

```
In [1]: import os
import torch
import torch.nn as nn
import torch.optim as optim
from torch.utils.data import Dataset, DataLoader, random_split
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.metrics import accuracy_score
from tqdm import tqdm
import warnings
warnings.filterwarnings('ignore')
torch.manual_seed(42)
np.random.seed(42)
```

1.数据集加载

```
In [2]: class MyDataset(Dataset):
        """自定义数据集类"""
        def __init__(self, csv_file):
            self.data = pd.read_csv(csv_file, header=None)
            # 随机打乱数据
            self.data = self.data.sample(frac=1, random_state=42).reset_index(drop=True)

        def __len__(self):
            return len(self.data)

        def __getitem__(self, idx):
            features = self.data.iloc[idx, :-1].values.astype(np.float32)
            label = self.data.iloc[idx, -1].astype(np.int64)
            label = label - 1
            return torch.tensor(features), torch.tensor(label)

# 创建数据集并查看基本信息
dataset = MyDataset("dataset.csv")
print("类别分布:")
print(dataset.data.iloc[:, -1].value_counts().sort_index().rename_axis(None))

# 划分训练集和测试集
train_size = int(0.9 * len(dataset))
test_size = len(dataset) - train_size
train_dataset, test_dataset = random_split(dataset, [train_size, test_size])
print(f"训练集大小: {len(train_dataset)}")
print(f"测试集大小: {len(test_dataset)}")
# # 创建数据加载器
# batch_size = 64
# train_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True)
# test_loader = DataLoader(test_dataset, batch_size=batch_size, shuffle=False)
```

类别分布:

```
1    1000
2    1000
3    1000
4    1000
```

Name: count, dtype: int64

训练集大小: 3600

测试集大小: 400

```
In [3]: def plot_dataset(dataset):
        """可视化数据集"""
        features = dataset.data.iloc[:, :-1].values
        labels = dataset.data.iloc[:, -1].values
```

```

label_names = ['Class 1', 'Class 2', 'Class 3', 'Class 4']

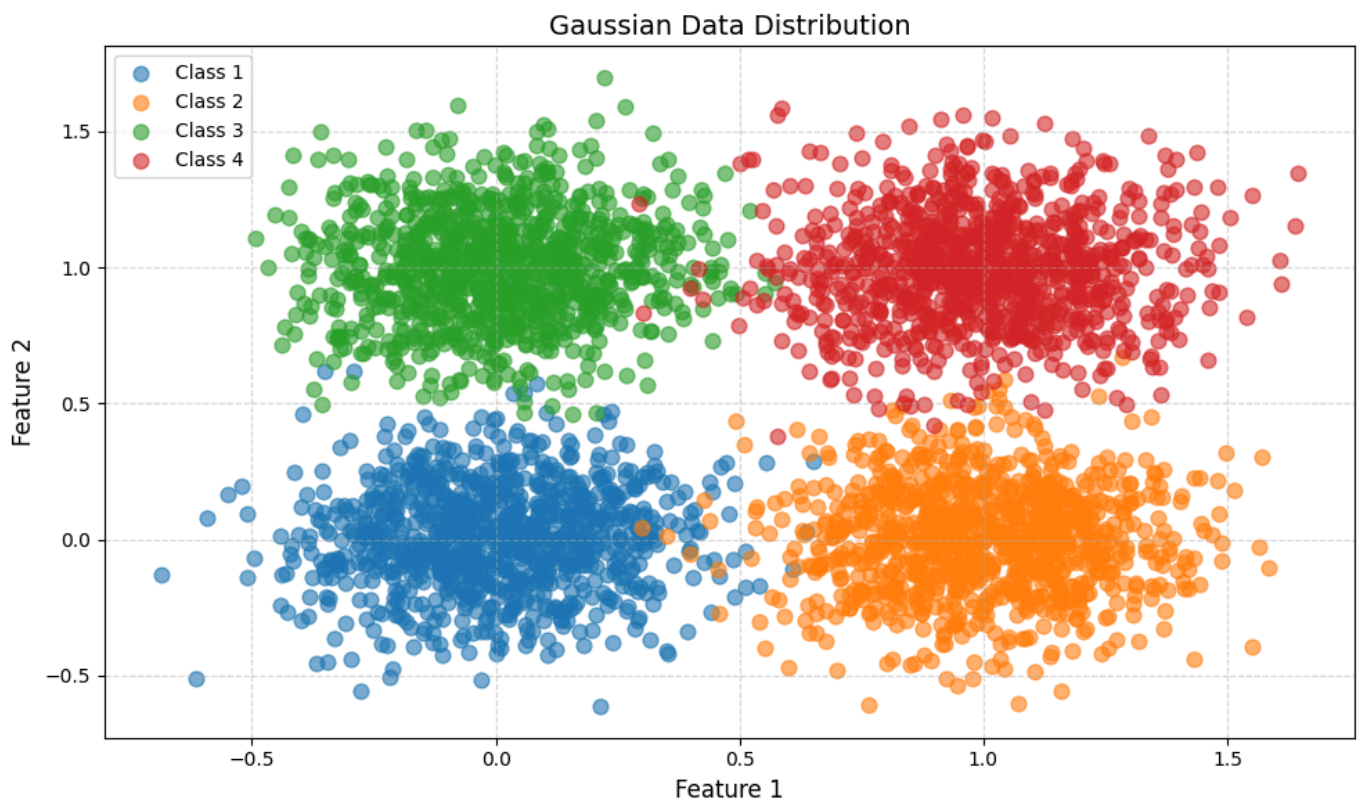
plt.figure(figsize=(10, 6))

for i, color in enumerate(label_names):
    class_features = features[labels == i + 1]
    plt.scatter(class_features[:, 0], class_features[:, 1],
                label=label_names[i], alpha=0.6, s=60)

plt.xlabel('Feature 1', fontsize=12)
plt.ylabel('Feature 2', fontsize=12)
plt.title('Gaussian Data Distribution', fontsize=14)
plt.grid(True, linestyle='--', alpha=0.5)
plt.legend()
plt.tight_layout()
plt.show()

plot_dataset(dataset)

