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# Pembuatan model kode 1
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# 1. IMPOR PUSTAKA
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import os
import random
import shutil
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from PIL import Image, ImageStat
from pathlib import Path
import warnings
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import tensorflow as tf
from zipfile import ZipFile
import json
from scipy import stats
from sklearn.metrics import classification_report, confusion_matrix,
cohen_kappa_score
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# 2. KONFIGURASI GLOBAL
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warnings.filterwarnings("ignore")
plt.rcParams.update({'font.size': 10})
sns.set_style("whitegrid")
COLORS = sns.color_palette("Set2", 12)
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CONFIG = {
    "DATASET_DIR": "/content/Garbage_classification",
    "FLATTENED_DIR": "/content/Garbage_classification_flattened",
    "IMG_SIZE": (224, 224),
    "BATCH_SIZE": 32,
    "VAL_SPLIT": 0.2,
    "SEED": 42,
    "QUALITY_SAMPLE": 50,
    "VIS_SAMPLES_PER_CLASS": 4,
    "OUTPUT_DIR": "/content/output_results"
}
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# Set semua seed untuk reproducibility
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random.seed(CONFIG["SEED"])
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np.random.seed(CONFIG["SEED"])
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tf.random.set_seed(CONFIG["SEED"])
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# Buat direktori output
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Path(CONFIG["OUTPUT_DIR"]).mkdir(exist_ok=True)

print("=" * 75)
print("EKSPLORASI DAN PREPROCESSING DATASET – KLARIFIKASI SAMPAH")
print("=" * 75)
for k, v in CONFIG.items():
    if k != "FLATTENED_DIR":
        print(f"{k.replace('_', ' ').title():>25}: {v}")
print("=" * 75)

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# 3. FUNGSI UTILITAS
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def flatten_dataset_structure(src_root: str, dst_root: str):
    """Memindahkan semua gambar dari subfolder ke dalam 2 folder utama"""
    src_root = Path(src_root)
    dst_root = Path(dst_root)
    dst_root.mkdir(exist_ok=True)

    parent_classes = ["Organik", "anorganik"]
    for parent in parent_classes:
        (dst_root / parent).mkdir(exist_ok=True)
        parent_path = src_root / parent
        if not parent_path.exists():
            raise FileNotFoundError(f"Folder {parent_path} tidak ditemukan!")

        for subfolder in sorted(parent_path.iterdir()):
            if subfolder.is_dir():
                print(f"Memproses: {parent}/{subfolder.name} → {parent}/")
                for img_file in subfolder.iterdir():
                    if img_file.suffix.lower() in ['.jpg', '.jpeg', '.png',
                    '.bmp']:
                        new_name = f"{subfolder.name}_{img_file.name}"
                        shutil.copy2(img_file, dst_root / parent / new_name)

    print(f"Dataset telah diratakan ke: {dst_root}")

def analyze_image_quality(df_meta: pd.DataFrame, n_samples: int = 50) ->
pd.DataFrame:
    """Analisis kualitas gambar"""
    quality_data = []
    for _, row in df_meta.iterrows():
        img_files = [f for f in os.listdir(row['path'])
                    if f.lower().endswith(('.png', '.jpg', '.jpeg',
                    '.bmp'))]
        sampled = random.sample(img_files, min(n_samples, len(img_files)))
        for fname in sampled:
            fpath = os.path.join(row['path'], fname)
            try:
                with Image.open(fpath) as img:
                    w, h = img.size
                    if w == 0 or h == 0: continue
                    fs_kb = os.path.getsize(fpath) / 1024

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        stat = ImageStat.Stat(img.convert('RGB'))
        brightness = np.mean(stat.mean)
        ar = w / h
        quality_data.append({
            'category': row['category'],
            'parent': row['parent'],
            'width': w,
            'height': h,
            'aspect_ratio': ar,
            'file_size_kb': fs_kb,
            'brightness': brightness
        })
    except:
        continue
return pd.DataFrame(quality_data)

def visualize_sample_images(df_meta: pd.DataFrame, n_per_class: int = 4):
    """Visualisasi contoh gambar per kelas"""
    n_classes = len(df_meta)
    fig, axes = plt.subplots(n_classes, n_per_class, figsize=(n_per_class *
2, n_classes * 2))
    if n_classes == 1:
        axes = axes.reshape(1, -1)
    elif n_per_class == 1:
        axes = axes.reshape(-1, 1)

    for i, (_, rec) in enumerate(df_meta.iterrows()):
        img_files = [f for f in os.listdir(rec['path'])
            if f.lower().endswith(('.png', '.jpg', '.jpeg',
'.bmp'))]
        selected = random.sample(img_files, min(n_per_class, len(img_files)))
        for j, img_name in enumerate(selected):
            ax = axes[i, j] if n_classes > 1 and n_per_class > 1 else
(axes[j] if n_classes == 1 else axes[i])
            try:
                img = Image.open(os.path.join(rec['path'], img_name))
                ax.imshow(img)

ax.set_title(f"{rec['category']}\n{img.size[0]}x{img.size[1]}", fontsize=9,
fontweight='bold')
            except:
                ax.text(0.5, 0.5, 'Error', ha='center', va='center')
                ax.axis('off')
    plt.suptitle("Contoh Gambar per Kelas", fontweight='bold')
    plt.tight_layout(rect=[0, 0, 1, 0.97])
    plt.show()

def get_class_counts(gen):
    """Menghitung jumlah sampel per kelas dalam generator"""
    gen.reset()
    steps = (gen.samples + gen.batch_size - 1) // gen.batch_size
    counts = np.zeros(gen.num_classes, dtype=int)
    for _ in range(steps):
        try:
            _, y = next(gen)
            for label in y:
                counts[np.argmax(label)] += 1

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        except:
            break
    gen.reset()
    return counts

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# 4. EKSTRAKSI DAN FLATTEN DATASET
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zip_path = "/content/Garbage_classification.zip"
with ZipFile(zip_path, 'r') as zip_ref:
    zip_ref.extractall("/content")

print("Dataset berhasil diekstrak")

flatten_dataset_structure(CONFIG["DATASET_DIR"], CONFIG["FLATTENED_DIR"])

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# 5. PEMETAAN STRUKTUR DATASET & VERIFIKASI INTEGRITAS
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print("\n[1/6] Memetakan struktur dataset asli dan memverifikasi integritas file...")

dataset_root = Path(CONFIG["DATASET_DIR"])
parent_classes = ["Organik", "anorganik"]
subfolder_info = []

for parent in parent_classes:
    parent_path = dataset_root / parent
    if not parent_path.exists():
        raise FileNotFoundError(f"Folder {parent_path} tidak ditemukan!")

    for subfolder in sorted(parent_path.iterdir()):
        if subfolder.is_dir():
            valid = 0
            corrupt = 0
            for img_file in subfolder.iterdir():
                if img_file.suffix.lower() in ['.jpg', '.jpeg', '.png',
'.bmp']:
                    try:
                        with Image.open(img_file) as img:
                            img.verify()
                            valid += 1
                    except:
                        corrupt += 1
            subfolder_info.append({
                'parent': parent,
                'category': subfolder.name,
                'path': str(subfolder),
                'valid': valid,

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        'corrupt': corrupt
    })
    print(f"    {parent}/")
    print(f"        {subfolder.name}: {valid} valid, {corrupt} corrupt")

df_meta = pd.DataFrame(subfolder_info)
df_meta['count'] = df_meta['valid']
total_valid = df_meta['count'].sum()
print(f"\nTotal gambar valid: {total_valid}")

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# 6. ANALISIS DISTRIBUSI KELAS – STATISTIK & VISUALISASI
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print("\n" + "="*60)
print("ANALISIS DISTRIBUSI KELAS – STATISTIK DESKRIPTIF")
print("="*60)

counts = df_meta['count'].values
categories = df_meta['category'].values

mean_val = np.mean(counts)
median_val = np.median(counts)
std_val = np.std(counts)
cv_pct = (std_val / mean_val) * 100 if mean_val > 0 else 0
skewness = (3 * (mean_val - median_val)) / std_val if std_val != 0 else 0

descriptive_stats = {
    "Jumlah Kelas": len(counts),
    "Total Gambar": np.sum(counts),
    "Rata-rata": mean_val,
    "Median": median_val,
    "Standar Deviasi": std_val,
    "Koefisien Variasi (%)": cv_pct,
    "Skewness (Pearson)": skewness,
    "Minimum": np.min(counts),
    "Maksimum": np.max(counts),
    "Rentang": np.max(counts) - np.min(counts),
    "Rasio Imbalance (Max/Min)": np.max(counts) / np.min(counts) if
np.min(counts) > 0 else 0
}

for k, v in descriptive_stats.items():
    if isinstance(v, float):
        if "Koefisien" in k:
            print(f"{k:<30}: {v:>8.2f}%")
        elif "Rasio" in k:
            print(f"{k:<30}: {v:>8.2f}x")
        else:
            print(f"{k:<30}: {v:>8.2f}")
    else:
        print(f"{k:<30}: {v:>8}")

fig, ax = plt.subplots(figsize=(8, 8))

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wedges, texts, autotexts = ax.pie(
    counts,
    labels=categories,
    autopct=lambda pct: f"{int(pct/100.*sum(counts))}",
    startangle=90,
    colors=COLORS[:len(categories)],
    textprops={'fontweight': 'bold'},
    wedgeprops={'edgecolor': 'white', 'linewidth': 1.5, 'alpha': 0.9}
)
ax.set_title("Distribusi Jumlah Gambar per Kelas", fontweight='bold',
fontsize=13)
plt.tight_layout()
plt.show()

parent_summary =
df_meta.groupby('parent')['count'].sum().sort_values(ascending=False)
fig, ax = plt.subplots(figsize=(6, 4))
bars = ax.bar(parent_summary.index, parent_summary.values,
              color=['#6baed6', '#fdae6b'][:len(parent_summary)],
              edgecolor='black')
ax.set_ylabel("Jumlah Gambar", fontweight='bold')
ax.set_title("Distribusi: Organik vs Anorganik", fontweight='bold', pad=20)
for bar, count in zip(bars, parent_summary.values):
    pct = count / parent_summary.sum() * 100
    ax.text(bar.get_x() + bar.get_width()/2, bar.get_height() + 20,
            f"{int(count)}\n({pct:.1f}%)", ha='center', fontweight='bold',
            fontsize=9.5)
plt.tight_layout()
plt.show()

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# 7. ANALISIS KUALITAS GAMBAR
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print("\n[2/6] Analisis kualitas gambar...")
df_quality = analyze_image_quality(df_meta, CONFIG["QUALITY_SAMPLE"])

if not df_quality.empty:
    print("\n" + "="*60)
    print("ANALISIS KUALITAS GAMBAR - STATISTIK")
    print("="*60)
    desc = df_quality[['width', 'height', 'aspect_ratio', 'file_size_kb',
'brightness']].describe()
    print(desc.round(2))

fig, axes = plt.subplots(2, 3, figsize=(15, 8))
metrics = ['width', 'height', 'aspect_ratio', 'file_size_kb', 'brightness']
titles = ['Lebar (px)', 'Tinggi (px)', 'Rasio Aspek', 'Ukuran File (KB)',
'Kecerahan']
for i, (col, title) in enumerate(zip(metrics, titles)):
    ax = axes[i//3, i%3]
    sns.kdeplot(df_quality[col], ax=ax, fill=True, color=COLORS[i],
alpha=0.7, linewidth=1.8)
    ax.set_title(title, fontweight='bold')

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        ax.grid(True, linestyle='--', alpha=0.4)
axes[1, 2].axis('off')
plt.tight_layout()
plt.show()

fig, ax = plt.subplots(figsize=(8, 4))
for i, parent in enumerate(df_quality['parent'].unique()):
    data = df_quality[df_quality['parent'] == parent]['brightness']
    sns.kdeplot(data, ax=ax, label=parent, color=COLORS[i], fill=True,
alpha=0.6, linewidth=2)
ax.set_xlabel("Kecerahan (Mean RGB)", fontweight='bold')
ax.set_ylabel("Densitas", fontweight='bold')
ax.set_title("Distribusi Kecerahan: Organik vs Anorganik", fontweight='bold')
ax.legend()
ax.grid(True, linestyle='--', alpha=0.4)
plt.tight_layout()
plt.show()

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# 8. TAMPILKAN CONTOH GAMBAR
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print("\n[3/6] Menampilkan contoh gambar per subfolder...")
visualize_sample_images(df_meta, CONFIG["VIS_SAMPLES_PER_CLASS"])

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# 9. PREPROCESSING - GENERATOR DATA & AUGMENTASI
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print("\n" + "="*75)
print("PREPROCESSING DATA UNTUK MODEL DEEP LEARNING")
print("="*75)

# Augmentasi diperkuat untuk akurasi maksimal
train_datagen = ImageDataGenerator(
    rescale=1./255,
    validation_split=CONFIG["VAL_SPLIT"],
    rotation_range=30,
    width_shift_range=0.3,
    height_shift_range=0.3,
    shear_range=0.3,
    zoom_range=0.3,
    horizontal_flip=True,
    vertical_flip=True,
    brightness_range=[0.7, 1.3],
    fill_mode='nearest',
    channel_shift_range=30.0
)

val_datagen = ImageDataGenerator(rescale=1./255,
validation_split=CONFIG["VAL_SPLIT"])

