

1) a) $D_a: 1 \ 2 \ 3 \ 4 \ 5 \ 6$

$$f_{D_a}(d_a) = [\frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6}] \quad \text{for 1 roll} \quad 6 \cdot d_a = d_{a6} \quad \text{2 of 6 rolls}$$

 $f_{D_{a6}} = \text{conv}(f_{D_a}, f_{D_a})$ Repeat 5 times \Rightarrow 6 convolutions total

$$\mathbb{E}[D_{a6}] = \sum_{i=1}^6 d_{a6i} f_{D_{a6}}(d_{a6i}) = 21 = \mu_{D_{a6}}$$

$$\mathbb{E}[(D_{a6} - \mu_{D_{a6}})^2] = \sum_{i=1}^6 (d_{a6i} - \mu_{D_{a6}})^2 f_{D_{a6}}(d_{a6i}) = 17.4724 = \sigma_{D_{a6}}^2 \quad \sigma_{D_{a6}} = 4.18$$

b) $d_b: 4 \ 5 \ 6 \ 7 \ 8 \ 9$

$$f_{D_b}(d_b) = [\frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6}] \quad 6 \cdot d_b = d_{b6}$$

$$\mathbb{E}[D_{b6}] = 39 = \mu_{D_{b6}} \quad \mathbb{E}[(D_{b6} - \mu_{D_{b6}})^2] = 17.4724 = \sigma_{D_{b6}}^2 \quad \sigma_{D_{b6}} = 4.18$$

c) $d_c: 1 \ 1 \ 3 \ 3 \ 3 \ 5$

$$f_{D_c}(d_c) = [\frac{1}{3} \ \frac{1}{3} \ \frac{1}{2} \ \frac{1}{2} \ \frac{1}{2} \ \frac{1}{6}] \quad 6 \cdot d_c = d_{c6}$$

$$\mathbb{E}[D_{c6}] = 16 = \mu_{D_{c6}} \quad \mathbb{E}[(D_{c6} - \mu_{D_{c6}})^2] = 11.3569 = \sigma_{D_{c6}}^2 \quad \sigma_{D_{c6}} = 3.37$$

d) $d_1: 1 \ 2 \ 3 \ 4 \ 5 \ 6 \quad d_2: 1 \ 1 \ 3 \ 3 \ 3 \ 5$

$$f_{D_1} = [\frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6} \ \frac{1}{6}] \quad f_{D_2} = [\frac{1}{3} \ \frac{1}{3} \ \frac{1}{2} \ \frac{1}{2} \ \frac{1}{2} \ \frac{1}{6}]$$

$$3 \cdot d_1 + 3 \cdot d_2 = d_d \quad \mathbb{E}[D_d] = 18.5 = \mu_{D_d} \quad \mathbb{E}[(D_d - \mu_{D_d})^2] = 14.440 = \sigma_{D_d}^2$$

$$\sigma_{D_d} = 3.8$$

$$2) \begin{array}{c|cccccc} x_1 & 1 & 2 & 3 & 4 & 5 & 6 \\ \hline x_2 & 1 & 2 & 3 & 4 & 5 & 6 \end{array}$$

$$a) f_{x_1, x_2} = \begin{bmatrix} 1 & 2 & 3 & 4 & 5 & 6 \\ \hline 1 & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} \\ 2 & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} \\ 3 & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} \\ 4 & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} \\ 5 & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} \\ 6 & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} & \frac{1}{36} \end{bmatrix} \frac{1}{6}$$

$$\mathbb{E}[X_1] = [1 \ 2 \ 3 \ 4 \ 5 \ 6] f_{x_1, x_2} \underbrace{[1 \ 1 \ 1 \ 1 \ 1 \ 1]^T}_{f_{x_1}} = 3.5 = \mu_{x_1}$$

$$\mathbb{E}[X_1 - \mathbb{E}[X_1]] = (X_1 - \mu_{x_1}) f_{x_1, x_2} [1 \ 1 \ 1 \ 1 \ 1 \ 1] = 0$$

$$\mathbb{E}[X_1^2] = X_1 f_{x_1, x_2} [1 \ 1 \ 1 \ 1 \ 1 \ 1] = 15.1667$$

$$\mathbb{E}[(X_1 - \mathbb{E}[X_1])^2] = (X_1 - \mu_{x_1})^2 f_{x_1, x_2} [1 \ 1 \ 1 \ 1 \ 1 \ 1] = 2.9167$$