```



2.前馈神经网络

```

In [4]: class FeedForwardNN(nn.Module):
        """前馈神经网络"""
        def __init__(
            self,
            input_size=2,
            hidden_sizes=[64],
            output_size=4,
            activation='relu',
            dropout_rate=0.0
        ):
            super(FeedForwardNN, self).__init__()
            # 构建网络
            layers = []
            prev_size = input_size
            for hidden_size in hidden_sizes:
                layers.append(nn.Linear(prev_size, hidden_size))
            # 激活函数
            if activation == 'relu':
                layers.append(nn.ReLU())

```

```

        elif activation == 'sigmoid':
            layers.append(nn.Sigmoid())
        elif activation == 'tanh':
            layers.append(nn.Tanh())
        elif activation == 'leaky_relu':
            layers.append(nn.LeakyReLU())
        else:
            layers.append(nn.ReLU())
        # 添加dropout
        if dropout_rate > 0.0:
            layers.append(nn.Dropout(dropout_rate))
        prev_size = hidden_size
        layers.append(nn.Linear(prev_size, output_size))
        self.network = nn.Sequential(*layers)
        # 保存网络参数
        self.hidden_sizes = hidden_sizes
        self.activation = activation

    def forward(self, x):
        return self.network(x)

def create_model(model_config):
    """根据配置创建模型"""
    return FeedForwardNN(
        input_size=2,
        hidden_sizes=model_config.get('hidden_sizes', [64]),
        output_size=4,
        activation=model_config.get('activation', 'relu'),
        dropout_rate=model_config.get('dropout_rate', 0.0)
    )