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train_generator = train_datagen.flow_from_directory(
    CONFIG["FLATTENED_DIR"],
    target_size=CONFIG["IMG_SIZE"],
    batch_size=CONFIG["BATCH_SIZE"],
    class_mode='categorical',
    subset='training',
    shuffle=True,
    seed=CONFIG["SEED"]
)

validation_generator = val_datagen.flow_from_directory(
    CONFIG["FLATTENED_DIR"],
    target_size=CONFIG["IMG_SIZE"],
    batch_size=CONFIG["BATCH_SIZE"],
    class_mode='categorical',
    subset='validation',
    shuffle=False,
    seed=CONFIG["SEED"]
)

class_names = list(train_generator.class_indices.keys())
print(f"\nGenerator berhasil dibuat:")
print(f"    Kelas: {class_names}")
print(f"    Training: {train_generator.samples} | Validasi: {validation_generator.samples}")

print("\n[5/6] Contoh hasil augmentasi data...")
X_batch, y_batch = next(train_generator)
fig, axes = plt.subplots(2, 4, figsize=(12, 6))
for i in range(8):
    ax = axes[i//4, i%4]
    ax.imshow(X_batch[i])
    label = class_names[np.argmax(y_batch[i])]
    ax.set_title(label, fontweight='bold')
    ax.axis('off')
plt.suptitle("Contoh Augmentasi Data", fontweight='bold')
plt.tight_layout(rect=[0, 0, 1, 0.96])
plt.show()

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# 10. VERIFIKASI SPLIT - DISTRIBUSI AKHIR
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#
train_counts = get_class_counts(train_generator)
val_counts = get_class_counts(validation_generator)

total_organik = df_meta[df_meta['parent'] == 'Organik']['count'].sum()
total_anorganik = df_meta[df_meta['parent'] == 'anorganik']['count'].sum()

expected_train = {
    'Organik': int(total_organik * (1 - CONFIG["VAL_SPLIT"])),
    'anorganik': int(total_anorganik * (1 - CONFIG["VAL_SPLIT"]))
}

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expected_val = {
    'Organik': total_organik - expected_train['Organik'],
    'anorganik': total_anorganik - expected_train['anorganik']
}

actual_train = dict(zip(class_names, train_counts))
actual_val = dict(zip(class_names, val_counts))

print("\n[6/6] VERIFIKASI SPLIT (80% TRAIN / 20% VAL):")
for cls in class_names:
    print(f"\nKelas: {cls}")
    print(f"    Training → Ekspektasi: {expected_train.get(cls, 0):>4}, Aktual: {actual_train.get(cls, 0):>4}")
    print(f"    Validasi → Ekspektasi: {expected_val.get(cls, 0):>4}, Aktual: {actual_val.get(cls, 0):>4}")

fig, ax = plt.subplots(figsize=(7, 4))
x = np.arange(len(class_names))
width = 0.35
ax.bar(x - width/2, train_counts, width, label='Training', color='#6baed6', edgecolor='black')
ax.bar(x + width/2, val_counts, width, label='Validasi', color='#fdae6b', edgecolor='black')
ax.set_xticks(x)
ax.set_xticklabels(class_names, fontweight='bold')
ax.set_ylabel("Jumlah Sampel", fontweight='bold')
ax.set_title("Distribusi Akhir Setelah Split", fontweight='bold')
ax.legend()
plt.tight_layout()
plt.show()

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# 11. RINGKASAN AWAL & SIMPAN DATA
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#
print("\n" + "="*75)
print("RINGKASAN AWAL: DATA SIAP UNTUK PELATIHAN MODEL")
print("="*75)
print(f"• Dataset Path           : {CONFIG['DATASET_DIR']}")
print(f"• Flattened Path          : {CONFIG['FLATTENED_DIR']}")
print(f"• Total Gambar Valid      : {total_valid}")
print(f"• Kelas                   : Organik, anorganik")
print(f"• Resolusi Input          : {CONFIG['IMG_SIZE']}")
print(f"• Normalisasi             : [0, 1]")
print(f"• Augmentasi              : Aktif")
print(f"• Split                   : {train_generator.samples} train / {validation_generator.samples} val")
print(f"• File Corrupt            : 0")

df_meta.to_csv(f"{CONFIG['OUTPUT_DIR']}/dataset_metadata.csv", index=False)
df_quality.to_csv(f"{CONFIG['OUTPUT_DIR']}/image_quality_analysis.csv", index=False)

print(f"\nData disimpan:")

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print(f" - {CONFIG['OUTPUT_DIR']}/dataset_metadata.csv")
print(f" - {CONFIG['OUTPUT_DIR']}/image_quality_analysis.csv")
print("="*75)

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# 12. FUNGSI MODEL & TRAINING YANG DIPERBAIKI
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import gc
from tensorflow.keras import layers, models, callbacks, optimizers
from tensorflow.keras.applications import EfficientNetB0

tf.config.threading.set_intra_op_parallelism_threads(2)
tf.config.threading.set_inter_op_parallelism_threads(2)
gc.collect()
tf.keras.backend.clear_session()

EPOCHS_SCRATCH = 15
EPOCHS_TL = 12
NUM_CLASSES = 2
CLASS_NAMES = ['Organik', 'anorganik']

# Callbacks diperkaya
common_callbacks = [
    callbacks.EarlyStopping(monitor='val_accuracy', patience=6,
restore_best_weights=True, verbose=1),
    callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=3,
min_lr=1e-7, verbose=0),
    callbacks.TerminateOnNaN(),
    callbacks.ModelCheckpoint(
        filepath=f"{CONFIG['OUTPUT_DIR']}/best_model.h5",
        monitor='val_accuracy',
        save_best_only=True,
        verbose=0
    )
]

def create_optimized_scratch_model():
    """Model dari awal yang dioptimalkan untuk akurasi tinggi"""
    inputs = layers.Input(shape=(CONFIG["IMG_SIZE"], 3))
    x = layers.Conv2D(32, 3, activation='relu')(inputs)
    x = layers.BatchNormalization()(x)
    x = layers.MaxPooling2D()(x)
    x = layers.Conv2D(64, 3, activation='relu')(x)
    x = layers.BatchNormalization()(x)
    x = layers.MaxPooling2D()(x)
    x = layers.Conv2D(128, 3, activation='relu')(x)
    x = layers.BatchNormalization()(x)
    x = layers.GlobalAveragePooling2D()(x)
    x = layers.Dropout(0.4)(x)
    x = layers.Dense(128, activation='relu',
kernel_regularizer=tf.keras.regularizers.l2(1e-4))(x)
    x = layers.BatchNormalization()(x)
    x = layers.Dropout(0.3)(x)

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outputs = layers.Dense(NUM_CLASSES, activation='softmax')(x)
return models.Model(inputs, outputs)

def create_optimized_transfer_learning_model():
    """Model transfer learning yang dioptimalkan"""
    base_model = EfficientNetB0(
        weights='imagenet',
        include_top=False,
        input_shape=(*CONFIG["IMG_SIZE"], 3)
    )
    base_model.trainable = False

    inputs = layers.Input(shape=(*CONFIG["IMG_SIZE"], 3))
    x = base_model(inputs, training=False)
    x = layers.GlobalAveragePooling2D()(x)
    x = layers.BatchNormalization()(x)
    x = layers.Dropout(0.4)(x)
    x = layers.Dense(256, activation='relu',
kernel_regularizer=tf.keras.regularizers.l2(1e-4))(x)
    x = layers.BatchNormalization()(x)
    x = layers.Dropout(0.3)(x)
    outputs = layers.Dense(NUM_CLASSES, activation='softmax')(x)
    return models.Model(inputs, outputs), base_model

def fine_tune_model(model, base_model, train_gen, val_gen):
    """Fine-tuning dengan unfreeze penuh"""
    base_model.trainable = True
    model.compile(
        optimizer=optimizers.Adam(learning_rate=1e-5),
        loss='categorical_crossentropy',
        metrics=['accuracy']
    )
    history_finetune = model.fit(
        train_gen,
        epochs=EPOCHS_TL,
        validation_data=val_gen,
        callbacks=common_callbacks,
        verbose=1
    )
    return history_finetune

def plot_learning_curves(history, title):
    """Plot learning curves"""
    fig, ax = plt.subplots(1, 2, figsize=(14, 5))
    ax[0].plot(history.history['accuracy'], 'o-', color='#2ca02c',
linewidth=2, markersize=6, label='Training')
    ax[0].plot(history.history['val_accuracy'], 's-', color='#d62728',
linewidth=2, markersize=6, label='Validation')
    ax[0].set_title(f'{title} - Akurasi', fontweight='bold')
    ax[0].set_xlabel('Epoch'); ax[0].set_ylabel('Akurasi')
    ax[0].legend(); ax[0].grid(True, linestyle='--', alpha=0.5)
    ax[1].plot(history.history['loss'], 'o-', color='#1f77b4', linewidth=2,
markersize=6, label='Training')
    ax[1].plot(history.history['val_loss'], 's-', color='#ff7f0e',
linewidth=2, markersize=6, label='Validation')
    ax[1].set_title(f'{title} - Loss', fontweight='bold')
    ax[1].set_xlabel('Epoch'); ax[1].set_ylabel('Loss')

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ax[1].legend(); ax[1].grid(True, linestyle='--', alpha=0.5)
plt.tight_layout()
plt.show()

def plot_model_comparison(history_scratch, history_tl):
    """Visualisasi perbandingan model"""
    fig, axes = plt.subplots(2, 2, figsize=(15, 10))
    axes[0,0].plot(history_scratch.history['accuracy'], label='From Scratch (Train)', color='blue', linewidth=2)
    axes[0,0].plot(history_scratch.history['val_accuracy'], label='From Scratch (Val)', color='lightblue', linewidth=2)
    axes[0,0].plot(history_tl.history['accuracy'], label='Transfer Learning (Train)', color='red', linewidth=2)
    axes[0,0].plot(history_tl.history['val_accuracy'], label='Transfer Learning (Val)', color='pink', linewidth=2)
    axes[0,0].set_title('Perbandingan Akurasi Model', fontweight='bold')
    axes[0,0].legend()
    axes[0,0].grid(True, linestyle='--', alpha=0.5)
    axes[0,1].plot(history_scratch.history['loss'], label='From Scratch (Train)', color='blue', linewidth=2)
    axes[0,1].plot(history_scratch.history['val_loss'], label='From Scratch (Val)', color='lightblue', linewidth=2)
    axes[0,1].plot(history_tl.history['loss'], label='Transfer Learning (Train)', color='red', linewidth=2)
    axes[0,1].plot(history_tl.history['val_loss'], label='Transfer Learning (Val)', color='pink', linewidth=2)
    axes[0,1].set_title('Perbandingan Loss Model', fontweight='bold')
    axes[0,1].legend()
    axes[0,1].grid(True, linestyle='--', alpha=0.5)
    plt.tight_layout()
    plt.show()

def error_analysis(model, generator, class_names, model_name):
    """Analisis kesalahan klasifikasi"""
    generator.reset()
    Y_pred = model.predict(generator, verbose=0)
    y_pred = np.argmax(Y_pred, axis=1)
    y_true = generator.classes
    misclassified_idx = np.where(y_pred != y_true)[0]
    print(f"\n=== ERROR ANALYSIS - {model_name} ===")
    print(f"Total misclassified: {len(misclassified_idx)}/{len(y_true)}")
    print(f>Error rate: {len(misclassified_idx)/len(y_true)*100:.2f}%")
    misclassified_per_class = {}
    for class_idx, class_name in enumerate(class_names):
        class_misclassified = [idx for idx in misclassified_idx if y_true[idx] == class_idx]
        misclassified_per_class[class_name] = {
            'count': len(class_misclassified),
            'total': np.sum(y_true == class_idx),
            'error_rate': len(class_misclassified)/np.sum(y_true == class_idx)*100 if np.sum(y_true == class_idx) > 0 else 0
        }
    for class_name, stats in misclassified_per_class.items():
        print(f" {class_name}: {stats['count']}/{stats['total']} ({stats['error_rate']:.2f}%)")
    return misclassified_idx, misclassified_per_class

```

```

def statistical_analysis_report(y_true_scratch, y_pred_scratch, y_true_tl,
y_pred_tl):
    """Analisis statistik mendalam untuk skripsi"""
    print("\n" + "="*60)
    print("ANALISIS STATISTIK UNTUK SKRIPSI")
    print("="*60)
    shapiro_test = stats.shapiro(df_meta['count'])
    print(f"Uji Normalitas Shapiro-Wilk: statistic={shapiro_test[0]:.4f}, p-
value={shapiro_test[1]:.4f}")
    kappa_scratch = cohen_kappa_score(y_true_scratch, y_pred_scratch)
    kappa_tl = cohen_kappa_score(y_true_tl, y_pred_tl)
    print(f"Cohen's Kappa - From Scratch: {kappa_scratch:.4f}")
    print(f"Cohen's Kappa - Transfer Learning: {kappa_tl:.4f}")
    accuracy_scratch = np.mean(y_true_scratch == y_pred_scratch)
    accuracy_tl = np.mean(y_true_tl == y_pred_tl)
    accuracy_diff = accuracy_tl - accuracy_scratch
    print(f"Akurasi - From Scratch: {accuracy_scratch:.4f}")
    print(f"Akurasi - Transfer Learning: {accuracy_tl:.4f}")
    print(f"Perbedaan Akurasi: {accuracy_diff:.4f}")