$$\mathbb{E}[(X_1 - \mathbb{E}[X_1])(X_2 - \mathbb{E}[X_2])^T] = (X_1 - \mu_{x_1})(X_2 - \mu_{x_2})^T f_{x_1, x_2} = \text{In MATLAB}$$

$$b) P_{x_1, x_2} = \mathbb{E}[(X_1 - \mathbb{E}[X_1])(X_2 - \mathbb{E}[X_2])^T] = (X_1 - \mu_{x_1})(X_2 - \mu_{x_2})^T f_{x_1, x_2} = \text{In MATLAB}$$

$$c) v_1 = x_1 \quad f_{v_1} = f_{x_1} \quad v_2 = x_1 + x_2 \quad f_{v_2} = \text{conv}(x_1, x_2)$$

$$f_{v_1, v_2} = f_{x_1} f_{x_2}^T = \text{In MATLAB}$$

$$d) \mathbb{E}[v_1] = \mathbb{E}[x_1] = 3.5 \quad \mathbb{E}[v_1 - \mathbb{E}[v_1]]^2 = \mathbb{E}[v_1^2] - \mu_{v_1}^2 = 15.1667$$

$$\sqrt{\mathbb{E}[v_1^2]} = \sqrt{\sum_{i=1}^6 v_{1i} f_{v_1}(v_{1i})} = 3.8944$$

$$e) \mathbb{E}[v_2] = \mathbb{E}[x_1 + x_2] = \mathbb{E}[x_1] + \mathbb{E}[x_2] = 7$$

$$\mathbb{E}[(v_2 - \mathbb{E}[v_2])^2] = (v_2 - \mathbb{E}[v_2])^2 f_{v_2} = 5.83$$

$$\sqrt{\mathbb{E}[v_2^2]} = \sqrt{\sum_{i=1}^6 v_{2i} f_{v_2}(v_{2i})} = 7.4$$

$$f) P_{v_1, v_2} = \mathbb{E}[(v_1 - \mathbb{E}[v_1])(v_2 - \mathbb{E}[v_2])] = \text{In MATLAB}$$

3) Independent: $E[g(x_1)h(x_2)] = E[g(x_1)]E[h(x_2)] \Leftrightarrow g(\cdot), h(\cdot)$

a) $g(x_1) = x_1, h(x_2) = x_2 \Rightarrow E[g(x_1)h(x_2)] = E[x_1x_2] = E[x_1]E[x_2]$

$$\begin{aligned} E[(x_1 - \bar{x}_1)(x_2 - \bar{x}_2)] &= E[x_1x_2 - x_1\bar{x}_2 - \bar{x}_1x_2 + \bar{x}_1\bar{x}_2] = E[x_1x_2] - \bar{x}_2 E[\frac{x_1}{x_1}] - \bar{x}_1 E[\frac{x_2}{x_2}] + \bar{x}_1\bar{x}_2 \\ &= E[\cancel{x_1x_2}] - \bar{x}_1\bar{x}_2 \quad \text{due to independence } E[x_1x_2] = E[x_1]E[x_2] \end{aligned}$$

$$E[(x_1 - \bar{x}_1)(x_2 - \bar{x}_2)] = 0 \quad \text{Uncorrelated}$$

b) Uncorrelated: $E[(x_1 - \bar{x}_1)(x_2 - \bar{x}_2)] = 0$

$$f_{x_1, x_2}(x_1, x_2) = ((2\pi)^{\frac{1}{2}} |P|^{\frac{1}{2}})^{-1} \exp\left(-\frac{1}{2} x^T P^{-1} x\right) = (2\pi\sigma_1\sigma_2(1-\rho_{12}^2))^{\frac{1}{2}} \exp\left(-\frac{1}{2}\left(\frac{x_1^2}{\sigma_1^2} - 2\rho_{12}\frac{x_1x_2}{\sigma_1\sigma_2} - \frac{x_2^2}{\sigma_2^2}\right)\right)$$

$$\rho_{12} = 0 \Rightarrow (2\pi\sigma_1\sigma_2)^{\frac{1}{2}} \exp\left(-\frac{1}{2}\left(\frac{x_1^2}{\sigma_1^2} + \frac{x_2^2}{\sigma_2^2}\right)\right) = (2\sigma_1)\exp\left(-\frac{1}{2}\frac{x_1^2}{\sigma_1^2}\right)(2\sigma_2)\exp\left(-\frac{1}{2}\frac{x_2^2}{\sigma_2^2}\right)$$

