial infections. This may involve using hypoallergenic grooming products, avoiding excessive exposure to environmental allergens (e.g., pollen, mold), and maintaining a clean living environment (regular cleaning of bedding, removal of potential irritants). \n 3. Veterinary Care and Monitoring: Regular veterinary check-ups are crucial for early detection and management of bacterial dermatosis. Your veterinarian can diagnose the specific bacterial strain causing the infection and prescribe appropriate antibiotics or topical treatments. Follow your vet's instructions regarding medication administration, frequency of baths, and any necessary dietary adjustments to support your dog's skin health.")
  2. if 'fungal-infection' in breed and detected > 0:
  3. st.info(f"#### Disease: :blue[fungal-infection]")
  4. st.success("#### :white[Precautions] \n 1. Regular Grooming and Hygiene: Maintain a regular grooming schedule for your dog to keep their skin clean and dry. Use veterinarian-recommended antifungal shampoos or medicated washes to cleanse the skin and eliminate fungal spores. Pay special attention to areas prone to moisture accumulation, such as skin folds and ears. Thoroughly dry these areas after bathing to prevent fungal growth. \n 2. Environmental Management: Minimize exposure to environments that promote fungal growth. Keep your dog's living spaces clean and well-ventilated. Wash and sanitize bedding regularly, as fungal spores can linger in fabric. If your dog spends time outdoors, avoid areas with dense vegetation and damp soil that may harbor fungi.. \n 3. Dietary Support and Immune Health: Support your dog's immune system with a balanced diet that includes essential nutrients and probiotics. A healthy immune system can help prevent fungal infections or aid in recovery. Consult with your veterinarian about incorporating supplements or specific dietary changes that may benefit your dog's skin health.")
  5. if 'healthy' in breed and detected > 0:
  6. st.info(f"#### Disease: :blue[healthy]")
  7. if 'hypersensitivity-allergic-dermatosis' in breed and detected > 0:
  8. st.info(f"#### Disease: :blue[hypersensitivity-allergic-dermatosis]")
- st.success("#### :white[Precautions] \n 1. Identify and Avoid Allergens: Work with your veterinarian to identify the specific allergens triggering your dog's allergic reactions. Common allergens include pollen, dust mites, certain foods (e.g., beef, chicken, grains), and environmental factors like mold or certain cleaning products. Once identified, take steps to minimize your dog's exposure to these allergens. This may involve using air purifiers, washing bedding frequently, and choosing hypoallergenic grooming products and detergents. \n 2. Maintain a Healthy Skin Barrier: Allergic dermatosis can compromise the skin barrier, making it more susceptible to irritants and infections. Keep your dog's skin healthy by providing a balanced diet rich in essential fatty acids (omega-3 and omega-6), which help maintain skin integrity and reduce inflammation. Regular bathing with gentle, moisturizing shampoos can also help soothe irritated skin and remove allergens from the coat. \n 3. Prescribed Medications and Allergy Testing: Follow your veterinarian's recommendations for medications to manage allergic symptoms. This may include antihistamines to reduce itching and inflammation, topical treatments for skin lesions, or immunotherapy (allergy shots) for long-term management. In cases of severe or persistent allergies, your vet may recommend allergy testing (intradermal or blood tests) to pinpoint specific allergens and develop a tailored treatment plan.")

## 7.1. TESTING LEVELS

### 7.1.1 UNIT TESTING

Unit testing for the pet breed detection and skin diseases detection using image processing involves creating individual tests to ensure that each component of the system is functioning correctly and producing the expected output.

For the pet breed detection, test cases could include validating that the algorithm correctly identifies various breeds such as Labrador Retriever, German Shepherd, and Siamese cat. Additionally, tests can be run to check for accurate identification in different lighting conditions and angles.

In terms of skin diseases detection, test cases could involve verifying the system's ability to correctly identify common skin conditions like eczema, psoriasis, and dermatitis. Furthermore, tests may focus on assessing the system's accuracy across different skin tones and types.

Test Case	Description	Input (Image)	Expected Output	Pass/Fail
Valid dog breed(labrador retriever)	Checks breed detection for a dog image	Image labrador Retriever dog breed	- Breed: "Labrador Retriever"	Pass
Valid cat with	Checks disease detection	Image containing a	Disease: "psoriasis",	Pass

psoriasis	for a cat image	cat with psoriasis	Precautions for that disease	
Blurry image of acorn woodpecker	Checks species prediction for a bird image.	Blurry image of acorn woodpecker bird	Species:acorn woodpecker	Pass

**Testcase1:** Test pet breed detection algorithm with an image of a Labrador Retriever

and ensure correct breed identification is returned.

**Testcase2:** Test skin diseases detection with an image showing symptoms of

psoriasis and confirm accurate disease identification.