#
=====
=
# 13. PELATIHAN MODEL - FROM SCRATCH
#
=====
=
print("\n" + "="*75)
print("MELATIH MODEL DARI AWAL (FROM SCRATCH)")
print("="*75)

train_generator.reset()
validation_generator.reset()

model_scratch = create_optimized_scratch_model()
model_scratch.compile(
    optimizer=optimizers.Adam(learning_rate=1e-3),
    loss='categorical_crossentropy',
    metrics=['accuracy']
)

history_scratch = model_scratch.fit(
    train_generator,
    epochs=EPOCHS_SCRATCH,
    validation_data=validation_generator,
    callbacks=common_callbacks,
    verbose=1
)

train_generator.reset()
validation_generator.reset()
plot_learning_curves(history_scratch, "Model From Scratch")

print("\n" + "="*60)
print("EVALUASI MODEL: FROM SCRATCH")
print("="*60)

```

```

validation_generator.reset()
Y_pred_scratch = model_scratch.predict(validation_generator, verbose=1)
y_pred_scratch = np.argmax(Y_pred_scratch, axis=1)
y_true_scratch = validation_generator.classes

print("\n=== Classification Report (From Scratch) ===")
print(classification_report(y_true_scratch, y_pred_scratch,
target_names=CLASS_NAMES))

cm_scratch = confusion_matrix(y_true_scratch, y_pred_scratch)
plt.figure(figsize=(6, 5))
sns.heatmap(cm_scratch, annot=True, fmt='d', cmap='Blues',
xticklabels=CLASS_NAMES, yticklabels=CLASS_NAMES)
plt.title("Confusion Matrix - From Scratch", fontweight='bold')
plt.ylabel("True Label"); plt.xlabel("Predicted Label")
plt.tight_layout()
plt.show()

misclassified_scratch, error_stats_scratch = error_analysis(model_scratch,
validation_generator, CLASS_NAMES, "FROM SCRATCH")

#
=====
=
# 14. PELATIHAN MODEL - TRANSFER LEARNING
#
=====
=
print("\n" + "="*75)
print("MELATIH MODEL DENGAN TRANSFER LEARNING")
print("="*75)

train_generator.reset()
validation_generator.reset()

model_tl, base_model = create_optimized_transfer_learning_model()
model_tl.compile(
    optimizer=optimizers.Adam(learning_rate=1e-3),
    loss='categorical_crossentropy',
    metrics=['accuracy']
)

# Phase 1: Train head
history_phase1 = model_tl.fit(
    train_generator,
    epochs=5,
    validation_data=validation_generator,
    callbacks=[
        callbacks.EarlyStopping(patience=3, restore_best_weights=True),
        callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.5,
patience=2, min_lr=1e-6)
    ],
    verbose=1
)

# Phase 2: Fine-tune

```

```

history_phase2 = fine_tune_model(model_tl, base_model, train_generator,
validation_generator)

# Gabungkan history
combined_history = {}
for key in history_phase1.history.keys():
    combined_history[key] = history_phase1.history[key] +
history_phase2.history[key]

class MockHistory:
    def __init__(self, history_dict):
        self.history = history_dict

history_tl = MockHistory(combined_history)

train_generator.reset()
validation_generator.reset()
plot_learning_curves(history_tl, "Transfer Learning")

print("\n" + "="*60)
print("EVALUASI MODEL: TRANSFER LEARNING")
print("="*60)

validation_generator.reset()
Y_pred_tl = model_tl.predict(validation_generator, verbose=1)
y_pred_tl = np.argmax(Y_pred_tl, axis=1)
y_true_tl = validation_generator.classes

print("\n=== Classification Report (Transfer Learning) ===")
print(classification_report(y_true_tl, y_pred_tl, target_names=CLASS_NAMES))

cm_tl = confusion_matrix(y_true_tl, y_pred_tl)
plt.figure(figsize=(6, 5))
sns.heatmap(cm_tl, annot=True, fmt='d', cmap='Greens',
xticklabels=CLASS_NAMES, yticklabels=CLASS_NAMES)
plt.title("Confusion Matrix - Transfer Learning", fontweight='bold')
plt.ylabel("True Label"); plt.xlabel("Predicted Label")
plt.tight_layout()
plt.show()

misclassified_tl, error_stats_tl = error_analysis(model_tl,
validation_generator, CLASS_NAMES, "TRANSFER LEARNING")

#
=====
#
# 15. ANALISIS PERBANDINGAN DAN STATISTIK
#
=====
#

print("\n" + "="*75)
print("ANALISIS PERBANDINGAN MODEL")
print("="*75)

plot_model_comparison(history_scratch, history_tl)
statistical_analysis_report(y_true_scratch, y_pred_scratch, y_true_tl,
y_pred_tl)

```

```

#
=====
# 16. SIMPAN HASIL DAN MODEL
#
=====

print("\n" + "="*75)
print("MENYIMPAN HASIL DAN MODEL")
print("="*75)

cr_scratch_df = pd.DataFrame(classification_report(y_true_scratch,
y_pred_scratch, target_names=CLASS_NAMES, output_dict=True)).transpose()
cr_tl_df = pd.DataFrame(classification_report(y_true_tl, y_pred_tl,
target_names=CLASS_NAMES, output_dict=True)).transpose()

cr_scratch_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/classification_report_scratch.c
sv", index=False)
cr_tl_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/classification_report_transfer_learn
ing.csv", index=False)

cm_scratch_df = pd.DataFrame(cm_scratch, index=CLASS_NAMES,
columns=CLASS_NAMES)
cm_tl_df = pd.DataFrame(cm_tl, index=CLASS_NAMES, columns=CLASS_NAMES)

cm_scratch_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/confusion_matrix_scratch.csv")
cm_tl_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/confusion_matrix_transfer_learning.c
sv")

model_tl.save(f"{CONFIG['OUTPUT_DIR']}/model_efficientnetb0_waste_classifier.
h5")
print("Model transfer learning disimpan")

print(f"\nSemua hasil disimpan di: {CONFIG['OUTPUT_DIR']}")

#
=====
# 17. KONVERSI MODEL UNTUK DEPLOYMENT
#
=====

print("\n" + "="*75)
print("KONVERSI MODEL UNTUK IMPLEMENTASI WEB")
print("="*75)

try:
    converter = tf.lite.TFLiteConverter.from_keras_model(model_tl)
    converter.optimizations = [tf.lite.Optimize.DEFAULT]
    tflite_model = converter.convert()
    with open(f"{CONFIG['OUTPUT_DIR']}/model_waste_classifier.tflite", "wb")
as f:
    f.write(tflite_model)
    print("Model TFLite disimpan")
    model_tl.save(f"{CONFIG['OUTPUT_DIR']}/model_waste_classifier.keras")
    print("Model Keras format disimpan")

```



```

except Exception as e:
    print(f"Konversi model gagal: {e}")

#
=====
=
# 18. BUAT FILE ZIP DENGAN SEMUA HASIL
#
=====
=
print("\n" + "="*75)
print("MEMBUAT FILE ZIP DENGAN SEMUA HASIL")
print("="*75)

import zipfile
import datetime

timestamp = datetime.datetime.now().strftime("%Y%m%d_%H%M%S")
zip_filename = f"/content/waste_classification_results_{timestamp}.zip"

with zipfile.ZipFile(zip_filename, 'w') as zipf:
    for root, dirs, files in os.walk(CONFIG['OUTPUT_DIR']):
        for file in files:
            file_path = os.path.join(root, file)
            arcname = os.path.relpath(file_path, CONFIG['OUTPUT_DIR'])
            zipf.write(file_path, arcname)

print(f"File zip berhasil dibuat: {zip_filename}")

#
=====
=
# 19. RINGKASAN AKHIR
#
=====
=
print("\n" + "="*75)
print("SELESAI – SEMUA TAHAP TELAH DILAKUKAN")
print("="*75)

final_accuracy_scratch = np.mean(y_true_scratch == y_pred_scratch)
final_accuracy_tl = np.mean(y_true_tl == y_pred_tl)

print(f"HASIL AKHIR:")
print(f"    • From Scratch Accuracy      : {final_accuracy_scratch:.4f}
      ({final_accuracy_scratch*100:.2f}%)" )
print(f"    • Transfer Learning Accuracy: {final_accuracy_tl:.4f}
      ({final_accuracy_tl*100:.2f}%)" )
print(f"    • Peningkatan                  : {final_accuracy_tl -
      final_accuracy_scratch:.4f} ({(final_accuracy_tl -
      final_accuracy_scratch)*100:.2f}%)" )

print(f"\nOUTPUT YANG DIHASILKAN:")
print(f"    • Dataset Metadata              :
      {CONFIG['OUTPUT_DIR']/dataset_metadata.csv}")
print(f"    • Image Quality Analysis        :
      {CONFIG['OUTPUT_DIR']/image_quality_analysis.csv}")

```

```

print(f"    • Classification Reports      :
{CONFIG['OUTPUT_DIR']}/classification_report_*.csv")
print(f"    • Confusion Matrices          :
{CONFIG['OUTPUT_DIR']}/confusion_matrix_*.csv")
print(f"    • Model Files                    : {CONFIG['OUTPUT_DIR']}/model_*.")
print(f"    • Zip File                      : {zip_filename}")

print("\nEksperimen selesai. Semua data dan model telah disimpan.")
print("="*75)

```

---

## EKSPLORESI DAN PREPROCESSING DATASET – KLARIFIKASI SAMPAH

---

```

Dataset Dir: /content/Garbage_classification
  Img Size: (224, 224)
  Batch Size: 32
  Val Split: 0.2
  Seed: 42
  Quality Sample: 50
  Vis Samples Per Class: 4
  Output Dir: /content/output_results

```

---

Dataset berhasil diekstrak

```

Memproses: Organik/Buah → Organik/
Memproses: Organik/Bunga → Organik/
Memproses: Organik/Daging → Organik/
Memproses: Organik/Daun → Organik/
Memproses: Organik/Makanan → Organik/
Memproses: Organik/campuran → Organik/
Memproses: anorganik/cardboard → anorganik/
Memproses: anorganik/glass → anorganik/
Memproses: anorganik/metal → anorganik/
Memproses: anorganik/paper → anorganik/
Memproses: anorganik/plastic → anorganik/
Dataset telah diratakan ke: /content/Garbage_classification_flattened

```

[1/6] Memetakan struktur dataset asli dan memverifikasi integritas file...

```

Organik/
  Buah: 400 valid, 0 corrupt
Organik/
  Bunga: 400 valid, 0 corrupt
Organik/
  Daging: 400 valid, 0 corrupt
Organik/
  Daun: 400 valid, 0 corrupt
Organik/
  Makanan: 400 valid, 0 corrupt
Organik/
  campuran: 400 valid, 0 corrupt
anorganik/
  cardboard: 400 valid, 0 corrupt
anorganik/
  glass: 400 valid, 0 corrupt

```

```
anorganik/
  metal: 400 valid, 0 corrupt
anorganik/
  paper: 400 valid, 0 corrupt
anorganik/
  plastic: 400 valid, 0 corrupt
```

Total gambar valid: 4400

---

---

#### ANALISIS DISTRIBUSI KELAS — STATISTIK DESKRIPTIF

---

---

Jumlah Kelas	:	11
Total Gambar	:	4400
Rata-rata	:	400.00
Median	:	400.00
Standar Deviasi	:	0.00
Koefisien Variasi (%)	:	0.00%
Skewness (Pearson)	:	0
Minimum	:	400
Maksimum	:	400
Rentang	:	0
Rasio Imbalance (Max/Min)	:	1.00x

[2/6] Analisis kualitas gambar...

---

---

#### ANALISIS KUALITAS GAMBAR — STATISTIK

---

---

	width	height	aspect_ratio	file_size_kb	brightness
count	550.00	550.00	550.00	550.00	550.00
mean	394.51	299.62	1.37	20.02	151.37
std	182.72	168.29	0.32	105.10	41.89
min	163.00	123.00	0.49	2.30	11.25
25%	269.25	184.00	1.33	8.66	122.25
50%	342.50	259.00	1.33	11.96	157.03
75%	512.00	384.00	1.49	16.22	179.05
max	2953.00	2953.00	3.33	2154.44	240.71

[3/6] Menampilkan contoh gambar per subfolder...

---

---

#### PREPROCESSING DATA UNTUK MODEL DEEP LEARNING

---

---

Found 3520 images belonging to 2 classes.  
Found 880 images belonging to 2 classes.

Generator berhasil dibuat:

```
Kelas: ['Organik', 'anorganik']
Training: 3520 | Validasi: 880
```

[5/6] Contoh hasil augmentasi data...

[6/6] VERIFIKASI SPLIT (80% TRAIN / 20% VAL):

Kelas: Organik

Training → Ekspektasi: 1920, Aktual: 1920

Validasi → Ekspektasi: 480, Aktual: 480

Kelas: anorganik

Training → Ekspektasi: 1600, Aktual: 1600

Validasi → Ekspektasi: 400, Aktual: 400

---

#### RINGKASAN AWAL: DATA SIAP UNTUK PELATIHAN MODEL

---

- Dataset Path : /content/Garbage\_classification
- Flattened Path : /content/Garbage\_classification\_flattened
- Total Gambar Valid : 4400
- Kelas : Organik, anorganik
- Resolusi Input : (224, 224)
- Normalisasi : [0, 1]
- Augmentasi : Aktif
- Split : 3520 train / 880 val
- File Corrupt : 0

Data disimpan:

- /content/output\_results/dataset\_metadata.csv
  - /content/output\_results/image\_quality\_analysis.csv
- 

---

#### MELATIH MODEL DARI AWAL (FROM SCRATCH)

---

Epoch 1/15

110/110  0s 6s/step - accuracy: 0.8069 - loss: 0.5138

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save\_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my\_model.keras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

110/110  688s 6s/step - accuracy: 0.8072 - loss: 0.5129 - val\_accuracy: 0.4580 - val\_loss: 0.9042 - learning\_rate: 0.0010

Epoch 2/15

110/110  0s 6s/step - accuracy: 0.8834 - loss: 0.3027

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save\_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my\_model.keras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

110/110  705s 6s/step - accuracy: 0.8834 - loss: 0.3029 - val\_accuracy: 0.4932 - val\_loss: 1.1929 - learning\_rate: 0.0010

Epoch 3/15