```

3.网络训练

```

In [ ]: def train_model(model, config, num_epochs=100):
    """训练模型"""
    # 配置信息
    name = config['name']
    batch_size = config.get('batch_size', 32)
    learning_rate = config.get('learning_rate', 0.001)

    # 创建日志文件夹和文件
    log_dir = "output"
    os.makedirs(log_dir, exist_ok=True)
    log_file = os.path.join(log_dir, f"{name}.log")

    with open(log_file, "w") as log:
        log.write("Epoch,Train Loss,Train Accuracy,Test Loss,Test Accuracy\n")

    # 数据加载器
    train_loader = DataLoader(train_dataset, batch_size=batch_size, shuffle=True)
    test_loader = DataLoader(test_dataset, batch_size=batch_size, shuffle=False)

    # 定义损失函数和优化器
    criterion = nn.CrossEntropyLoss()
    optimizer = optim.SGD(model.parameters(), lr=learning_rate)

    # 记录训练过程
    train_losses = []
    test_losses = []
    train_accuracies = []
    test_accuracies = []

    # 训练循环
    for epoch in tqdm(range(num_epochs), desc=f"Training {config['name'][:12]}"):

```

```

# 训练模式
model.train()
epoch_train_loss = 0.0
train_preds = []
train_labels = []

for batch_idx, (data, target) in enumerate(train_loader):
    optimizer.zero_grad()
    output = model(data)
    loss = criterion(output, target)
    loss.backward()
    optimizer.step()

    epoch_train_loss += loss.item()

    # 收集预测结果
    _, predicted = torch.max(output.data, 1)
    train_preds.extend(predicted.cpu().numpy())
    train_labels.extend(target.cpu().numpy())

# 计算训练损失和准确率
train_loss = epoch_train_loss / len(train_loader)
train_losses.append(train_loss)
train_acc = accuracy_score(train_labels, train_preds)
train_accuracies.append(train_acc)

# 测试模式
model.eval()
epoch_test_loss = 0.0
test_preds = []
test_labels = []

with torch.no_grad():
    for data, target in test_loader:
        output = model(data)
        loss = criterion(output, target)
        epoch_test_loss += loss.item()

        _, predicted = torch.max(output.data, 1)
        test_preds.extend(predicted.cpu().numpy())
        test_labels.extend(target.cpu().numpy())

# 计算测试损失和准确率
test_loss = epoch_test_loss / len(test_loader)
test_losses.append(test_loss)
test_acc = accuracy_score(test_labels, test_preds)
test_accuracies.append(test_acc)

# 写入日志文件
with open(log_file, "a") as log:
    log.write(f"{epoch+1},{train_loss:.4f},{train_acc:.4f},{test_loss:.4f},{test_acc:.4f}")

# 最终评估
final_train_accuracy = train_accuracies[-1]
final_test_accuracy = test_accuracies[-1]

return {
    'model': model,
    'train_losses': train_losses,
    'test_losses': test_losses,
    'train_accuracies': train_accuracies,
    'test_accuracies': test_accuracies,
    'final_train_accuracy': final_train_accuracy,
    'final_test_accuracy': final_test_accuracy,
}