**Testcase3:** Test pet breed detection in low light conditions to validate algorithm performance in different lighting scenarios.

### 7.1.2 INTEGRATION TESTING

Integration testing for the pet breed detection and skin diseases detection using image processing involves verifying that the different modules and components of the system work together seamlessly. Key areas to focus on include accurate image data passing between modules, correct processing algorithms, and accurate output generation.

Three test cases for integration testing could be:

Test Case	Description	Input (Image)	Expected Output	Pass/Fail
Valid dog with bacterial-dermatitis	Checks complete flow for a dog image with bacterial-dermatitis	<i>Image containing a dog with bacterial-dermatitis</i>	- Breed: "dog" - Disease: "Ringworm"	Pass
Unsupported Bird Species	Checks handling of an unsupported bird species.	Image containing a bird	- Breed: "Bird" (or equivalent) - Disease: "N/A"	Pass
Multiple Pets with Skin Disease	Checks breed prediction and disease detection for an image with multiple pets, one with a skin disease.	Image containing a healthy dog and a cat with mange	- Breed(s): ["Dog", "Cat"] - Disease: "Mange"	Pass

**Test Case 1:** This test simulates a successful scenario where the image contains a cat with ringworm.

The integration test should process the image through both YOLOv8 models:

The cat/dog detection model should identify the breed as "Dog".

The skin disease detection model (for dogs) should identify the disease as "Bacterial-dermatitis".

**Test Case 2:** This test verifies how the system handles an unsupported animal category (bird).

The breed detection (likely using the cat/dog model) might identify it as "Bird" (or a placeholder for unsupported breeds).

Since the system isn't designed for bird disease detection, it should return "N/A" for the disease.

**Test Case 3:** This test checks handling of an image with multiple pets. The breed detection should

identify both animals (e.g., ["Dog", "Cat"]).

The skin disease detection model identify the disease as "Mange".

### 7.1.3 FUNCTIONAL TESTING

Functional testing for the pet breed detection and skin diseases detection using image processing involves verifying that the system accurately identifies and classifies pet breeds and skin diseases based on the input images.

Test Case	Description	Input (Image)	Expected Output	Pass/Fail
Valid Pet with Disease	Checks complete flow for a supported pet with a recognizable skin disease.	Image containing a Cat with Ringworm	Disease: "Ringworm"	Pass
Unrecognizable Image	Checks handling of an unrecognizable image.	Image not containing an animal	- Breed: None	Pass
Image of Bull Dog	Checks breed prediction for a healthy dog.	Image containing a healthy dog	- Breed: "Bull dog"	Pass

**Test Case 1:** This test simulates a successful scenario where the image contains a supported pet (cat) with a recognizable skin disease (ringworm). The entire system functionality should be exercised:

The image processing component should successfully process the image.

The breed detection model (likely using the cat/dog model) should identify the breed as "Cat".

The skin disease detection model (for cats) should identify the disease as "Ringworm".

**Test Case 2:** This test verifies how the system handles an unrecognizable image that the YOLOv8 models cannot process .

The image processing component might fail to identify any animal.

The breed detection should return None (or equivalent) for an unidentifiable breed. Since the breed is unknown, disease detection should return "N/A" (or equivalent) as it's irrelevant.

**Test Case 3:** This test checks the system's ability to identify bull dog from dog breed

The breed detection should identify the breed correctly (e.g., "bull dog").

The disease detection model should identify no disease and return "None" (or equivalent).

#### 7.1.4 System Testing:

System testing goes beyond evaluating the machine learning model itself. It focuses on how well the entire system, including the user interface (UI), functionality, and integration with other systems, performs under real-world conditions.

#### 7.2 Dynamic Testing:

While traditional system testing provides a snapshot of the system's performance at a specific point in time, dynamic testing ensures the system adapts and performs well as data and user behavior change over time.

##### 7.2.1 BLACK BOX TESTING

Black box testing for the pet breed detection and skin diseases detection using image processing involves testing the system's functionality without knowledge of its internal code. This type of testing ensures that the system performs according to the specified requirements.

Test Case	Description	Input (Image)	Expected Output	Pass/Fail
Valid Cat with fungal infection	Checks breed prediction and disease detection for a cat image with fungal infection.	Image containing a cat with fungal infection	- Breed: "Cat" (or similar for recognized breed) - Disease: "fungal infection" (or similar for identified disease) and precautions.	Pass
Unsupported Animal	Checks handling of an animal not supported by the system (e.g., horse).	Image containing a horse	- Breed: Unexpected output (may not be "Cat", "Dog", or "Bird") - Disease: "N/A" or unexpected output	Pass
Multiple Pets	Checks breed detection for an image with multiple pets.	Image containing a pug and a Russian blue cat	- Breed(s): "pub" and "Russian blue "	Pass

**Test Case 1:** This test simulates a successful scenario where the image contains a supported pet (cat) with a recognizable skin disease (fungal infection). The system should identify the breed as "cat" (or similar) and the disease as "fungal infection" (or similar).