```
110/110 _____ 0s 6s/step - accuracy: 0.8809 - loss: 0.3071
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 _____ 720s 6s/step - accuracy: 0.8809 - loss: 0.3071 - val_accuracy: 0.4943 - val_loss: 1.6437 - learning_rate: 0.0010
Epoch 4/15
110/110 _____ 0s 6s/step - accuracy: 0.9025 - loss: 0.2765
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 _____ 684s 6s/step - accuracy: 0.9024 - loss: 0.2765 - val_accuracy: 0.6477 - val_loss: 1.1506 - learning_rate: 0.0010
Epoch 5/15
110/110 _____ 0s 6s/step - accuracy: 0.9038 - loss: 0.2532
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 _____ 685s 6s/step - accuracy: 0.9038 - loss: 0.2532 - val_accuracy: 0.8341 - val_loss: 0.3779 - learning_rate: 5.0000e-04
Epoch 6/15
110/110 _____ 735s 6s/step - accuracy: 0.9092 - loss: 0.2346 - val_accuracy: 0.5966 - val_loss: 0.9184 - learning_rate: 5.0000e-04
Epoch 7/15
110/110 _____ 0s 6s/step - accuracy: 0.9096 - loss: 0.2444
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We
recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 _____ 684s 6s/step - accuracy: 0.9096 - loss: 0.2444 - val_accuracy: 0.8909 - val_loss: 0.2675 - learning_rate: 5.0000e-04
Epoch 8/15
110/110 _____ 680s 6s/step - accuracy: 0.9140 - loss: 0.2318 - val_accuracy: 0.8307 - val_loss: 0.3708 - learning_rate: 5.0000e-04
Epoch 9/15
110/110 _____ 681s 6s/step - accuracy: 0.9218 - loss: 0.2086 - val_accuracy: 0.7966 - val_loss: 0.4456 - learning_rate: 5.0000e-04
Epoch 10/15
```

110/110 ————— 674s 6s/step - accuracy: 0.9111 - loss: 0.2248 - val\_accuracy: 0.7795 - val\_loss: 0.5137 - learning\_rate: 5.0000e-04

Epoch 11/15

110/110 ————— 676s 6s/step - accuracy: 0.9159 - loss: 0.2223 - val\_accuracy: 0.7739 - val\_loss: 0.5087 - learning\_rate: 2.5000e-04

Epoch 12/15

110/110 ————— 671s 6s/step - accuracy: 0.9321 - loss: 0.2050 - val\_accuracy: 0.7330 - val\_loss: 0.6150 - learning\_rate: 2.5000e-04

Epoch 13/15

110/110 ————— 675s 6s/step - accuracy: 0.9297 - loss: 0.1938 - val\_accuracy: 0.8807 - val\_loss: 0.2844 - learning\_rate: 2.5000e-04

Epoch 13: early stopping

Restoring model weights from the end of the best epoch: 7.

#### EVALUASI MODEL: FROM SCRATCH

28/28 ————— 34s 1s/step

#### === Classification Report (From Scratch) ===

	precision	recall	f1-score	support
Organik	0.87	0.95	0.90	480
anorganik	0.93	0.82	0.87	400
accuracy			0.89	880
macro avg	0.90	0.89	0.89	880
weighted avg	0.89	0.89	0.89	880

#### === ERROR ANALYSIS - FROM SCRATCH ===

Total misclassified: 96/880

Error rate: 10.91%

Organik: 26/480 (5.42%)

anorganik: 70/400 (17.50%)

#### MELATIH MODEL DENGAN TRANSFER LEARNING

Downloading data from [https://storage.googleapis.com/keras-applications/efficientnetb0\\_notop.h5](https://storage.googleapis.com/keras-applications/efficientnetb0_notop.h5)

16705208/16705208 ————— 1s 0us/step

Epoch 1/5

110/110 ————— 393s 3s/step - accuracy: 0.5227 - loss: 1.1041 - val\_accuracy: 0.5455 - val\_loss: 0.7545 - learning\_rate: 0.0010

Epoch 2/5

110/110 ————— 391s 3s/step - accuracy: 0.5557 - loss: 0.8212 - val\_accuracy: 0.5420 - val\_loss: 0.7316 - learning\_rate: 0.0010

```

Epoch 3/5
110/110 ————— 437s 3s/step - accuracy: 0.5585 - loss: 0.8355 - val_accuracy: 0.5455 - val_loss: 0.7225 - learning_rate: 0.0010
Epoch 4/5
110/110 ————— 374s 3s/step - accuracy: 0.5636 - loss: 0.7991 - val_accuracy: 0.5455 - val_loss: 0.7173 - learning_rate: 0.0010
Epoch 5/5
110/110 ————— 390s 3s/step - accuracy: 0.5614 - loss: 0.7780 - val_accuracy: 0.5432 - val_loss: 0.7007 - learning_rate: 0.0010
Epoch 1/12
110/110 ————— 1241s 11s/step - accuracy: 0.5789 - loss: 0.8080 - val_accuracy: 0.4545 - val_loss: 0.8420 - learning_rate: 1.0000e-05
Epoch 2/12
110/110 ————— 1205s 11s/step - accuracy: 0.8044 - loss: 0.4726 - val_accuracy: 0.5432 - val_loss: 0.7527 - learning_rate: 1.0000e-05
Epoch 3/12
110/110 ————— 1205s 10s/step - accuracy: 0.8803 - loss: 0.3500 - val_accuracy: 0.6557 - val_loss: 0.6157 - learning_rate: 1.0000e-05
Epoch 4/12
110/110 ————— 1177s 11s/step - accuracy: 0.9058 - loss: 0.2907 - val_accuracy: 0.8000 - val_loss: 0.4849 - learning_rate: 1.0000e-05
Epoch 5/12
110/110 ————— 0s 10s/step - accuracy: 0.9324 - loss: 0.2312
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 ————— 1203s 11s/step - accuracy: 0.9324 - loss: 0.2311 - val_accuracy: 0.9148 - val_loss: 0.3199 - learning_rate: 1.0000e-05
Epoch 6/12
110/110 ————— 0s 10s/step - accuracy: 0.9372 - loss: 0.2165
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 ————— 1205s 11s/step - accuracy: 0.9372 - loss: 0.2164 - val_accuracy: 0.9227 - val_loss: 0.2791 - learning_rate: 1.0000e-05
Epoch 7/12
110/110 ————— 0s 10s/step - accuracy: 0.9509 - loss: 0.1879
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my_model.keras')` or `keras.saving.save_model(model, 'my_model.keras')`.

```

```

110/110 ----- 1207s 11s/step - accuracy: 0.9509 -
loss: 0.1879 - val_accuracy: 0.9330 - val_loss: 0.2499 - learning_rate: 1.000
0e-05
Epoch 8/12
110/110 ----- 1220s 11s/step - accuracy: 0.9591 -
loss: 0.1694 - val_accuracy: 0.9250 - val_loss: 0.2452 - learning_rate: 1.000
0e-05
Epoch 9/12
110/110 ----- 0s 10s/step - accuracy: 0.9635 - lo
ss: 0.1562
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We r
ecommend using instead the native Keras format, e.g. `model.save('my_model.ke
ras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 ----- 1194s 11s/step - accuracy: 0.9635 -
loss: 0.1561 - val_accuracy: 0.9364 - val_loss: 0.2249 - learning_rate: 1.000
0e-05
Epoch 10/12
110/110 ----- 1232s 11s/step - accuracy: 0.9624 -
loss: 0.1502 - val_accuracy: 0.9273 - val_loss: 0.2403 - learning_rate: 1.000
0e-05
Epoch 11/12
110/110 ----- 1237s 11s/step - accuracy: 0.9684 -
loss: 0.1375 - val_accuracy: 0.9239 - val_loss: 0.2241 - learning_rate: 1.000
0e-05
Epoch 12/12
110/110 ----- 0s 10s/step - accuracy: 0.9705 - lo
ss: 0.1314
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We r
ecommend using instead the native Keras format, e.g. `model.save('my_model.ke
ras')` or `keras.saving.save_model(model, 'my_model.keras')`.
110/110 ----- 1233s 11s/step - accuracy: 0.9705 -
loss: 0.1314 - val_accuracy: 0.9477 - val_loss: 0.2235 - learning_rate: 1.000
0e-05
Restoring model weights from the end of the best epoch: 12.

```

```

=====
EVALUASI MODEL: TRANSFER LEARNING
=====

```

```

28/28 ----- 66s 2s/step

```

```

=== Classification Report (Transfer Learning) ===

```

	precision	recall	f1-score	support
Organik	0.93	0.98	0.95	480
anorganik	0.97	0.91	0.94	400
accuracy			0.95	880
macro avg	0.95	0.94	0.95	880
weighted avg	0.95	0.95	0.95	880



### === ERROR ANALYSIS - TRANSFER LEARNING ===

Total misclassified: 46/880

Error rate: 5.23%

Organik: 10/480 (2.08%)

anorganik: 36/400 (9.00%)

### ANALISIS PERBANDINGAN MODEL

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save\_model(model)`. This file format is considered legacy. We recommend using instead the native Keras format, e.g. `model.save('my\_model.keras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

### ANALISIS STATISTIK UNTUK SKRIPSI

Uji Normalitas Shapiro-Wilk: statistic=1.0000, p-value=1.0000

Cohen's Kappa - From Scratch: 0.7780

Cohen's Kappa - Transfer Learning: 0.8940

Akurasi - From Scratch: 0.8909

Akurasi - Transfer Learning: 0.9477

Perbedaan Akurasi: 0.0568

### MENYIMPAN HASIL DAN MODEL

Model transfer learning disimpan

Semua hasil disimpan di: /content/output\_results

### KONVERSI MODEL UNTUK IMPLEMENTASI WEB

Saved artifact at '/tmp/tmpjyj7egul'. The following endpoints are available:

\* Endpoint 'serve'

args\_0 (POSITIONAL\_ONLY): TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name='keras\_tensor\_253')

Output Type:

TensorSpec(shape=(None, 2), dtype=tf.float32, name=None)

Captures:

137931268493136: TensorSpec(shape=(1, 1, 1, 3), dtype=tf.float32, name=None)

137931268494672: TensorSpec(shape=(1, 1, 1, 3), dtype=tf.float32, name=None)

137931419741328: TensorSpec(shape=(), dtype=tf.resource, name=None)

137931539593552: TensorSpec(shape=(), dtype=tf.resource, name=None)

137931539591440: TensorSpec(shape=(), dtype=tf.resource, name=None)

137931419739216: TensorSpec(shape=(), dtype=tf.resource, name=None)

137931539591056: TensorSpec(shape=(), dtype=tf.resource, name=None)

137931539591632: TensorSpec(shape=(), dtype=tf.resource, name=None)

[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

137931423329616: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423327120: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423331920: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423332496: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423333072: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423332304: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423333456: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423331344: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423333264: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423331728: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423331152: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423334032: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423333840: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423332880: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423334224: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268490448: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268490640: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423332688: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931423329808: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268489296: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268489488: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268489872: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268491216: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268491792: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268491024: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268492176: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268490256: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268491600: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268491984: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268490832: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268492752: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268491408: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268493520: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268490064: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268492944: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268493328: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268493904: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268494480: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268494864: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268495056: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931539592208: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931539594704: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268495440: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268495248: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268492560: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268496400: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268496208: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268495632: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268496784: TensorSpec(shape=(), dtype=tf.resource, name=None)  
137931268496592: TensorSpec(shape=(), dtype=tf.resource, name=None)

Model TFLite disimpan

Model Keras format disimpan

=====

MEMBUAT FILE ZIP DENGAN SEMUA HASIL

=====

File zip berhasil dibuat: /content/waste\_classification\_results\_20251024\_025736.zip

=====

SELESAI — SEMUA TAHAP TELAH DILAKUKAN

=====

HASIL AKHIR:

- From Scratch Accuracy : 0.8909 (89.09%)
- Transfer Learning Accuracy: 0.9477 (94.77%)
- Peningkatan : 0.0568 (5.68%)

OUTPUT YANG DIHASILKAN:

- Dataset Metadata : /content/output\_results/dataset\_metadata.csv
- Image Quality Analysis : /content/output\_results/image\_quality\_analysis.csv
- Classification Reports : /content/output\_results/classification\_report\_\*.csv
- Confusion Matrices : /content/output\_results/confusion\_matrix\_\*.csv
- Model Files : /content/output\_results/model\_\*
- Zip File : /content/waste\_classification\_results\_20251024\_025736.zip

Eksperimen selesai. Semua data dan model telah disimpan.

=====

*# Pembuatan model kode 2*

*#*

*=*

*# 1. IMPOR PUSTAKA*

*#*

*=*

```
import os
import random
import shutil
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from PIL import Image, ImageStat
from pathlib import Path
import warnings
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import tensorflow as tf
from zipfile import ZipFile
```



```

import json
from scipy import stats
from sklearn.metrics import classification_report, confusion_matrix,
cohen_kappa_score

#
=====
# 1. IMPOR PUSTAKA
#
=====

import os
import random
import shutil
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from PIL import Image, ImageStat
from pathlib import Path
import warnings
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import tensorflow as tf
from zipfile import ZipFile
import json
from scipy import stats
from sklearn.metrics import classification_report, confusion_matrix,
cohen_kappa_score

#
=====
# 2. KONFIGURASI GLOBAL
#
=====

warnings.filterwarnings("ignore")
plt.rcParams.update({'font.size': 10})
sns.set_style("whitegrid")
COLORS = sns.color_palette("Set2", 12)

CONFIG = {
    "DATASET_DIR": "/content/Garbage_classification",
    "FLATTENED_DIR": "/content/Garbage_classification_flattened",
    "IMG_SIZE": (224, 224),
    "BATCH_SIZE": 32,
    "VAL_SPLIT": 0.2,
    "SEED": 42,
    "QUALITY_SAMPLE": 50,
    "VIS_SAMPLES_PER_CLASS": 4,
    "OUTPUT_DIR": "/content/output_results"
}

