

An Improved Densenet Deep Neural Network Model for Tuberculosis Detection Using Chest X-Ray Images

Spring 2025: Neural Networks & Deep Learning- Mini Project

Name: Veera Manikanta Kumar Allada

Student ID: 700756934

Github Link: <https://github.com/maniallada9/Neural-Networks-deep-Learning>

```
[1] # Mount Google Drive
    from google.colab import drive
    drive.mount('/content/drive')
```

↪ Mounted at /content/drive

```

▶ def prepare_dataset():
    # Google Drive mounted path
    MONT_PATH = "/content/drive/MyDrive/DataSet/MontgomerySet/CXR_png"
    SHENZHEN_PATH = "/content/drive/MyDrive/DataSet/ChinaSet_AllFiles"
    TBX11K_PATH = "/content/drive/MyDrive/TB_Dataset/TBnNormal"

    OUTPUT_PATH = "/content/filtered_dataset"
    TB_DIR = os.path.join(OUTPUT_PATH, "TB")
    NORMAL_DIR = os.path.join(OUTPUT_PATH, "Normal")

    os.makedirs(TB_DIR, exist_ok=True)
    os.makedirs(NORMAL_DIR, exist_ok=True)

    def copy_resize(img_path, label, count):
        target_dir = TB_DIR if label == "TB" else NORMAL_DIR
        try:
            img = Image.open(img_path)
            img.verify() # Check image validity
            img = Image.open(img_path).convert("RGB").resize((224, 224))
            img.save(os.path.join(target_dir, f"{label}_{count}.jpg"))
        except (UnidentifiedImageError, OSError) as e:
            print(f"Skipped {img_path}: {e}")

    image_records = []

    # Montgomery: TB images have '_1' in name, Normal have '_0'
    for img in Path(MONT_PATH).rglob("*.png"):

```

```

elif "_1.png" in name:
    label = "TB"
else:
    continue
image_records.append((img, label))

# Shenzhen: Same logic
for img in Path(SHENZHEN_PATH).rglob("*.png"):
    name = img.name
    if "_0.png" in name:
        label = "Normal"
    elif "_1.png" in name:
        label = "TB"
    else:
        continue
    image_records.append((img, label))

# TBX11K - already separated
tbx_tb_path = os.path.join(TBX11K_PATH, "PULMONARY_TUBERCULOSIS")
tbx_normal_path = os.path.join(TBX11K_PATH, "NORMAL")

for img in Path(tbx_tb_path).rglob("*.jpg"):
    image_records.append((img, "TB"))
for img in Path(tbx_normal_path).rglob("*.jpg"):
    image_records.append((img, "Normal"))

# Shuffle and limit
random.shuffle(image_records)
tb_images = [img for img in image_records if img[1] == "TB"][:2500]
normal_images = [img for img in image_records if img[1] == "Normal"][:2500]

print(f"Total TB images found: {len(tb_images)}")
print(f"Total Normal images found: {len(normal_images)}")

```

```

▶ def get_data_loaders(data_dir, batch_size=32):
    transform = transforms.Compose([
        transforms.Resize((224, 224)),
        transforms.RandomHorizontalFlip(),
        transforms.ToTensor(),
        transforms.Normalize([0.485, 0.456, 0.406],
                              [0.229, 0.224, 0.225])
    ])

    dataset = datasets.ImageFolder(data_dir, transform=transform)
    train_size = int(0.8 * len(dataset))
    test_size = len(dataset) - train_size
    train_dataset, test_dataset = torch.utils.data.random_split(dataset, [train_size, test_size])

    train_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True)
    test_loader = DataLoader(test_dataset, batch_size=batch_size, shuffle=False)

    return train_loader, test_loader

```

```

▶ class ChannelAttention(nn.Module):
    def __init__(self, in_planes, reduction=16):
        super(ChannelAttention, self).__init__()
        self.avg_pool = nn.AdaptiveAvgPool2d(1)
        self.max_pool = nn.AdaptiveMaxPool2d(1)
        self.fc = nn.Sequential(
            nn.Conv2d(in_planes, in_planes // reduction, 1, bias=False),
            nn.ReLU(),
            nn.Conv2d(in_planes // reduction, in_planes, 1, bias=False)
        )
        self.sigmoid = nn.Sigmoid()

    def forward(self, x):
        avg_out = self.fc(self.avg_pool(x))
        max_out = self.fc(self.max_pool(x))
        return self.sigmoid(avg_out + max_out)

class SpatialAttention(nn.Module):
    def __init__(self, kernel_size=7):
        super(SpatialAttention, self).__init__()
        padding = kernel_size // 2
        self.conv = nn.Conv2d(2, 1, kernel_size, padding=padding, bias=False)
        self.sigmoid = nn.Sigmoid()

    def forward(self, x):
        avg = torch.mean(x, dim=1, keepdim=True)
        max_, _ = torch.max(x, dim=1, keepdim=True)
        x = torch.cat([avg, max_], dim=1)

```

```

▶ def train_and_evaluate():
    device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
    train_loader, test_loader = get_dataloaders("/content/filtered_dataset", batch_size=32)
    model = CBAMWDnet(num_classes=2).to(device)
    criterion = nn.CrossEntropyLoss()
    optimizer = optim.Adam(model.parameters(), lr=1e-4)

    train_losses = []

    for epoch in range(10):
        model.train()
        total_loss = 0
        for imgs, labels in train_loader:
            imgs, labels = imgs.to(device), labels.to(device)
            optimizer.zero_grad()
            outputs = model(imgs)
            loss = criterion(outputs, labels)
            loss.backward()
            optimizer.step()
            total_loss += loss.item()
        print(f"Epoch {epoch+1}, Loss: {total_loss / len(train_loader):.4f}")

    # Evaluation
    model.eval()
    all_preds, all_labels = [], []
    with torch.no_grad():
        for imgs, labels in test_loader:
            imgs = imgs.to(device)
            outputs = model(imgs)
            preds = torch.argmax(outputs, dim=1).cpu()

```