```

```
'config': config  
}
```

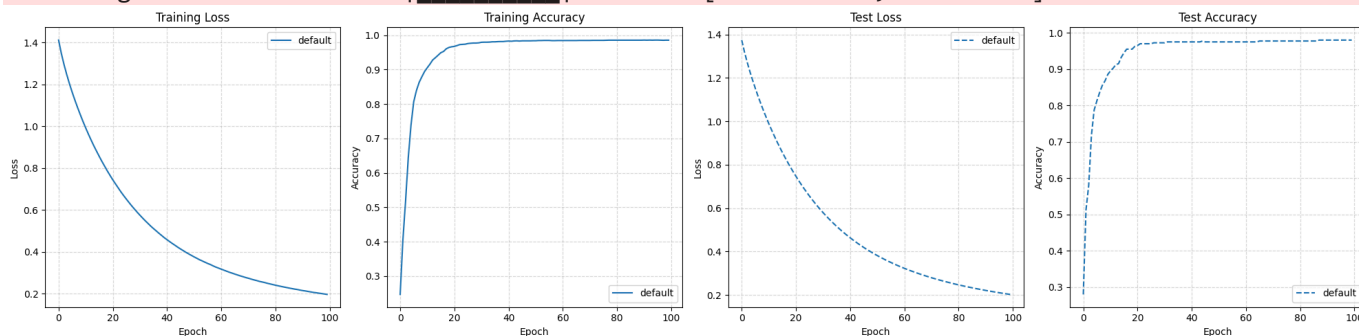
4.绘制结果

```
In [6]: def plot_results(results):  
        """绘制训练结果"""  
        plt.figure(figsize=(20, 5))  
  
        # 训练损失曲线  
        ax1 = plt.subplot(141)  
        for model_key, result in results.items():  
            ax1.plot(result['train_losses'], label=f"{result['config']['name']}")  
        ax1.set_xlabel('Epoch')  
        ax1.set_ylabel('Loss')  
        ax1.set_title('Training Loss')  
        ax1.legend()  
        ax1.grid(True, linestyle='--', alpha=0.5)  
  
        # 训练准确率曲线  
        ax2 = plt.subplot(142)  
        for model_key, result in results.items():  
            ax2.plot(result['train_accuracies'], label=f"{result['config']['name']}")  
        ax2.set_xlabel('Epoch')  
        ax2.set_ylabel('Accuracy')  
        ax2.set_title('Training Accuracy')  
        ax2.legend()  
        ax2.grid(True, linestyle='--', alpha=0.5)  
  
        # 测试损失曲线  
        ax3 = plt.subplot(143)  
        for model_key, result in results.items():  
            ax3.plot(result['test_losses'], linestyle='--', label=f"{result['config']['name']}")  
        ax3.set_xlabel('Epoch')  
        ax3.set_ylabel('Loss')  
        ax3.set_title('Test Loss')  
        ax3.legend()  
        ax3.grid(True, linestyle='--', alpha=0.5)  
  
        # 测试准确率曲线  
        ax4 = plt.subplot(144)  
        for model_key, result in results.items():  
            ax4.plot(result['test_accuracies'], linestyle='--', label=f"{result['config']['name']}")  
        ax4.set_xlabel('Epoch')  
        ax4.set_ylabel('Accuracy')  
        ax4.set_title('Test Accuracy')  
        ax4.legend()  
        ax4.grid(True, linestyle='--', alpha=0.5)  
  
        plt.tight_layout()  
        plt.show()
```

```
In [7]: # 默认配置  
default_config = {  
    'name': 'default',  
    'hidden_sizes': [64],  
    'activation': 'relu',  
    'dropout_rate': 0.0,  
    'batch_size': 32,  
    'learning_rate': 0.001  
}  
  
results = {}  
model = create_model(default_config)
```

```
results['default'] = train_model(model, default_config, num_epochs=100)
plot_results(results)
```

Training default : 100% |██████████| 100/100 [00:59<00:00, 1.69it/s]



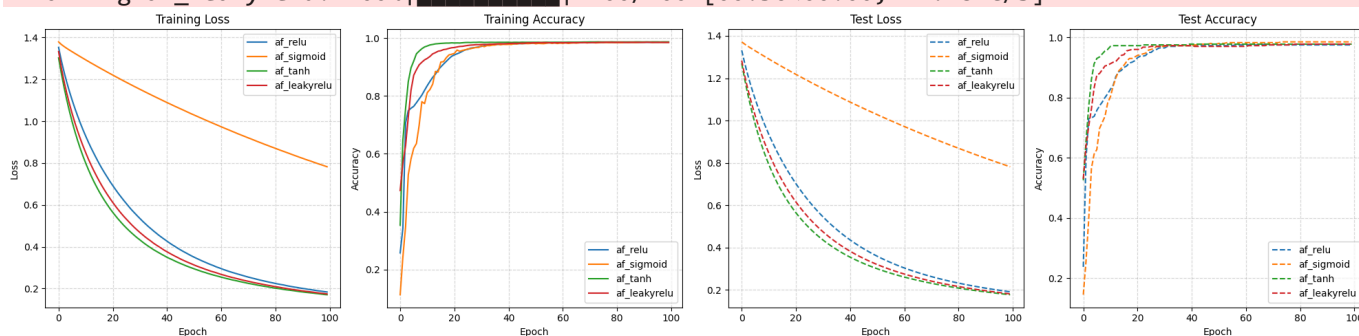
In [8]: # 不同激活函数

```
configs = {
    '1': {
        'name': 'af_relu',
        'activation': 'relu'
    },
    '2': {
        'name': 'af_sigmoid',
        'activation': 'sigmoid'
    },
    '3': {
        'name': 'af_tanh',
        'activation': 'tanh'
    },
    '4': {
        'name': 'af_leakyrelu',
        'activation': 'leaky_relu'
    }
}

results = {}
for model_key, config in configs.items():
    # 创建模型
    model = create_model(config)
    # 训练模型
    results[model_key] = train_model(model, config, num_epochs=100)

plot_results(results)
```