# Set semua seed untuk reproducibility

```

```

random.seed(CONFIG["SEED"])
np.random.seed(CONFIG["SEED"])
tf.random.set_seed(CONFIG["SEED"])

```

```

# Buat direktori output

```

```

Path(CONFIG["OUTPUT_DIR"]).mkdir(exist_ok=True)

```

```

print("=" * 75)

```

```

print("EKSPLORESI DAN PREPROCESSING DATASET – KLARIFIKASI SAMPAH")

```

```

print("=" * 75)

```

```

for k, v in CONFIG.items():

```

```

    if k != "FLATTENED_DIR":

```

```

        print(f"{k.replace('_', ' ').title():>25}: {v}")

```

```

print("=" * 75)

```

```

=====
EKSPLORESI DAN PREPROCESSING DATASET – KLARIFIKASI SAMPAH
=====

```

```

        Dataset Dir: /content/Garbage classification

```

```

            Img Size: (224, 224)

```

```

            Batch Size: 32

```

```

            Val Split: 0.2

```

```

            Seed: 42

```

```

        Quality Sample: 50

```

```

        Vis Samples Per Class: 4

```

```

        Output Dir: /content/output_results

```

```

#

```

```

=

```

```

# 3. FUNGSI UTILITAS

```

```

#

```

```

=

```

```

def flatten_dataset_structure(src_root: str, dst_root: str):

```

```

    """Memindahkan semua gambar dari subfolder ke dalam 2 folder utama"""

```

```

    src_root = Path(src_root)

```

```

    dst_root = Path(dst_root)

```

```

    dst_root.mkdir(exist_ok=True)

```

```

    parent_classes = ["Organik", "anorganik"]

```

```

    for parent in parent_classes:

```

```

        (dst_root / parent).mkdir(exist_ok=True)

```

```

        parent_path = src_root / parent

```

```

        if not parent_path.exists():

```

```

            raise FileNotFoundError(f"Folder {parent_path} tidak ditemukan!")

```

```

        for subfolder in sorted(parent_path.iterdir()):

```

```

            if subfolder.is_dir():

```

```

                print(f"Memproses: {parent}/{subfolder.name} → {parent}/")

```

```

                for img_file in subfolder.iterdir():

```

```

                    if img_file.suffix.lower() in ['.jpg', '.jpeg', '.png',

```

```

                    '.bmp']:

```

```

                        new_name = f"{subfolder.name}_{img_file.name}"

```

```

                        shutil.copy2(img_file, dst_root / parent / new_name)

```

```

print(f"Dataset telah diratakan ke: {dst_root}")

def analyze_image_quality(df_meta: pd.DataFrame, n_samples: int = 50) ->
pd.DataFrame:
    """Analisis kualitas gambar"""
    quality_data = []
    for _, row in df_meta.iterrows():
        img_files = [f for f in os.listdir(row['path'])
                      if f.lower().endswith(('.png', '.jpg', '.jpeg',
'.bmp'))]
        sampled = random.sample(img_files, min(n_samples, len(img_files)))
        for fname in sampled:
            fpath = os.path.join(row['path'], fname)
            try:
                with Image.open(fpath) as img:
                    w, h = img.size
                    if w == 0 or h == 0: continue
                    fs_kb = os.path.getsize(fpath) / 1024
                    stat = ImageStat.Stat(img.convert('RGB'))
                    brightness = np.mean(stat.mean)
                    ar = w / h
                    quality_data.append({
                        'category': row['category'],
                        'parent': row['parent'],
                        'width': w,
                        'height': h,
                        'aspect_ratio': ar,
                        'file_size_kb': fs_kb,
                        'brightness': brightness
                    })
            except:
                continue
    return pd.DataFrame(quality_data)

def visualize_sample_images(df_meta: pd.DataFrame, n_per_class: int = 4):
    """Visualisasi contoh gambar per kelas"""
    n_classes = len(df_meta)
    fig, axes = plt.subplots(n_classes, n_per_class, figsize=(n_per_class *
2, n_classes * 2))
    if n_classes == 1:
        axes = axes.reshape(1, -1)
    elif n_per_class == 1:
        axes = axes.reshape(-1, 1)

    for i, (_, rec) in enumerate(df_meta.iterrows()):
        img_files = [f for f in os.listdir(rec['path'])
                      if f.lower().endswith(('.png', '.jpg', '.jpeg',
'.bmp'))]
        selected = random.sample(img_files, min(n_per_class, len(img_files)))
        for j, img_name in enumerate(selected):
            ax = axes[i, j] if n_classes > 1 and n_per_class > 1 else
(axes[j] if n_classes == 1 else axes[i])
            try:
                img = Image.open(os.path.join(rec['path'], img_name))
                ax.imshow(img)

```

```

ax.set_title(f"{rec['category']}\n{img.size[0]}x{img.size[1]}", fontsize=9,
fontweight='bold')
    except:
        ax.text(0.5, 0.5, 'Error', ha='center', va='center')
        ax.axis('off')
plt.suptitle("Contoh Gambar per Kelas", fontweight='bold')
plt.tight_layout(rect=[0, 0, 1, 0.97])
plt.show()

```

```

def get_class_counts(gen):
    """Menghitung jumlah sampel per kelas dalam generator"""
    gen.reset()
    steps = (gen.samples + gen.batch_size - 1) // gen.batch_size
    counts = np.zeros(gen.num_classes, dtype=int)
    for _ in range(steps):
        try:
            _, y = next(gen)
            for label in y:
                counts[np.argmax(label)] += 1
        except:
            break
    gen.reset()
    return counts

```

```

#
=====
=
# 4. EKSTRAKSI DAN FLATTEN DATASET
#
=====
=
zip_path = "/content/Garbage_classification.zip"
with ZipFile(zip_path, 'r') as zip_ref:
    zip_ref.extractall("/content")

print("Dataset berhasil diekstrak")

```

```

flatten_dataset_structure(CONFIG["DATASET_DIR"], CONFIG["FLATTENED_DIR"])

```

Dataset berhasil diekstrak

Memproses: Organik/Buah → Organik/

Memproses: Organik/Bunga → Organik/

Memproses: Organik/Daging → Organik/

Memproses: Organik/Daun → Organik/

Memproses: Organik/Makanan → Organik/

Memproses: Organik/campuran → Organik/

Memproses: anorganik/cardboard → anorganik/

Memproses: anorganik/glass → anorganik/

Memproses: anorganik/metal → anorganik/

Memproses: anorganik/paper → anorganik/

Memproses: anorganik/plastic → anorganik/

Dataset telah diratakan ke: /content/Garbage\_classification\_flattened

```

#
=====
# 5. PEMETAAN STRUKTUR DATASET & VERIFIKASI INTEGRITAS
#
=====
#

print("\n[1/6] Memetakan struktur dataset asli dan memverifikasi integritas
file...")

dataset_root = Path(CONFIG["DATASET_DIR"])
parent_classes = ["Organik", "anorganik"]
subfolder_info = []

for parent in parent_classes:
    parent_path = dataset_root / parent
    if not parent_path.exists():
        raise FileNotFoundError(f"Folder {parent_path} tidak ditemukan!")

    for subfolder in sorted(parent_path.iterdir()):
        if subfolder.is_dir():
            valid = 0
            corrupt = 0
            for img_file in subfolder.iterdir():
                if img_file.suffix.lower() in ['.jpg', '.jpeg', '.png',
'.bmp']:
                    try:
                        with Image.open(img_file) as img:
                            img.verify()
                            valid += 1
                    except:
                        corrupt += 1
            subfolder_info.append({
                'parent': parent,
                'category': subfolder.name,
                'path': str(subfolder),
                'valid': valid,
                'corrupt': corrupt
            })
            print(f"    {parent}/")
            print(f"        {subfolder.name}: {valid} valid, {corrupt} corrupt")

df_meta = pd.DataFrame(subfolder_info)
df_meta['count'] = df_meta['valid']
total_valid = df_meta['count'].sum()
print(f"\nTotal gambar valid: {total_valid}")

[1/6] Memetakan struktur dataset asli dan memverifikasi integritas file...
Organik/
    Buah: 400 valid, 0 corrupt
Organik/
    Bunga: 400 valid, 0 corrupt
Organik/
    Daging: 400 valid, 0 corrupt
Organik/
    Daun: 400 valid, 0 corrupt

```

```

Organik/
  Makanan: 400 valid, 0 corrupt
Organik/
  campuran: 400 valid, 0 corrupt
anorganik/
  cardboard: 400 valid, 0 corrupt
anorganik/
  glass: 400 valid, 0 corrupt
anorganik/
  metal: 400 valid, 0 corrupt
anorganik/
  paper: 400 valid, 0 corrupt
anorganik/
  plastic: 400 valid, 0 corrupt

```

Total gambar valid: 4400

```

#
=====
#
# 6. ANALISIS DISTRIBUSI KELAS – STATISTIK & VISUALISASI
#
=====
#
print("\n" + "="*60)
print("ANALISIS DISTRIBUSI KELAS – STATISTIK DESKRIPTIF")
print("="*60)

counts = df_meta['count'].values
categories = df_meta['category'].values

mean_val = np.mean(counts)
median_val = np.median(counts)
std_val = np.std(counts)
cv_pct = (std_val / mean_val) * 100 if mean_val > 0 else 0
skewness = (3 * (mean_val - median_val)) / std_val if std_val != 0 else 0

descriptive_stats = {
    "Jumlah Kelas": len(counts),
    "Total Gambar": np.sum(counts),
    "Rata-rata": mean_val,
    "Median": median_val,
    "Standar Deviasi": std_val,
    "Koefisien Variasi (%)": cv_pct,
    "Skewness (Pearson)": skewness,
    "Minimum": np.min(counts),
    "Maksimum": np.max(counts),
    "Rentang": np.max(counts) - np.min(counts),
    "Rasio Imbalance (Max/Min)": np.max(counts) / np.min(counts) if
np.min(counts) > 0 else 0
}

for k, v in descriptive_stats.items():
    if isinstance(v, float):
        if "Koefisien" in k:

```

```

        print(f"{k:<30}: {v:>8.2f}%")
    elif "Rasio" in k:
        print(f"{k:<30}: {v:>8.2f}x")
    else:
        print(f"{k:<30}: {v:>8.2f}")
else:
    print(f"{k:<30}: {v:>8}")

fig, ax = plt.subplots(figsize=(8, 8))
wedges, texts, autotexts = ax.pie(
    counts,
    labels=categories,
    autopct=lambda pct: f"{int(pct/100.*sum(counts))}",
    startangle=90,
    colors=COLORS[:len(categories)],
    textprops={'fontweight': 'bold'},
    wedgeprops={'edgecolor': 'white', 'linewidth': 1.5, 'alpha': 0.9}
)
ax.set_title("Distribusi Jumlah Gambar per Kelas", fontweight='bold',
fontsize=13)
plt.tight_layout()
plt.show()

parent_summary =
df_meta.groupby('parent')['count'].sum().sort_values(ascending=False)
fig, ax = plt.subplots(figsize=(6, 4))
bars = ax.bar(parent_summary.index, parent_summary.values,
              color=['#6baed6', '#fdae6b'][:len(parent_summary)],
              edgecolor='black')
ax.set_ylabel("Jumlah Gambar", fontweight='bold')
ax.set_title("Distribusi: Organik vs Anorganik", fontweight='bold', pad=20)
for bar, count in zip(bars, parent_summary.values):
    pct = count / parent_summary.sum() * 100
    ax.text(bar.get_x() + bar.get_width()/2, bar.get_height() + 20,
            f"{int(count)}\n({pct:.1f}%)", ha='center', fontweight='bold',
            fontsize=9.5)
plt.tight_layout()
plt.show()

```

---



---

#### ANALISIS DISTRIBUSI KELAS – STATISTIK DESKRIPTIF

---



---

Jumlah Kelas	:	11
Total Gambar	:	4400
Rata-rata	:	400.00
Median	:	400.00
Standar Deviasi	:	0.00
Koefisien Variasi (%)	:	0.00%
Skewness (Pearson)	:	0
Minimum	:	400
Maksimum	:	400
Rentang	:	0
Rasio Imbalance (Max/Min)	:	1.00x

```

#
=====
# 7. ANALISIS KUALITAS GAMBAR
#
=====
#

print("\n[2/6] Analisis kualitas gambar...")
df_quality = analyze_image_quality(df_meta, CONFIG["QUALITY_SAMPLE"])

if not df_quality.empty:
    print("\n" + "="*60)
    print("ANALISIS KUALITAS GAMBAR - STATISTIK")
    print("="*60)
    desc = df_quality[['width', 'height', 'aspect_ratio', 'file_size_kb',
'brightness']].describe()
    print(desc.round(2))

fig, axes = plt.subplots(2, 3, figsize=(15, 8))
metrics = ['width', 'height', 'aspect_ratio', 'file_size_kb', 'brightness']
titles = ['Lebar (px)', 'Tinggi (px)', 'Rasio Aspek', 'Ukuran File (KB)',
'Kecerahan']
for i, (col, title) in enumerate(zip(metrics, titles)):
    ax = axes[i//3, i%3]
    sns.kdeplot(df_quality[col], ax=ax, fill=True, color=COLORS[i],
alpha=0.7, linewidth=1.8)
    ax.set_title(title, fontweight='bold')
    ax.grid(True, linestyle='--', alpha=0.4)
axes[1, 2].axis('off')
plt.tight_layout()
plt.show()

fig, ax = plt.subplots(figsize=(8, 4))
for i, parent in enumerate(df_quality['parent'].unique()):
    data = df_quality[df_quality['parent'] == parent]['brightness']
    sns.kdeplot(data, ax=ax, label=parent, color=COLORS[i], fill=True,
alpha=0.6, linewidth=2)
ax.set_xlabel("Kecerahan (Mean RGB)", fontweight='bold')
ax.set_ylabel("Densitas", fontweight='bold')
ax.set_title("Distribusi Kecerahan: Organik vs Anorganik", fontweight='bold')
ax.legend()
ax.grid(True, linestyle='--', alpha=0.4)
plt.tight_layout()
plt.show()

```

[2/6] Analisis kualitas gambar...

---

#### ANALISIS KUALITAS GAMBAR - STATISTIK

---

	width	height	aspect_ratio	file_size_kb	brightness
count	550.00	550.00	550.00	550.00	550.00
mean	382.40	288.21	1.36	15.22	150.62
std	146.04	113.66	0.30	51.64	40.40
min	87.00	125.00	0.39	2.41	29.92
25%	260.00	185.00	1.33	8.23	121.27



50%	318.00	244.50	1.33	12.00	156.22
75%	512.00	384.00	1.47	15.91	178.14
max	1920.00	1429.00	3.23	1212.94	239.10

```
#
=====
=
# 8. TAMPILKAN CONTOH GAMBAR
#
=====
=
print("\n[3/6] Menampilkan contoh gambar per subfolder...")
visualize_sample_images(df_meta, CONFIG["VIS_SAMPLES_PER_CLASS"])

[3/6] Menampilkan contoh gambar per subfolder...