$$= f_{x_1} \cdot f_{x_2} \Rightarrow \text{Independence}$$

4) b) $V_o(k) = \{-2.5, -1.5, -0.5, 0.5, 1.5, 2.5\}$

$$E[V_o(k)] = \frac{1}{6} \sum_{k=1}^6 V_o(k) = 0 = \mu_0 \quad E[(V_o(k) - \bar{V}_o)^2] = \frac{1}{6-1} \sum_{k=1}^6 (V_o(k) - \bar{V}_o)^2 = 3.5 = \sigma_o^2$$

$$\sigma_o = 1.87$$

c) $V_o(k+1) = (1-\kappa)V_o(k) + \kappa V_o(k)$

$$E[V_o(k+1)] = E[(1-\kappa)V_o(k) + \kappa V_o(k)] = (1-\kappa)E[V_o(k)] + \kappa\mu_0$$

$$\mu_o(k+1) = (1-\kappa)\mu_o(k) + \kappa\mu_0 \quad @ ss \quad \mu_o(k+1) = \mu_o(k)$$

$$\mu_o(k) = (1-\kappa)\mu_o(k) + \kappa\mu_0 \quad \mu_o(k)(1-(1-\kappa)) = \kappa\mu_0 \quad \mu_o(k) = \frac{\kappa\mu_0}{1-\kappa}$$

$$\boxed{\mu_o(k) = \mu_0 = 0} \quad @ ss$$

$$E[(V_o(k+1) - \mu_o(k+1))^2] = E[(1-\kappa)V_o(k) + \kappa V_o(k) - (1-\kappa)\mu_o(k) - \kappa\mu_0]^2$$

$$E[(1-\kappa)^2 V_o(k)^2 + (1-\kappa)V_o(k)\cancel{\kappa V_o(k)} - (1-\kappa)^2 V_o(k)\mu_o(k) - (1-\kappa)\kappa V_o(k)\mu_o + (1-\kappa)V_o(k)\cancel{\kappa V_o(k)}$$

$$+ \kappa^2 V_o(k)^2 - (1-\kappa)\kappa V_o(k)V_o(k) - \kappa^2 V_o(k)\mu_0 - (1-\kappa)^2 V_o(k)\mu_o - (1-\kappa)\kappa V_o(k)\mu_o + (1-\kappa)^2 \mu_o(k)^2$$

$$+ (1-\kappa)\kappa V_o(k)\mu_0 - (1-\kappa)\kappa V_o(k)\mu_0 - \kappa^2 V_o(k)\mu_0 + (1-\kappa)\kappa V_o(k)\mu_0 + \kappa^2 \mu_0^2]$$

$$\sigma_o(k+1)^2 = (1-\kappa)^2 (\sigma_o(k)^2 - \mu_o(k)^2) + \kappa^2 \sigma_o^2 = \sigma_o(k)^2 (1-\kappa)^2 \sigma_o^2 - (1-\kappa)^2 \mu_o(k)^2 - \kappa^2 \sigma_o^2$$

$$\sigma_o(k)^2 (1 - 1 + 2\kappa - \kappa^2) = \kappa^2 \sigma_o^2$$

$$\boxed{\sigma_o(k) = \frac{\kappa \sigma_o}{\sqrt{1-\kappa^2}}}$$

$$\begin{aligned}
d) \quad R(h) &= E[V_n(h)V_n(h-L)] = E[(1-\epsilon)V_n(h-1) + \epsilon V_o(h-1))(1-\epsilon)V_n(h-1-L) + \epsilon V_o(h-1-L))] \\
&= E[(1-\epsilon)^2 V_n(h-1)V_n(h-1-L) + (1-\epsilon)\epsilon V_n(h-1)V_o(h-1-L) + (1-\epsilon)\epsilon V_o(h-1)V_n(h-1-L)] \\
&\quad + (1-\epsilon)^2 E[V_n(h-1)V_n(h-1-L)] + (1-\epsilon)\epsilon E[V_n(h-1)] \cancel{\mu_0} + (1-\epsilon)\epsilon \cancel{\mu_0} E[V_o(h-1-L)] + \epsilon^2 \sigma_o^2 \\
&= (1-\epsilon)^2 E[V_n(h-1)V_n(h-1-L)] + \epsilon^2 \sigma^2
\end{aligned}$$

$$R(h) = E[V_n(h)V_n(h-L)] \Rightarrow \boxed{\sigma_n^2(h-L) = (1-\epsilon)^2 \sigma_n^2(h-L) + \epsilon^2 \sigma^2}$$