Training af_relu : 100% |██████████| 100/100 [01:06<00:00, 1.50it/s]
 Training af_sigmoid : 100% |██████████| 100/100 [01:24<00:00, 1.19it/s]
 Training af_tanh : 100% |██████████| 100/100 [01:09<00:00, 1.44it/s]
 Training af_leakyrelu: 100% |██████████| 100/100 [00:56<00:00, 1.78it/s]



In [9]: # 不同隐藏层大小

```
configs = {
    '1': {
        'name': 'hs_16',
        'hidden_sizes': [16]
    },
    '2': {
```

```

        'name': 'hs_32',
        'hidden_sizes': [32]
    },
    '3': {
        'name': 'hs_64',
        'hidden_sizes': [64]
    },
    '4': {
        'name': 'hs_128',
        'hidden_sizes': [128]
    }
}

results = {}
for model_key, config in configs.items():
    # 创建模型
    model = create_model(config)
    # 训练模型
    results[model_key] = train_model(model, config, num_epochs=100)

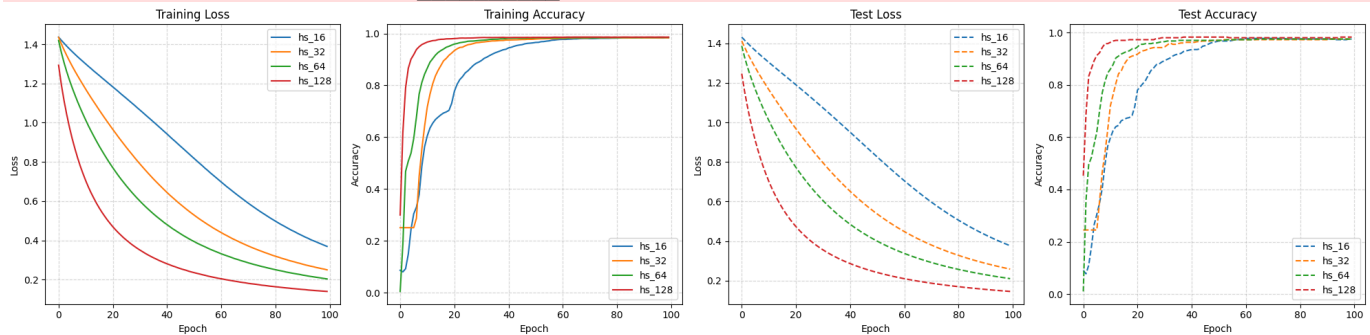
plot_results(results)

```

```

Training hs_16      : 100%|██████████| 100/100 [00:56<00:00, 1.76it/s]
Training hs_32      : 100%|██████████| 100/100 [01:05<00:00, 1.53it/s]
Training hs_64      : 100%|██████████| 100/100 [01:24<00:00, 1.19it/s]
Training hs_128     : 100%|██████████| 100/100 [01:27<00:00, 1.14it/s]