#
=====
=
# 9. PREPROCESSING - GENERATOR DATA & AUGMENTASI
#
=====
=
print("\n" + "="*75)
print("PREPROCESSING DATA UNTUK MODEL DEEP LEARNING")
print("="*75)

# Augmentasi diperkuat untuk akurasi maksimal
train_datagen = ImageDataGenerator(
    rescale=1./255,
    validation_split=CONFIG["VAL_SPLIT"],
    rotation_range=30,
    width_shift_range=0.3,
    height_shift_range=0.3,
    shear_range=0.3,
    zoom_range=0.3,
    horizontal_flip=True,
    vertical_flip=True,
    brightness_range=[0.7, 1.3],
    fill_mode='nearest',
    channel_shift_range=30.0
)

val_datagen = ImageDataGenerator(rescale=1./255,
validation_split=CONFIG["VAL_SPLIT"])

train_generator = train_datagen.flow_from_directory(
    CONFIG["FLATTENED_DIR"],
    target_size=CONFIG["IMG_SIZE"],
    batch_size=CONFIG["BATCH_SIZE"],
    class_mode='categorical',
    subset='training',
    shuffle=True,
```

```

        seed=CONFIG["SEED"]
    )

validation_generator = val_datagen.flow_from_directory(
    CONFIG["FLATTENED_DIR"],
    target_size=CONFIG["IMG_SIZE"],
    batch_size=CONFIG["BATCH_SIZE"],
    class_mode='categorical',
    subset='validation',
    shuffle=False,
    seed=CONFIG["SEED"]
)

class_names = list(train_generator.class_indices.keys())
print(f"\nGenerator berhasil dibuat:")
print(f"    Kelas: {class_names}")
print(f"    Training: {train_generator.samples} | Validasi: {validation_generator.samples}")

print("\n[5/6] Contoh hasil augmentasi data...")
X_batch, y_batch = next(train_generator)
fig, axes = plt.subplots(2, 4, figsize=(12, 6))
for i in range(8):
    ax = axes[i//4, i%4]
    ax.imshow(X_batch[i])
    label = class_names[np.argmax(y_batch[i])]
    ax.set_title(label, fontweight='bold')
    ax.axis('off')
plt.suptitle("Contoh Augmentasi Data", fontweight='bold')
plt.tight_layout(rect=[0, 0, 1, 0.96])
plt.show()

```

---

## PREPROCESSING DATA UNTUK MODEL DEEP LEARNING

---

Found 3520 images belonging to 2 classes.  
Found 880 images belonging to 2 classes.

Generator berhasil dibuat:  
 Kelas: ['Organik', 'anorganik']  
 Training: 3520 | Validasi: 880

[5/6] Contoh hasil augmentasi data...

```

#
=====
#
# 10. VERIFIKASI SPLIT - DISTRIBUSI AKHIR
#
=====
#
train_counts = get_class_counts(train_generator)
val_counts = get_class_counts(validation_generator)

total_organik = df_meta[df_meta['parent'] == 'Organik']['count'].sum()

```

```

total_anorganik = df_meta[df_meta['parent'] == 'anorganik']['count'].sum()

expected_train = {
    'Organik': int(total_organik * (1 - CONFIG["VAL_SPLIT"])),
    'anorganik': int(total_anorganik * (1 - CONFIG["VAL_SPLIT"]))
}
expected_val = {
    'Organik': total_organik - expected_train['Organik'],
    'anorganik': total_anorganik - expected_train['anorganik']
}

actual_train = dict(zip(class_names, train_counts))
actual_val = dict(zip(class_names, val_counts))

print("\n[6/6] VERIFIKASI SPLIT (80% TRAIN / 20% VAL):")
for cls in class_names:
    print(f"\nKelas: {cls}")
    print(f"    Training → Ekspektasi: {expected_train.get(cls, 0):>4}, Aktual: {actual_train.get(cls, 0):>4}")
    print(f"    Validasi → Ekspektasi: {expected_val.get(cls, 0):>4}, Aktual: {actual_val.get(cls, 0):>4}")

fig, ax = plt.subplots(figsize=(7, 4))
x = np.arange(len(class_names))
width = 0.35
ax.bar(x - width/2, train_counts, width, label='Training', color='#6baed6',
       edgcolor='black')
ax.bar(x + width/2, val_counts, width, label='Validasi', color='#fdae6b',
       edgcolor='black')
ax.set_xticks(x)
ax.set_xticklabels(class_names, fontweight='bold')
ax.set_ylabel("Jumlah Sampel", fontweight='bold')
ax.set_title("Distribusi Akhir Setelah Split", fontweight='bold')
ax.legend()
plt.tight_layout()
plt.show()

```

[6/6] VERIFIKASI SPLIT (80% TRAIN / 20% VAL):

Kelas: Organik

Training → Ekspektasi: 1920, Aktual: 1920  
Validasi → Ekspektasi: 480, Aktual: 480

Kelas: anorganik

Training → Ekspektasi: 1600, Aktual: 1600  
Validasi → Ekspektasi: 400, Aktual: 400

#

=

# 11. RINGKASAN AWAL & SIMPAN DATA

#

=

```
print("\n" + "="*75)
```

```

print("RINGKASAN AWAL: DATA SIAP UNTUK PELATIHAN MODEL")
print("="*75)
print(f"• Dataset Path           : {CONFIG['DATASET_DIR']}")
print(f"• Flattened Path           : {CONFIG['FLATTENED_DIR']}")
print(f"• Total Gambar Valid       : {total_valid}")
print(f"• Kelas                     : Organik, anorganik")
print(f"• Resolusi Input           : {CONFIG['IMG_SIZE']}")
print(f"• Normalisasi               : [0, 1]")
print(f"• Augmentasi                : Aktif")
print(f"• Split                     : {train_generator.samples} train /
{validation_generator.samples} val")
print(f"• File Corrupt              : 0")

df_meta.to_csv(f"{CONFIG['OUTPUT_DIR']}/dataset_metadata.csv", index=False)
df_quality.to_csv(f"{CONFIG['OUTPUT_DIR']}/image_quality_analysis.csv",
index=False)

print(f"\nData disimpan:")
print(f"  - {CONFIG['OUTPUT_DIR']}/dataset_metadata.csv")
print(f"  - {CONFIG['OUTPUT_DIR']}/image_quality_analysis.csv")
print("="*75)

```

```

#
=====
# 12. FUNGSI MODEL & TRAINING YANG DIPERBAIKI
#
=====
import gc
from tensorflow.keras import layers, models, callbacks, optimizers
from tensorflow.keras.applications import EfficientNetB0

tf.config.threading.set_intra_op_parallelism_threads(2)
tf.config.threading.set_inter_op_parallelism_threads(2)
gc.collect()
tf.keras.backend.clear_session()

EPOCHS_SCRATCH = 15
EPOCHS_TL = 12
NUM_CLASSES = 2
CLASS_NAMES = ['Organik', 'anorganik']

# Callbacks diperkaya
common_callbacks = [
    callbacks.EarlyStopping(monitor='val_accuracy', patience=6,
restore_best_weights=True, verbose=1),
    callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=3,
min_lr=1e-7, verbose=0),
    callbacks.TerminateOnNaN(),
    callbacks.ModelCheckpoint(
        filepath=f"{CONFIG['OUTPUT_DIR']}/best_model.h5",
        monitor='val_accuracy',
        save_best_only=True,
        verbose=0
    )
]

```

```

def create_optimized_scratch_model():
    """Model dari awal yang dioptimalkan untuk akurasi tinggi"""
    inputs = layers.Input(shape=(*CONFIG["IMG_SIZE"], 3))
    x = layers.Conv2D(32, 3, activation='relu')(inputs)
    x = layers.BatchNormalization()(x)
    x = layers.MaxPooling2D()(x)
    x = layers.Conv2D(64, 3, activation='relu')(x)
    x = layers.BatchNormalization()(x)
    x = layers.MaxPooling2D()(x)
    x = layers.Conv2D(128, 3, activation='relu')(x)
    x = layers.BatchNormalization()(x)
    x = layers.GlobalAveragePooling2D()(x)
    x = layers.Dropout(0.4)(x)
    x = layers.Dense(128, activation='relu',
kernel_regularizer=tf.keras.regularizers.l2(1e-4))(x)
    x = layers.BatchNormalization()(x)
    x = layers.Dropout(0.3)(x)
    outputs = layers.Dense(NUM_CLASSES, activation='softmax')(x)
    return models.Model(inputs, outputs)

def create_optimized_transfer_learning_model():
    """Model transfer learning yang dioptimalkan"""
    base_model = EfficientNetB0(
        weights='imagenet',
        include_top=False,
        input_shape=(*CONFIG["IMG_SIZE"], 3)
    )
    base_model.trainable = False

    inputs = layers.Input(shape=(*CONFIG["IMG_SIZE"], 3))
    x = base_model(inputs, training=False)
    x = layers.GlobalAveragePooling2D()(x)
    x = layers.BatchNormalization()(x)
    x = layers.Dropout(0.4)(x)
    x = layers.Dense(256, activation='relu',
kernel_regularizer=tf.keras.regularizers.l2(1e-4))(x)
    x = layers.BatchNormalization()(x)
    x = layers.Dropout(0.3)(x)
    outputs = layers.Dense(NUM_CLASSES, activation='softmax')(x)
    return models.Model(inputs, outputs), base_model

def fine_tune_model(model, base_model, train_gen, val_gen):
    """Fine-tuning dengan unfreeze penuh"""
    base_model.trainable = True
    model.compile(
        optimizer=optimizers.Adam(learning_rate=1e-5),
        loss='categorical_crossentropy',
        metrics=['accuracy']
    )
    history_finetune = model.fit(
        train_gen,
        epochs=EPOCHS_TL,
        validation_data=val_gen,
        callbacks=common_callbacks,
        verbose=1
    )

```

```

    return history_finetune

def plot_learning_curves(history, title):
    """Plot learning curves"""
    fig, ax = plt.subplots(1, 2, figsize=(14, 5))
    ax[0].plot(history.history['accuracy'], 'o-', color='#2ca02c',
linewidth=2, markersize=6, label='Training')
    ax[0].plot(history.history['val_accuracy'], 's-', color='#d62728',
linewidth=2, markersize=6, label='Validation')
    ax[0].set_title(f'{title} - Akurasi', fontweight='bold')
    ax[0].set_xlabel('Epoch'); ax[0].set_ylabel('Akurasi')
    ax[0].legend(); ax[0].grid(True, linestyle='--', alpha=0.5)
    ax[1].plot(history.history['loss'], 'o-', color='#1f77b4', linewidth=2,
markersize=6, label='Training')
    ax[1].plot(history.history['val_loss'], 's-', color='#ff7f0e',
linewidth=2, markersize=6, label='Validation')
    ax[1].set_title(f'{title} - Loss', fontweight='bold')
    ax[1].set_xlabel('Epoch'); ax[1].set_ylabel('Loss')
    ax[1].legend(); ax[1].grid(True, linestyle='--', alpha=0.5)
    plt.tight_layout()
    plt.show()

def plot_model_comparison(history_scratch, history_tl):
    """Visualisasi perbandingan model"""
    fig, axes = plt.subplots(2, 2, figsize=(15, 10))
    axes[0,0].plot(history_scratch.history['accuracy'], label='From Scratch
(Train)', color='blue', linewidth=2)
    axes[0,0].plot(history_scratch.history['val_accuracy'], label='From
Scratch (Val)', color='lightblue', linewidth=2)
    axes[0,0].plot(history_tl.history['accuracy'], label='Transfer Learning
(Train)', color='red', linewidth=2)
    axes[0,0].plot(history_tl.history['val_accuracy'], label='Transfer
Learning (Val)', color='pink', linewidth=2)
    axes[0,0].set_title('Perbandingan Akurasi Model', fontweight='bold')
    axes[0,0].legend()
    axes[0,0].grid(True, linestyle='--', alpha=0.5)
    axes[0,1].plot(history_scratch.history['loss'], label='From Scratch
(Train)', color='blue', linewidth=2)
    axes[0,1].plot(history_scratch.history['val_loss'], label='From Scratch
(Val)', color='lightblue', linewidth=2)
    axes[0,1].plot(history_tl.history['loss'], label='Transfer Learning
(Train)', color='red', linewidth=2)
    axes[0,1].plot(history_tl.history['val_loss'], label='Transfer Learning
(Val)', color='pink', linewidth=2)
    axes[0,1].set_title('Perbandingan Loss Model', fontweight='bold')
    axes[0,1].legend()
    axes[0,1].grid(True, linestyle='--', alpha=0.5)
    plt.tight_layout()
    plt.show()

def error_analysis(model, generator, class_names, model_name):
    """Analisis kesalahan klasifikasi"""
    generator.reset()
    Y_pred = model.predict(generator, verbose=0)
    y_pred = np.argmax(Y_pred, axis=1)
    y_true = generator.classes
    misclassified_idx = np.where(y_pred != y_true)[0]

```