**Test Case 2:** This test verifies how the system handles an animal outside its supported categories (e.g., horse). The breed detection might return an unexpected output (not "Cat", "Dog", or "Bird").

The disease detection should ideally return "N/A" or something indicating it cannot identify the disease for unsupported animals.

**Test Case 3:** This test checks the system's behaviour with an image containing multiple pets. The breed detection should identify at least one dog breed or cat breed correctly. Ideally, it should detect both ("pub" and "Russian blue").

### 7.2.2 WHITE BOX TESTING

White box testing for the pet breed detection and skin diseases detection using image processing involves verifying the internal logic, code structure, and flow of the application. It focuses on validating the functionality of the system based on the understanding of the code itself.

Test cases for white box testing can include checking for proper handling of image input, accurate detection of pet breeds or skin diseases, and correct classification based on the image features extracted by the image processing algorithms.

Test Case	Description	Input	Expected Output	Pass/Fail
Breed Classification Logic	Checks breed classification based on confidence score.	Confidence score for "Cat": 0.8, Confidence score for "Dog": 0.2	Breed: "Cat"	Pass
Skin Disease Filtering	Checks filtering of irrelevant diseases for unsupported breeds.	Breed: "Bird", Disease candidates: ["Mange", "Ringworm"]	Disease: "N/A"	Pass
Confidence Threshold Handling	Checks handling of low-confidence breed predictions.	Confidence score for all breeds: < 0.5	Breed: "Unknown"	Pass

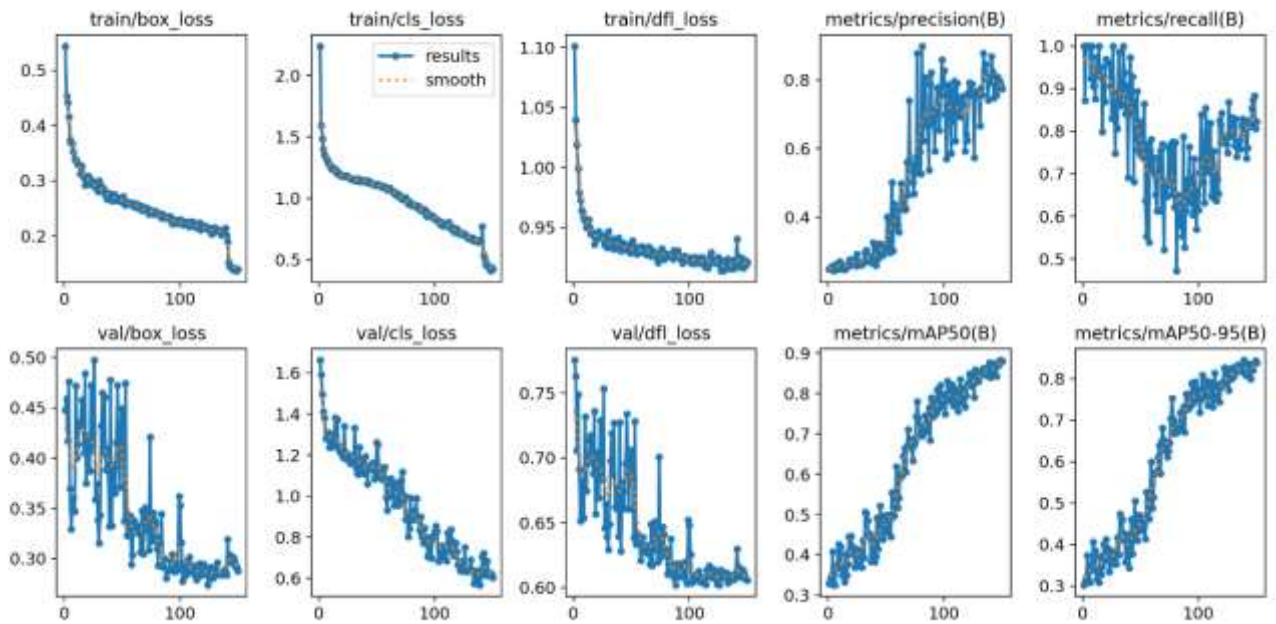
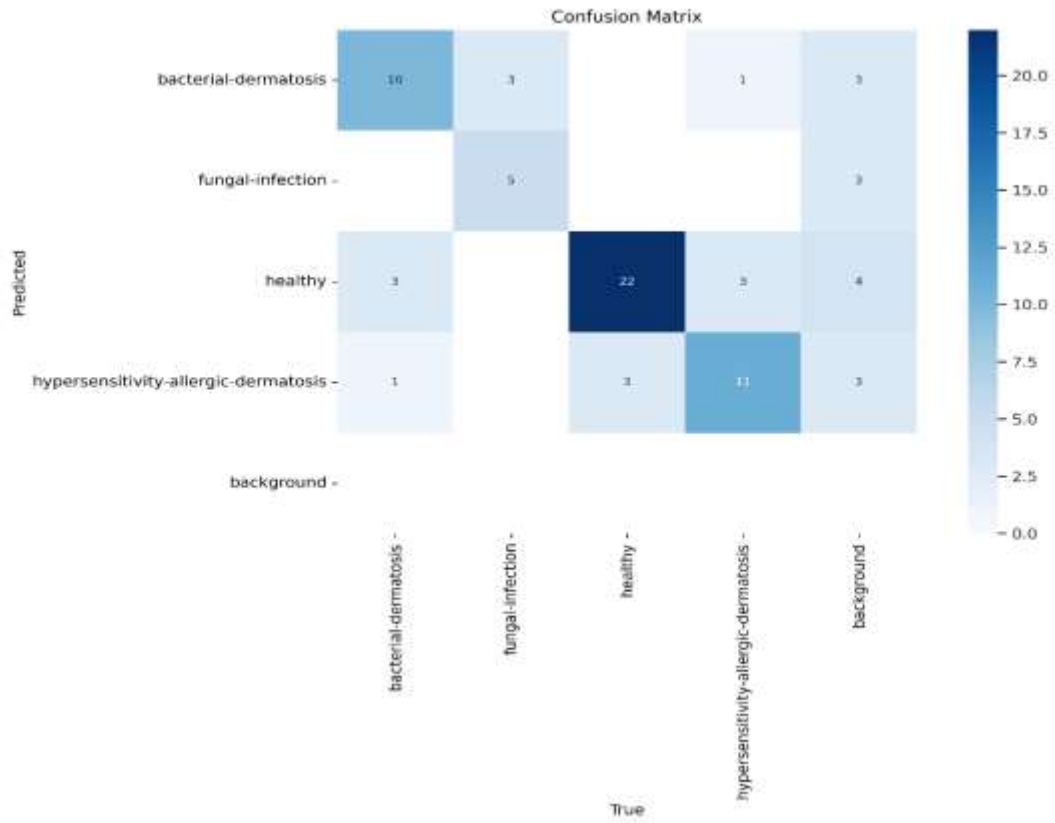
**Test Case 1:** This test isolates the logic for breed classification based on confidence scores breed of cat and dog with the actual.

