# **Probability Distribution Problems**

## **Problem 1: Simulating a Normal Distribution**

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Objective: Generate and plot a normal distribution with specified mean and standard deviation.

```
% Parameters
mu = 0; % Mean
sigma = 1; % Standard Deviation

% Generate a normal distribution
x = -3*sigma:0.1:3*sigma;
y = normpdf(x, mu, sigma);

% Plot
figure;
plot(x, y);
title('Normal Distribution');
xlabel('Value');
ylabel('Probability Density');
```

**Problem 2: Calculating Binomial Probabilities** 

Objective: Calculate and plot the probabilities for a binomial distribution.

```
% Parameters

n = 10; % Number of trials

p = 0.5; % Probability of success
```

```
% Calculate binomial probabilities

x = 0:n;

y = binopdf(x, n, p);

% Plot

figure;

bar(x, y);

title('Binomial Distribution');

xlabel('Number of Successes');

ylabel('Probability');
```

#### **Problem 3: Poisson Distribution**

Objective: Plot the Poisson distribution for different mean values.

```
% Parameters

lambda = [2, 4, 6]; % Different mean values

% Generate and plot Poisson distributions

figure;

for i = 1:length(lambda)

x = 0:15;

y = poisspdf(x, lambda(i));

subplot(1, length(lambda), i);

bar(x, y);
```

```
title(['Poisson Distribution with \lambda = ', num2str(lambda(i))]);
xlabel('Number of Events');
ylabel('Probability');
end
```

# Problem 4: Exponential Distribution

Objective: Generate an exponential distribution and plot its probability density function.

```
% Parameter

lambda = 1; % Rate parameter

% Generate exponential distribution

x = 0:0.1:10;

y = exppdf(x, 1/lambda);

% Plot

figure;

plot(x, y);

title('Exponential Distribution');

xlabel('Value');

ylabel('Probability Density');
```

## **Problem 5: Uniform Distribution**

Objective: Simulate and visualize a uniform distribution.

```
% Parameters
a = 0; % Lower bound
b = 1; % Upper bound
% Generate uniform distribution
x = a:0.01:b;
y = unifpdf(x, a, b);
 % Plot
 figure;
 plot(x, y);
 title('Uniform Distribution');
 xlabel('Value');
 ylabel('Probability Density');
  Problem 6: Cumulative Distribution Function (CDF)
  Objective: Plot the CDF of a normal distribution.
  % Parameters for normal distribution
  mu = 0; % Mean
  sigma = 1; % Standard deviation
  % Generate CDF
  x = -3*sigma:0.1:3*sigma;
  y = normcdf(x, mu, sigma);
```

```
% Plot
figure;
plot(x, y);
title('CDF of Normal Distribution');
xlabel('Value');
ylabel('Cumulative Probability');
```

## **Problem 7: Working with Random Variables**

Objective: Generate random numbers from a normal distribution and calculate their statistics.

```
% Generate random numbers
num_samples = 1000;
samples = normrnd(0, 1, [num_samples, 1]);
% Calculate statistics
mean_val = mean(samples);
std_dev = std(samples);
% Display results
fprintf('Mean: %.2f\n', mean_val);
fprintf('Standard Deviation: %.2f\n', std_dev);
```