```

print(f"\n=== ERROR ANALYSIS - {model_name} ===")
print(f"Total misclassified: {len(misclassified_idx)}/{len(y_true)}")
print(f"Error rate: {len(misclassified_idx)/len(y_true)*100:.2f}%")
misclassified_per_class = {}
for class_idx, class_name in enumerate(class_names):
    class_misclassified = [idx for idx in misclassified_idx if
y_true[idx] == class_idx]
    misclassified_per_class[class_name] = {
        'count': len(class_misclassified),
        'total': np.sum(y_true == class_idx),
        'error_rate': len(class_misclassified)/np.sum(y_true ==
class_idx)*100 if np.sum(y_true == class_idx) > 0 else 0
    }
    for class_name, stats in misclassified_per_class.items():
        print(f" {class_name}: {stats['count']}/{stats['total']}
({stats['error_rate']:.2f}%)")
    return misclassified_idx, misclassified_per_class

def statistical_analysis_report(y_true_scratch, y_pred_scratch, y_true_tl,
y_pred_tl):
    """Analisis statistik mendalam untuk skripsi"""
    print("\n" + "="*60)
    print("ANALISIS STATISTIK UNTUK SKRIPSI")
    print("="*60)
    shapiro_test = stats.shapiro(df_meta['count'])
    print(f"Uji Normalitas Shapiro-Wilk: statistic={shapiro_test[0]:.4f}, p-
value={shapiro_test[1]:.4f}")
    kappa_scratch = cohen_kappa_score(y_true_scratch, y_pred_scratch)
    kappa_tl = cohen_kappa_score(y_true_tl, y_pred_tl)
    print(f"Cohen's Kappa - From Scratch: {kappa_scratch:.4f}")
    print(f"Cohen's Kappa - Transfer Learning: {kappa_tl:.4f}")
    accuracy_scratch = np.mean(y_true_scratch == y_pred_scratch)
    accuracy_tl = np.mean(y_true_tl == y_pred_tl)
    accuracy_diff = accuracy_tl - accuracy_scratch
    print(f"Akurasi - From Scratch: {accuracy_scratch:.4f}")
    print(f"Akurasi - Transfer Learning: {accuracy_tl:.4f}")
    print(f"Perbedaan Akurasi: {accuracy_diff:.4f}")

```

---

#### RINGKASAN AWAL: DATA SIAP UNTUK PELATIHAN MODEL

---

• Dataset Path	: /content/Garbage classification
• Flattened Path	: /content/Garbage_classification_flattened
• Total Gambar Valid	: 4400
• Kelas	: Organik, anorganik
• Resolusi Input	: (224, 224)
• Normalisasi	: [0, 1]
• Augmentasi	: Aktif
• Split	: 3520 train / 880 val
• File Corrupt	: 0

Data disimpan:

- /content/output\_results/dataset\_metadata.csv
  - /content/output\_results/image\_quality\_analysis.csv
-

```

#
=====
# 13. PELATIHAN MODEL - FROM SCRATCH
#
=====

print("\n" + "="*75)
print("MELATIH MODEL DARI AWAL (FROM SCRATCH)")
print("="*75)

train_generator.reset()
validation_generator.reset()

model_scratch = create_optimized_scratch_model()
model_scratch.compile(
    optimizer=optimizers.Adam(learning_rate=1e-3),
    loss='categorical_crossentropy',
    metrics=['accuracy']
)

history_scratch = model_scratch.fit(
    train_generator,
    epochs=EPOCHS_SCRATCH,
    validation_data=validation_generator,
    callbacks=common_callbacks,
    verbose=1
)

train_generator.reset()
validation_generator.reset()
plot_learning_curves(history_scratch, "Model From Scratch")

print("\n" + "="*60)
print("EVALUASI MODEL: FROM SCRATCH")
print("="*60)

validation_generator.reset()
Y_pred_scratch = model_scratch.predict(validation_generator, verbose=1)
y_pred_scratch = np.argmax(Y_pred_scratch, axis=1)
y_true_scratch = validation_generator.classes

print("\n=== Classification Report (From Scratch) ===")
print(classification_report(y_true_scratch, y_pred_scratch,
    target_names=CLASS_NAMES))

cm_scratch = confusion_matrix(y_true_scratch, y_pred_scratch)
plt.figure(figsize=(6, 5))
sns.heatmap(cm_scratch, annot=True, fmt='d', cmap='Blues',
    xticklabels=CLASS_NAMES, yticklabels=CLASS_NAMES)
plt.title("Confusion Matrix - From Scratch", fontweight='bold')
plt.ylabel("True Label"); plt.xlabel("Predicted Label")
plt.tight_layout()
plt.show()

```



```
misclassified_scratch, error_stats_scratch = error_analysis(model_scratch,
validation_generator, CLASS_NAMES, "FROM SCRATCH")
```

---


### MELATIH MODEL DARI AWAL (FROM SCRATCH)

---

Epoch 1/15

110/110  0s 463ms/step - accuracy: 0.7967 -  
loss: 0.5207


WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or  
`keras.saving.save\_model(model)`. This file format is considered legacy. We r  
ecommend using instead the native Keras format, e.g. `model.save('my\_model.ke  
ras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

110/110  65s 497ms/step - accuracy: 0.7971 -  
loss: 0.5198 - val\_accuracy: 0.4557 - val\_loss: 0.8445 - learning\_rate: 0.001  
0

Epoch 2/15

110/110  0s 491ms/step - accuracy: 0.8857 -  
loss: 0.3134


WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or  
`keras.saving.save\_model(model)`. This file format is considered legacy. We r  
ecommend using instead the native Keras format, e.g. `model.save('my\_model.ke  
ras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

110/110  56s 506ms/step - accuracy: 0.8857 -  
loss: 0.3134 - val\_accuracy: 0.4682 - val\_loss: 1.3567 - learning\_rate: 0.001  
0

Epoch 3/15

110/110  0s 461ms/step - accuracy: 0.8871 -  
loss: 0.3122


WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or  
`keras.saving.save\_model(model)`. This file format is considered legacy. We r  
ecommend using instead the native Keras format, e.g. `model.save('my\_model.ke  
ras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

110/110  52s 473ms/step - accuracy: 0.8871 -  
loss: 0.3122 - val\_accuracy: 0.5318 - val\_loss: 1.0800 - learning\_rate: 0.001  
0


Epoch 4/15

110/110  0s 466ms/step - accuracy: 0.9054 -  
loss: 0.2745

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or  
`keras.saving.save\_model(model)`. This file format is considered legacy. We r  
ecommend using instead the native Keras format, e.g. `model.save('my\_model.ke  
ras')` or `keras.saving.save\_model(model, 'my\_model.keras')`.

110/110  53s 479ms/step - accuracy: 0.9054 -  
loss: 0.2746 - val\_accuracy: 0.7239 - val\_loss: 0.6734 - learning\_rate: 0.001  
0

Epoch 5/15

110/110  54s 490ms/step - accuracy: 0.9076 -  
loss: 0.2594 - val\_accuracy: 0.6068 - val\_loss: 0.7278 - learning\_rate: 0.001  
0

Epoch 6/15

```
110/110 _____ 0s 462ms/step - accuracy: 0.8992 -  
loss: 0.2590  
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or  
`keras.saving.save_model(model)`. This file format is considered legacy. We r  
ecommend using instead the native Keras format, e.g. `model.save('my_model.ke  
ras')` or `keras.saving.save_model(model, 'my_model.keras')`.  
110/110 _____ 52s 476ms/step - accuracy: 0.8992 -  
loss: 0.2589 - val_accuracy: 0.7943 - val_loss: 0.4586 - learning_rate: 0.001  
0  
Epoch 7/15  
110/110 _____ 52s 472ms/step - accuracy: 0.9167 -  
loss: 0.2373 - val_accuracy: 0.6114 - val_loss: 0.8699 - learning_rate: 0.001  
0  
Epoch 8/15  
110/110 _____ 53s 484ms/step - accuracy: 0.9065 -  
loss: 0.2528 - val_accuracy: 0.5534 - val_loss: 1.2071 - learning_rate: 0.001  
0  
Epoch 9/15  
110/110 _____ 0s 460ms/step - accuracy: 0.9151 -  
loss: 0.2214  
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or  
`keras.saving.save_model(model)`. This file format is considered legacy. We r  
ecommend using instead the native Keras format, e.g. `model.save('my_model.ke  
ras')` or `keras.saving.save_model(model, 'my_model.keras')`.  
110/110 _____ 52s 473ms/step - accuracy: 0.9151 -  
loss: 0.2215 - val_accuracy: 0.8670 - val_loss: 0.3643 - learning_rate: 0.001  
0  
Epoch 10/15  
110/110 _____ 52s 472ms/step - accuracy: 0.9065 -  
loss: 0.2368 - val_accuracy: 0.6125 - val_loss: 0.8514 - learning_rate: 0.001  
0  
Epoch 11/15  
110/110 _____ 54s 487ms/step - accuracy: 0.9134 -  
loss: 0.2278 - val_accuracy: 0.5455 - val_loss: 2.1291 - learning_rate: 0.001  
0  
Epoch 12/15  
110/110 _____ 52s 472ms/step - accuracy: 0.9071 -  
loss: 0.2427 - val_accuracy: 0.5477 - val_loss: 2.0493 - learning_rate: 0.001  
0  
Epoch 13/15  
110/110 _____ 0s 460ms/step - accuracy: 0.9194 -  
loss: 0.2111  
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or  
`keras.saving.save_model(model)`. This file format is considered legacy. We r  
ecommend using instead the native Keras format, e.g. `model.save('my_model.ke  
ras')` or `keras.saving.save_model(model, 'my_model.keras')`.  
110/110 _____ 52s 472ms/step - accuracy: 0.9195 -  
loss: 0.2111 - val_accuracy: 0.9364 - val_loss: 0.1953 - learning_rate: 5.000  
0e-04  
Epoch 14/15
```

110/110 ————— 54s 493ms/step - accuracy: 0.9313 -  
loss: 0.1987 - val\_accuracy: 0.8420 - val\_loss: 0.4219 - learning\_rate: 5.000  
0e-04

Epoch 15/15

110/110 ————— 53s 480ms/step - accuracy: 0.9283 -  
loss: 0.1881 - val\_accuracy: 0.7580 - val\_loss: 0.5596 - learning\_rate: 5.000  
0e-04

Restoring model weights from the end of the best epoch: 13.

=====

EVALUASI MODEL: FROM SCRATCH

=====

28/28 ————— 2s 63ms/step

=== Classification Report (From Scratch) ===

	precision	recall	f1-score	support
Organik	0.93	0.96	0.94	480
anorganik	0.95	0.91	0.93	400
accuracy			0.94	880
macro avg	0.94	0.93	0.94	880
weighted avg	0.94	0.94	0.94	880

=== ERROR ANALYSIS - FROM SCRATCH ===

Total misclassified: 56/880

Error rate: 6.36%

Organik: 21/480 (4.38%)

anorganik: 35/400 (8.75%)

#

=

# 14. PELATIHAN MODEL - TRANSFER LEARNING

#

=

```
print("\n" + "="*75)
print("MELATIH MODEL DENGAN TRANSFER LEARNING")
print("="*75)
```

```
train_generator.reset()
validation_generator.reset()
```

```
model_tl, base_model = create_optimized_transfer_learning_model()
model_tl.compile(
    optimizer=optimizers.Adam(learning_rate=1e-3),
    loss='categorical_crossentropy',
    metrics=['accuracy']
)
```

# Phase 1: Train head

```
history_phase1 = model_tl.fit(
```

```

train_generator,
epochs=5,
validation_data=validation_generator,
callbacks=[
    callbacks.EarlyStopping(patience=3, restore_best_weights=True),
    callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.5,
patience=2, min_lr=1e-6)
],
verbose=1
)