```



In [10]: # 不同网络层数

```

configs = {
    '1': {
        'name': 'layers_2',
        'hidden_sizes': [64]
    },
    '2': {
        'name': 'layers_3',
        'hidden_sizes': [64,64]
    },
    '3': {
        'name': 'layers_4',
        'hidden_sizes': [64,64,64]
    },
    '4': {
        'name': 'layers_5',
        'hidden_sizes': [64,64,64,64]
    }
}

results = {}
for model_key, config in configs.items():
    # 创建模型
    model = create_model(config)
    # 训练模型
    results[model_key] = train_model(model, config, num_epochs=100)

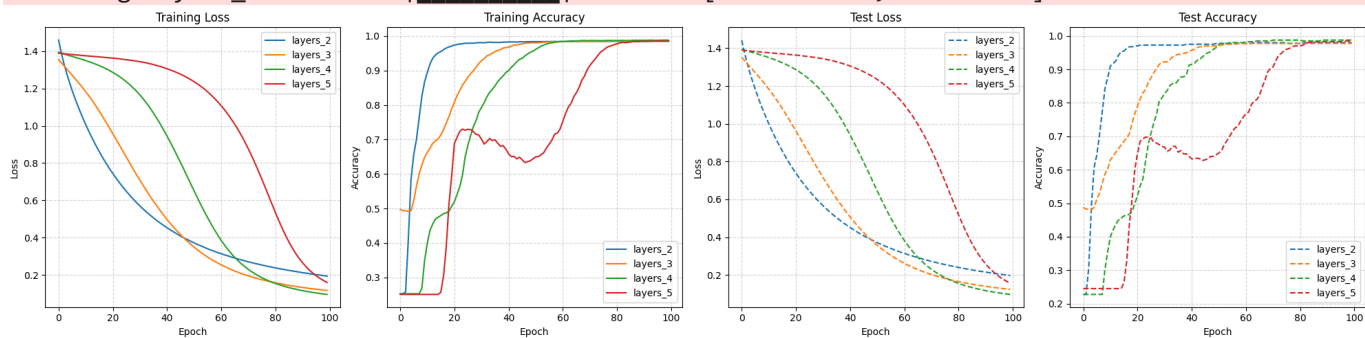
plot_results(results)

```

```

Training layers_2      : 100%|██████████| 100/100 [01:20<00:00, 1.24it/s]
Training layers_3      : 100%|██████████| 100/100 [01:14<00:00, 1.34it/s]
Training layers_4      : 100%|██████████| 100/100 [01:00<00:00, 1.65it/s]
Training layers_5      : 100%|██████████| 100/100 [01:29<00:00, 1.12it/s]

```



In [11]: # 不同batch size

```

configs = {
    '1': {
        'name': 'bs_16',
        'batch_size': 16
    },
    '2': {
        'name': 'bs_32',
        'batch_size': 32
    },
    '3': {
        'name': 'bs_64',
        'batch_size': 64
    },
    '4': {
        'name': 'bs_128',
        'batch_size': 128
    }
}

results = {}
for model_key, config in configs.items():
    # 创建模型
    model = create_model(config)
    # 训练模型
    results[model_key] = train_model(model, config, num_epochs=100)

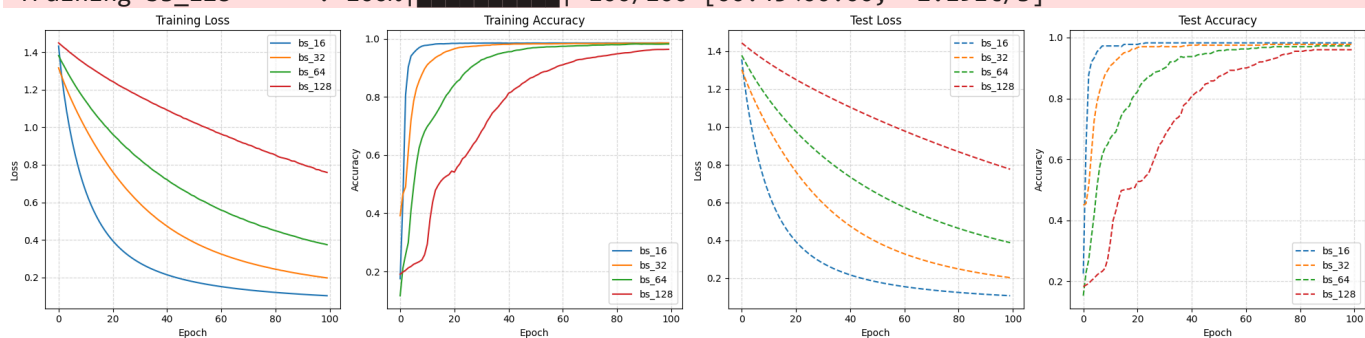
plot_results(results)