## **Basic Statistics with MATLAB**

1. Calculating Mean, Median, and Mode

Objective: Find the mean, median, and mode of a given data set.

```
% Data
data = [15, 9, 26, 13, 14, 12, 22, 19];
% Mean
mean_val = mean(data);
% Median
median_val = median(data);
% Mode
mode_val = mode(data);
% Display results
fprintf('Mean: %.2f\n', mean_val);
fprintf('Median: %.2f\n', median_val);
fprintf('Mode: \%.2f\n', mode\_val);
2. Standard Deviation and Variance
Objective: Compute the standard deviation and variance of a data set.
% Data
data = [15, 9, 26, 13, 14, 12, 22, 19];
```

```
% Standard Deviation
std_dev = std(data);
% Variance
variance = var(data);
% Display results
fprintf('Standard Deviation: %.2f\n', std_dev);
fprintf('Variance: %.2f\n', variance);
3. Linear Correlation Coefficient
Objective: Determine the linear correlation coefficient (Pearson's r) between two sets of data.
% Data
data_x = [1, 2, 3, 4, 5];
data_y = [2, 4, 5, 4, 5];
% Correlation Coefficient
corr_coeff = corrcoef(data_x, data_y);
% Display result
fprintf('Correlation Coefficient: %.2f\n', corr_coeff(1,2));
4. Histogram Plotting
Objective: Create a histogram to visualize the distribution of a data set.
```

```
% Data
data = [15, 9, 26, 13, 14, 12, 22, 19];
% Plot Histogram
figure;
histogram(data);
title('Data Distribution');
xlabel('Value');
ylabel('Frequency');
5. Boxplot for Data Distribution
Objective: Generate a boxplot to observe the spread and skewness of data.
% Data
data = [15, 9, 26, 13, 14, 12, 22, 19];
% Boxplot
figure;
boxplot(data);
title('Boxplot of Data');
6. Scatter Plot for Two Variables
Objective: Create a scatter plot to visualize the relationship between two variables.
% Data
data_x = [1, 2, 3, 4, 5];
data_y = [2, 4, 5, 4, 5];
```

```
% Scatter Plot
figure;
scatter(data_x, data_y);
title('Scatter Plot');
xlabel('X-axis');
ylabel('Y-axis');
7. Generating Random Data and Analyzing
Objective: Generate random data following a normal distribution and analyze it.
% Generate random data
data = normrnd(0, 1, [100, 1]); % 100 random numbers from N(0,1)
% Mean and Standard Deviation
mean_val = mean(data);
std_dev = std(data);
% Display results
fprintf('Mean of Random Data: %.2f\n', mean_val);
fprintf('Standard Deviation of Random Data: %.2f\n', std_dev);
% Plot Histogram
figure;
histogram(data);
title('Histogram of Random Data');
```

#### **Regression Analysis**

Regression analysis is a powerful statistical method for examining the relationship between a dependent variable and one or more independent variables. Here's how you can perform linear regression analysis in MATLAB:

#### **Simple Linear Regression**

Simple linear regression involves a single independent variable. Let's say we have some data x and y, and we want to fit a line y = a\*x + b.

```
% Sample Data

x = [1, 2, 3, 4, 5];

y = [2, 3, 4, 6, 5];

% Perform linear regression

p = polyfit(x, y, 1); % p(1) is slope, p(2) is intercept

% Create a linear model

y_fit = polyval(p, x);

% Plot

figure;

plot(x, y, 'o'); % original data

hold on;

plot(x, y_fit, '-'); % fitted line
```

```
title('Simple Linear Regression');
xlabel('x');
ylabel('y');
legend('Data', 'Fitted line');
```

#### **Multiple Linear Regression**

In multiple linear regression, we predict a dependent variable based on multiple independent variables.

```
% Sample Data
X = [123; 234; 345; 456; 567];% Each row is an observation
y = [2; 3; 4; 6; 5]; % Dependent variable
% Add a column of ones to X for the intercept
X = [ones(size(X, 1), 1) X];
% Perform regression
b = regress(y, X); % Returns the regression coefficients
% Predicted values
y_pred = X * b;
% Display the coefficients
disp('Coefficients (including intercept):');
disp(b);
% Plot - only practical if you have 1 or 2 independent variables
% For more variables, consider 3D plots or partial regression plots
```

#### **Polynomial Regression**

Polynomial regression fits a nonlinear relationship between the value of x and the corresponding conditional mean of y.

```
% Sample Data
 x = [1, 2, 3, 4, 5];
 y = [2, 4, 6, 8, 10];
 % Polynomial degree
 degree = 2;
 % Perform polynomial regression
 p = polyfit(x, y, degree);
 % Create a polynomial model
x_fit = linspace(min(x), max(x), 100); % 100 points for a smoother plot
y_fit = polyval(p, x_fit);
% Plot
figure;
plot(x, y, 'o'); % original data
hold on;
plot(x_fit, y_fit, '-'); % fitted polynomial
title('Polynomial Regression');
xlabel('x');
ylabel('y');
legend('Data', 'Fitted polynomial');
```