PROGRAM 1

PROBABILITY

```
total outcomes = 6
favorable_outcomes = 1
probability_4 = favorable_outcomes / total_outcomes
print(f"Probability of rolling a 4: {probability_4}")
import numpy as np
import matplotlib.pyplot as plt
from scipy.stats import norm, poisson, binom, expon
mean = 50
std dev = 10
samples = np.random.normal(mean, std dev, 1000)
plt.figure(figsize=(8, 6))
plt.hist(samples, bins=30, density=True, alpha=0.6, color='blue')
x = np.linspace(mean - 4*std dev, mean + 4*std dev, 100)
plt.plot(x, norm.pdf(x, mean, std dev), 'r-', lw=2, label='Normal Distribution')
plt.title('Normal Distribution Example (Quality Control)')
plt.xlabel('Values')
plt.ylabel('Probability Density')
plt.legend()
plt.grid(True)
plt.show()
lambda_param = 5
k = 3
prob 3 events = poisson.pmf(k, lambda param)
print(f"Probability of 3 events occurring in an hour: {prob 3 events}")
```

```
n = 10
p = 0.6
k_success = 7
prob 7 success = binom.pmf(k success, n, p)
print(f"Probability of 7 successes out of 10 trials: {prob 7 success}")
exp samples = np.random.exponential(scale=2, size=1000)
plt.figure(figsize=(8, 6))
plt.hist(exp samples, bins=30, density=True, alpha=0.6, color='green')
x exp = np.linspace(0, 10, 100)
plt.plot(x_exp, expon.pdf(x_exp, scale=2), 'r-', lw=2, label='Exponential
Distribution')
plt.title('Exponential Distribution Example (Reliability Analysis)')
plt.xlabel('Values')
plt.ylabel('Probability Density')
plt.legend()
plt.grid(True)
plt.show()
```

PROGRAM 2

REGRESSION

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split

```
from sklearn.metrics import mean squared error, r2 score
from sklearn.linear model import LogisticRegression
from sklearn.datasets import load iris
np.random.seed(42)
X = np.random.rand(100, 1) * 10
y = 2 * X.squeeze() + np.random.randn(100) * 2
plt.figure(figsize=(8, 4))
plt.scatter(X, y)
plt.title('Scatter Plot')
plt.xlabel('X')
plt.ylabel('Y')
plt.grid(True)
correlation coefficient = np.corrcoef(X.squeeze(), y)[0, 1]
print(f"Correlation Coefficient: {correlation coefficient}")
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
random state=42)
lin reg = LinearRegression()
lin reg.fit(X train, y train)
y_pred = lin_reg.predict(X_test)
mse = mean_squared_error(y_test, y_pred)
r2 = r2 score(y test, y pred)
print(f"Mean Squared Error: {mse}")
print(f"R-squared Score: {r2}")
plt.figure(figsize=(8, 4))
plt.scatter(X_test, y_test, color='black')
plt.plot(X_test, y_pred, color='blue', linewidth=3)
```

```
plt.title('Linear Regression Prediction')
plt.xlabel('X')
plt.ylabel('Y')
plt.grid(True)
iris = load iris()
X iris = iris.data[:, :2]
y_iris = iris.target
log_reg = LogisticRegression()
log reg.fit(X iris, y iris)
x_{min}, x_{max} = X_{iris}[:, 0].min() - 1, X_{iris}[:, 0].max() + 1
y_min, y_max = X_iris[:, 1].min() - 1, X_iris[:, 1].max() + 1
xx, yy = np.meshgrid(np.arange(x min, x max, 0.1), np.arange(y min, y max,
0.1))
Z = log_reg.predict(np.c_[xx.ravel(), yy.ravel()])
Z = Z.reshape(xx.shape)
plt.figure(figsize=(8, 6))
plt.contourf(xx, yy, Z, alpha=0.4)
plt.scatter(X iris[:, 0], X iris[:, 1], c=y iris, s=20, edgecolor='k')
plt.title('Logistic Regression (Iris dataset)')
plt.xlabel('Sepal Length')
plt.ylabel('Sepal Width')
plt.grid(True)
plt.show()
```