```

```

Training bs_16         : 100%|██████████| 100/100 [01:34<00:00, 1.06it/s]
Training bs_32         : 100%|██████████| 100/100 [01:18<00:00, 1.27it/s]
Training bs_64         : 100%|██████████| 100/100 [01:02<00:00, 1.60it/s]
Training bs_128        : 100%|██████████| 100/100 [00:45<00:00, 2.19it/s]

```



In [12]: # 不同learning rate

```

configs = {
    '1': {
        'name': 'lr_0.1',
        'learning_rate': 0.1
    },

```



```

'2': {
    'name': 'lr_0.01',
    'learning_rate': 0.01
},
'3': {
    'name': 'lr_0.001',
    'learning_rate': 0.001
},
'4': {
    'name': 'lr_0.0001',
    'learning_rate': 0.0001
}
}

results = {}
for model_key, config in configs.items():
    # 创建模型
    model = create_model(config)
    # 训练模型
    results[model_key] = train_model(model, config, num_epochs=100)

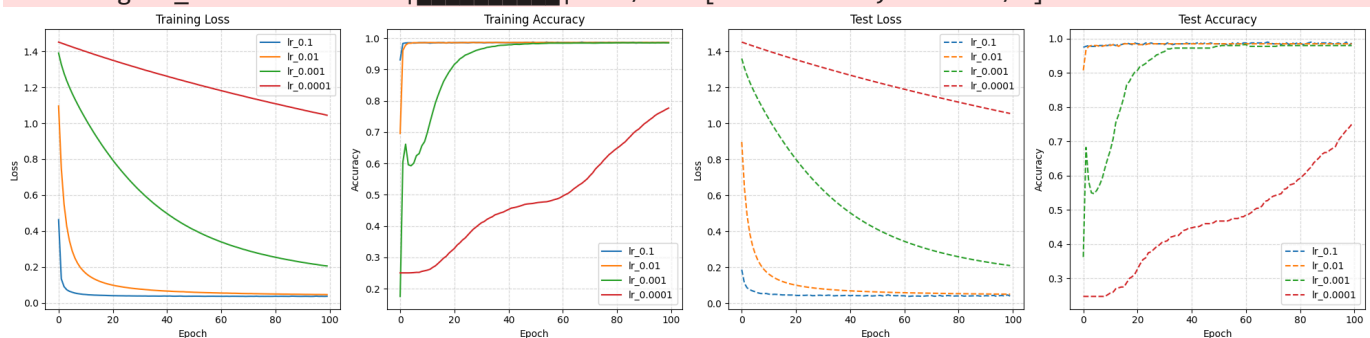
plot_results(results)

```

```

Training lr_0.1      : 100%|██████████| 100/100 [00:55<00:00, 1.80it/s]
Training lr_0.01     : 100%|██████████| 100/100 [00:55<00:00, 1.81it/s]
Training lr_0.001    : 100%|██████████| 100/100 [01:07<00:00, 1.48it/s]
Training lr_0.0001   : 100%|██████████| 100/100 [01:17<00:00, 1.28it/s]

```



In [13]: # 不同dropout rate

```

configs = {
    '1': {
        'name': 'dr_0.0',
        'dropout_rate': 0.0
    },
    '2': {
        'name': 'dr_0.1',
        'dropout_rate': 0.1
    },
    '3': {
        'name': 'dr_0.2',
        'dropout_rate': 0.2
    },
    '4': {
        'name': 'dr_0.3',
        'dropout_rate': 0.2
    }
}

results = {}
for model_key, config in configs.items():
    # 创建模型
    model = create_model(config)
    # 训练模型
    results[model_key] = train_model(model, config, num_epochs=100)

```

```
plot_results(results)
```

```
Training dr_0.0      : 100%|██████████| 100/100 [01:01<00:00, 1.62it/s]
Training dr_0.1      : 100%|██████████| 100/100 [01:19<00:00, 1.26it/s]
Training dr_0.2      : 100%|██████████| 100/100 [01:20<00:00, 1.24it/s]
Training dr_0.3      : 100%|██████████| 100/100 [01:25<00:00, 1.16it/s]
```