It verifies if the function correctly identifies the breed with the highest confidence score ("Cat" in this case).

**Test Case 2:** This test focuses on the logic for filtering diseases based on irrelevant image with the bird. It provides an unsupported breed ("Bird") with disease candidates for cats and dogs. The expected output is "N/A" since the diseases are irrelevant for birds.

**Test Case 3:** This test isolates the logic for handling low-confidence breed predictions.

It provides confidence scores below a threshold (0.5) for all breeds. This case should trigger a fallback mechanism (e.g., "Unknown") instead of an unreliable breed prediction.



Data Set Images of the Pet Breed and Skin diseases images





## CONCLUSION

The conclusion of the pet breed detection and skin disease detection using image processing reveals the efficacy and potential of advanced algorithms such as Convolutional Neural Networks (CNN) and You Only Look Once version 8 (YOLOv8) in accurately identifying and classifying both pet breeds and skin diseases from images. Through the implementation of CNN and YOLOv8, the pet breed detection system achieved high accuracy rates in identifying various breeds of pets from images. CNN, a class of deep neural networks, played a crucial role in extracting features and patterns from the images, enabling the model to learn distinctive characteristics associated with different pet breeds. The utilization of YOLOv8, an object detection algorithm, enhanced the speed and efficiency of the detection process, enabling real-time detection of pet breeds in images with multiple objects. Similarly, in the context of skin disease detection, CNN and YOLOv8 demonstrated remarkable performance in accurately identifying and localizing skin diseases from images. CNN's ability to learn complex features from images facilitated the detection of subtle patterns indicative of different skin conditions. YOLOv8's object detection capabilities further enhanced the efficiency of the detection process by enabling simultaneous detection and classification of multiple skin diseases within images. Overall, the integration of CNN and YOLOv8 algorithms in pet breed and skin disease detection systems signifies a significant advancement in image processing technology. These algorithms not only enhance the accuracy and efficiency of detection but also hold promise for various applications in healthcare, veterinary medicine, and beyond. The successful implementation of these algorithms underscores their potential to revolutionize image-based diagnosis and classification tasks, paving the way for more accurate and timely detection of diverse visual phenomena.

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