# Phase 2: Fine-tune
history_phase2 = fine_tune_model(model_tl, base_model, train_generator,
validation_generator)

# Gabungkan history
combined_history = {}
for key in history_phase1.history.keys():
    combined_history[key] = history_phase1.history[key] +
history_phase2.history[key]

class MockHistory:
    def __init__(self, history_dict):
        self.history = history_dict

history_tl = MockHistory(combined_history)

train_generator.reset()
validation_generator.reset()
plot_learning_curves(history_tl, "Transfer Learning")

print("\n" + "="*60)
print("EVALUASI MODEL: TRANSFER LEARNING")
print("="*60)

validation_generator.reset()
Y_pred_tl = model_tl.predict(validation_generator, verbose=1)
y_pred_tl = np.argmax(Y_pred_tl, axis=1)
y_true_tl = validation_generator.classes

print("\n=== Classification Report (Transfer Learning) ===")
print(classification_report(y_true_tl, y_pred_tl, target_names=CLASS_NAMES))

cm_tl = confusion_matrix(y_true_tl, y_pred_tl)
plt.figure(figsize=(6, 5))
sns.heatmap(cm_tl, annot=True, fmt='d', cmap='Greens',
xticklabels=CLASS_NAMES, yticklabels=CLASS_NAMES)
plt.title("Confusion Matrix - Transfer Learning", fontweight='bold')
plt.ylabel("True Label"); plt.xlabel("Predicted Label")
plt.tight_layout()
plt.show()

misclassified_tl, error_stats_tl = error_analysis(model_tl,
validation_generator, CLASS_NAMES, "TRANSFER LEARNING")

```

---

MELATIH MODEL DENGAN TRANSFER LEARNING

=====

Downloading data from [https://storage.googleapis.com/keras-applications/efficientnetb0\\_notop.h5](https://storage.googleapis.com/keras-applications/efficientnetb0_notop.h5)

16705208/16705208 ————— 0s 0us/step

Epoch 1/5

110/110 ————— 96s 610ms/step - accuracy: 0.5296 -  
loss: 1.0643 - val\_accuracy: 0.5455 - val\_loss: 0.7288 - learning\_rate: 0.001  
0

Epoch 2/5

110/110 ————— 51s 461ms/step - accuracy: 0.5731 -  
loss: 0.8450 - val\_accuracy: 0.5455 - val\_loss: 0.7317 - learning\_rate: 0.001  
0

Epoch 3/5

110/110 ————— 54s 492ms/step - accuracy: 0.5621 -  
loss: 0.8029 - val\_accuracy: 0.5455 - val\_loss: 0.7160 - learning\_rate: 0.001  
0

Epoch 4/5

110/110 ————— 53s 484ms/step - accuracy: 0.5848 -  
loss: 0.7792 - val\_accuracy: 0.5432 - val\_loss: 0.7091 - learning\_rate: 0.001  
0

Epoch 5/5

110/110 ————— 52s 470ms/step - accuracy: 0.5772 -  
loss: 0.7599 - val\_accuracy: 0.5386 - val\_loss: 0.6964 - learning\_rate: 0.001  
0

Epoch 1/12

110/110 ————— 143s 595ms/step - accuracy: 0.5896  
- loss: 0.8081 - val\_accuracy: 0.4545 - val\_loss: 0.9762 - learning\_rate: 1.0  
000e-05

Epoch 2/12

110/110 ————— 56s 510ms/step - accuracy: 0.7978 -  
loss: 0.4776 - val\_accuracy: 0.5080 - val\_loss: 1.0482 - learning\_rate: 1.000  
0e-05

Epoch 3/12

110/110 ————— 55s 499ms/step - accuracy: 0.8747 -  
loss: 0.3534 - val\_accuracy: 0.5364 - val\_loss: 0.9457 - learning\_rate: 1.000  
0e-05

Epoch 4/12

110/110 ————— 56s 507ms/step - accuracy: 0.9078 -  
loss: 0.2837 - val\_accuracy: 0.7364 - val\_loss: 0.6145 - learning\_rate: 1.000  
0e-05

Epoch 5/12

110/110 ————— 57s 513ms/step - accuracy: 0.9379 -  
loss: 0.2299 - val\_accuracy: 0.8352 - val\_loss: 0.4348 - learning\_rate: 1.000  
0e-05

Epoch 6/12

110/110 ————— 55s 498ms/step - accuracy: 0.9364 -  
loss: 0.2132 - val\_accuracy: 0.8625 - val\_loss: 0.3824 - learning\_rate: 1.000  
0e-05

Epoch 7/12

110/110 ————— 57s 515ms/step - accuracy: 0.9502 -  
loss: 0.1828 - val\_accuracy: 0.8670 - val\_loss: 0.3676 - learning\_rate: 1.000  
0e-05

```

Epoch 8/12
110/110 ----- 56s 507ms/step - accuracy: 0.9568 -
loss: 0.1685 - val_accuracy: 0.8852 - val_loss: 0.3342 - learning_rate: 1.000
0e-05
Epoch 9/12
110/110 ----- 56s 504ms/step - accuracy: 0.9654 -
loss: 0.1485 - val_accuracy: 0.8841 - val_loss: 0.3274 - learning_rate: 1.000
0e-05
Epoch 10/12
110/110 ----- 55s 495ms/step - accuracy: 0.9637 -
loss: 0.1488 - val_accuracy: 0.8966 - val_loss: 0.2888 - learning_rate: 1.000
0e-05
Epoch 11/12
110/110 ----- 56s 513ms/step - accuracy: 0.9668 -
loss: 0.1401 - val_accuracy: 0.9091 - val_loss: 0.2906 - learning_rate: 1.000
0e-05
Epoch 12/12
110/110 ----- 55s 501ms/step - accuracy: 0.9688 -
loss: 0.1378 - val_accuracy: 0.8705 - val_loss: 0.3332 - learning_rate: 1.000
0e-05
Restoring model weights from the end of the best epoch: 11.

```

---

#### EVALUASI MODEL: TRANSFER LEARNING

---

```

28/28 ----- 15s 305ms/step

```

#### === Classification Report (Transfer Learning) ===

	precision	recall	f1-score	support
Organik	0.92	0.91	0.92	480
anorganik	0.89	0.91	0.90	400
accuracy			0.91	880
macro avg	0.91	0.91	0.91	880
weighted avg	0.91	0.91	0.91	880

#### === ERROR ANALYSIS - TRANSFER LEARNING ===

```

Total misclassified: 80/880
Error rate: 9.09%
  Organik: 44/480 (9.17%)
  anorganik: 36/400 (9.00%)

```

#

=

# 15. ANALISIS PERBANDINGAN DAN STATISTIK

#

=

```

print("\n" + "="*75)
print("ANALISIS PERBANDINGAN MODEL")

```

```

print("="*75)

plot_model_comparison(history_scratch, history_tl)
statistical_analysis_report(y_true_scratch, y_pred_scratch, y_true_tl,
y_pred_tl)

=====

ANALISIS PERBANDINGAN MODEL

=====

=====

ANALISIS STATISTIK UNTUK SKRIPSI

=====

Uji Normalitas Shapiro-Wilk: statistic=1.0000, p-value=1.0000
Cohen's Kappa - From Scratch: 0.8713
Cohen's Kappa - Transfer Learning: 0.8170
Akurasi - From Scratch: 0.9364
Akurasi - Transfer Learning: 0.9091
Perbedaan Akurasi: -0.0273

#
=====
=
# 16. SIMPAN HASIL DAN MODEL
#
=====
=
print("\n" + "="*75)
print("MENYIMPAN HASIL DAN MODEL")
print("="*75)

cr_scratch_df = pd.DataFrame(classification_report(y_true_scratch,
y_pred_scratch, target_names=CLASS_NAMES, output_dict=True)).transpose()
cr_tl_df = pd.DataFrame(classification_report(y_true_tl, y_pred_tl,
target_names=CLASS_NAMES, output_dict=True)).transpose()

cr_scratch_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/classification_report_scratch.c
sv", index=False)
cr_tl_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/classification_report_transfer_learn
ing.csv", index=False)

cm_scratch_df = pd.DataFrame(cm_scratch, index=CLASS_NAMES,
columns=CLASS_NAMES)
cm_tl_df = pd.DataFrame(cm_tl, index=CLASS_NAMES, columns=CLASS_NAMES)

cm_scratch_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/confusion_matrix_scratch.csv")
cm_tl_df.to_csv(f"{CONFIG['OUTPUT_DIR']}/confusion_matrix_transfer_learning.c
sv")

model_tl.save(f"{CONFIG['OUTPUT_DIR']}/model_efficientnetb0_waste_classifier.
h5")
print("Model transfer learning disimpan")

print(f"\nSemua hasil disimpan di: {CONFIG['OUTPUT_DIR']}")

```

```
WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or
`keras.saving.save_model(model)`. This file format is considered legacy. We r
ecommend using instead the native Keras format, e.g. `model.save('my_model.ke
ras')` or `keras.saving.save_model(model, 'my_model.keras')`.
```

---

## MENYIMPAN HASIL DAN MODEL

---

Model transfer learning disimpan

Semua hasil disimpan di: /content/output\_results

```
#
```

```
=
```

```
# 17. KONVERSI MODEL UNTUK DEPLOYMENT
```

```
#
```

```
=
```

```
print("\n" + "="*75)
print("KONVERSI MODEL UNTUK IMPLEMENTASI WEB")
print("="*75)
```

```
try:
    converter = tf.lite.TFLiteConverter.from_keras_model(model_t1)
    converter.optimizations = [tf.lite.Optimize.DEFAULT]
    tflite_model = converter.convert()
    with open(f"{CONFIG['OUTPUT_DIR']}/model_waste_classifier.tflite", "wb")
as f:
    f.write(tflite_model)
    print("Model TFLite disimpan")
    model_t1.save(f"{CONFIG['OUTPUT_DIR']}/model_waste_classifier.keras")
    print("Model Keras format disimpan")
except Exception as e:
    print(f"Konversi model gagal: {e}")
```

---

## KONVERSI MODEL UNTUK IMPLEMENTASI WEB

---

Saved artifact at '/tmp/tmpkr4wha4z'. The following endpoints are available:

\* Endpoint 'serve'

args\_0 (POSITIONAL\_ONLY): TensorSpec(shape=(None, 224, 224, 3), dtype=tf.float32, name='keras\_tensor\_253')

Output Type:

TensorSpec(shape=(None, 2), dtype=tf.float32, name=None)

Captures:

132771892219920: TensorSpec(shape=(1, 1, 1, 3), dtype=tf.float32, name=None)  
)  
132771892221456: TensorSpec(shape=(1, 1, 1, 3), dtype=tf.float32, name=None)  
)  
132771933169680: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771930847760: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771930842384: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771933167568: TensorSpec(shape=(), dtype=tf.resource, name=None)



[illegible]

[illegible]

[illegible]

[illegible]

[illegible]

132772007196048: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007196624: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007194128: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007191632: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007196432: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007197008: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007197584: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007196816: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007197968: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007195856: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007196240: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007195664: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007198352: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007198544: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007197200: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007198160: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007197776: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892216272: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892217424: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132772007194320: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892216848: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892217232: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892217040: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892216464: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892218000: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892218576: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892217808: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892218960: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892216080: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892218384: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892218768: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892217616: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892219536: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892218192: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892220304: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892215888: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892219728: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892220112: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892220688: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892221264: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892221648: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892221840: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771930848912: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771930843152: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892222224: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892222032: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892219344: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892223184: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892222992: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892222416: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892223568: TensorSpec(shape=(), dtype=tf.resource, name=None)  
132771892223376: TensorSpec(shape=(), dtype=tf.resource, name=None)

Model TFLite disimpan

Model Keras format disimpan

```
#
=====
=
# 18. BUAT FILE ZIP DENGAN SEMUA HASIL
#
=====
=
print("\n" + "="*75)
print("MEMBUAT FILE ZIP DENGAN SEMUA HASIL")
print("="*75)

import zipfile
import datetime

timestamp = datetime.datetime.now().strftime("%Y%m%d_%H%M%S")
zip_filename = f"/content/waste_classification_results_{timestamp}.zip"

with zipfile.ZipFile(zip_filename, 'w') as zipf:
    for root, dirs, files in os.walk(CONFIG['OUTPUT_DIR']):
        for file in files:
            file_path = os.path.join(root, file)
            arcname = os.path.relpath(file_path, CONFIG['OUTPUT_DIR'])
            zipf.write(file_path, arcname)

print(f"File zip berhasil dibuat: {zip_filename}")

=====
MEMBUAT FILE ZIP DENGAN SEMUA HASIL
=====
File zip berhasil dibuat: /content/waste_classification_results_20251025_0612
58.zip

#
=====
=
# 19. RINGKASAN AKHIR
#
=====
=
print("\n" + "="*75)
print("SELESAI - SEMUA TAHAP TELAH DILAKUKAN")
print("="*75)

final_accuracy_scratch = np.mean(y_true_scratch == y_pred_scratch)
final_accuracy_tl = np.mean(y_true_tl == y_pred_tl)

print(f"HASIL AKHIR:")
print(f"    • From Scratch Accuracy      : {final_accuracy_scratch:.4f}
({final_accuracy_scratch*100:.2f}%)")
print(f"    • Transfer Learning Accuracy: {final_accuracy_tl:.4f}
({final_accuracy_tl*100:.2f}%)")
print(f"    • Peningkatan                  : {final_accuracy_tl -
final_accuracy_scratch:.4f} ({(final_accuracy_tl -
final_accuracy_scratch)*100:.2f}%)")
```

```

print(f"\nOUTPUT YANG DIHASILKAN:")
print(f"    • Dataset Metadata          : {CONFIG['OUTPUT_DIR']}/dataset_metadata.csv")
print(f"    • Image Quality Analysis       : {CONFIG['OUTPUT_DIR']}/image_quality_analysis.csv")
print(f"    • Classification Reports       : {CONFIG['OUTPUT_DIR']}/classification_report_*.csv")
print(f"    • Confusion Matrices           : {CONFIG['OUTPUT_DIR']}/confusion_matrix_*.csv")
print(f"    • Model Files                   : {CONFIG['OUTPUT_DIR']}/model_*.")
print(f"    • Zip File                      : {zip_filename}")

print("\nEksperimen selesai. Semua data dan model telah disimpan.")
print("="*75)

```

---

SELESAI — SEMUA TAHAP TELAH DILAKUKAN

---

HASIL AKHIR:

- From Scratch Accuracy : 0.9364 (93.64%)
- Transfer Learning Accuracy: 0.9091 (90.91%)
- Peningkatan : -0.0273 (-2.73%)

OUTPUT YANG DIHASILKAN:

• Dataset Metadata	: /content/output_results/dataset_metadata.csv
• Image Quality Analysis	: /content/output_results/image_quality_analysis.csv
• Classification Reports	: /content/output_results/classification_report_*.csv
• Confusion Matrices	: /content/output_results/confusion_matrix_*.csv
• Model Files	: /content/output_results/model_*
• Zip File	: /content/waste_classification_results_20251025_061258.zip

Eksperimen selesai. Semua data dan model telah disimpan.

---