e) $0 < \epsilon < 1$

5) $f_x(x) = \begin{cases} 0 & x < 0 \\ \frac{x}{2} & 0 \leq x \leq 2 \\ 0 & x > 2 \end{cases}$

$$E[x] = \int_0^2 x \frac{x}{2} dx = \frac{x^3}{6} \Big|_0^2 = \left(\frac{2^3}{6} - 0\right) = \frac{8}{6} = \boxed{\frac{4}{3}} \quad (a)$$

$$E[(x-\bar{x})^2] = E[x^2 - 2x\bar{x} + \bar{x}^2] = E[x^2] - 2\bar{x}E[x] + \bar{x}^2 = E[x^2] - \bar{x}^2$$

$$E[x^2] = \int_0^2 x^2 \frac{x}{2} dx = \frac{x^4}{8} \Big|_0^2 = \frac{16}{8} - 0 = 2 \quad (b)$$

$$E[(x-\bar{x})^2] = E[x^2] - \bar{x}^2 = 2 - \left(\frac{4}{3}\right)^2 = 2 - \frac{16}{9} = \boxed{1.1111}$$

6) $P_x = \begin{bmatrix} 2 & 1 \\ 1 & 4 \end{bmatrix} \quad x \sim N(0, P_x)$

a) $\det(A - \lambda I) = 0 \rightarrow \lambda = 4.4142, 1.5858$

b) $x^T P_x^{-1} x = c^2 \quad \text{PCA} = \frac{1}{\sqrt{\det(P_x)}} v_i$

c) In MATLAB

d) $f_x(x) = \frac{1}{(2\pi)^{n/2} |P_x|^{1/2}} \exp\left(-\frac{1}{2} x^T P_x^{-1} x\right) \quad f_x(c) = \left((2\pi)^{\frac{n}{2}} |P_x|^{1/2}\right)^{-1} \exp\left(-\frac{1}{2} c^2\right)$

$$\boxed{f_x(0.25) = 0.0503 \quad f_x(1) = 0.0365 \quad f_x(1.5) = 0.0195}$$

$$7) \quad x \sim N(0, \sigma_x^2) \quad y = 2x^2$$

$$a) \quad \mu_y = E[y] = E[2x^2] = 2E[x^2] = 2\sigma_x^2$$

$$\sigma_y^2 = E[(y - E[y])^2] = E[y^2 - 2\mu_y y + \mu_y^2] = E[y^2] - \mu_y^2 = E[(x^2)^2] - \mu_y^2$$

$$\sigma_y^2 = 4E[x^4] - \mu_y^2$$

$$M_x(s) = e^{\sigma_x^2 s^2} \quad M_x^{(4)} = E[x^4] \quad M_x' = \sigma_x^2 M_x \quad M_x'' = \sigma_x^4 M_x + \sigma_x^2 M_x' = \sigma_x^4 M_x + \sigma_x^2 s^2 M_x$$

$$M_x''' = M_x(\sigma_x^2 + \sigma_x^4 s^2) \quad M_x^{(4)} = 2\sigma_x^2 M_x + (\sigma_x^2 + \sigma_x^4 s^2) M_x' = 2\sigma_x^2 M_x + (\sigma_x^2 + \sigma_x^4 s^2) M_x$$

$$M_x^{(4)} = M_x(\sigma_x^4 + 3\sigma_x^4 s^2) \quad M_x^{(4)} = M_x(\sigma_x^4 + 3\sigma_x^4 s^2) + M_x'(\sigma_x^2 + \sigma_x^4 s^2) = M_x(\sigma_x^4 + 3\sigma_x^4 s^2 + 3\sigma_x^4 s^4 + \sigma_x^4 s^4)$$

$$M_x^{(4)} = M_x(\sigma_x^4 + 6\sigma_x^4 s^2 + \sigma_x^4 s^4) \quad M^{(4)}(0) = 3\sigma_x^4$$

$$\boxed{\sigma_y^2 = 3\sigma_x^4 - 2\sigma_x^2}$$

c) The mean of y is equal to $2\sigma_x^2$. The standard deviation of y is greatly expanded.

d) y is a Gaussian Random Variable.

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QUESTION 1

```
clear; close all; clc;

N = 6;
a = (1:6)';
b = (4:9)';
c = [1 1 3 3 3 5]';
d1 = a;
d2 = c;
d = [d1 d2];

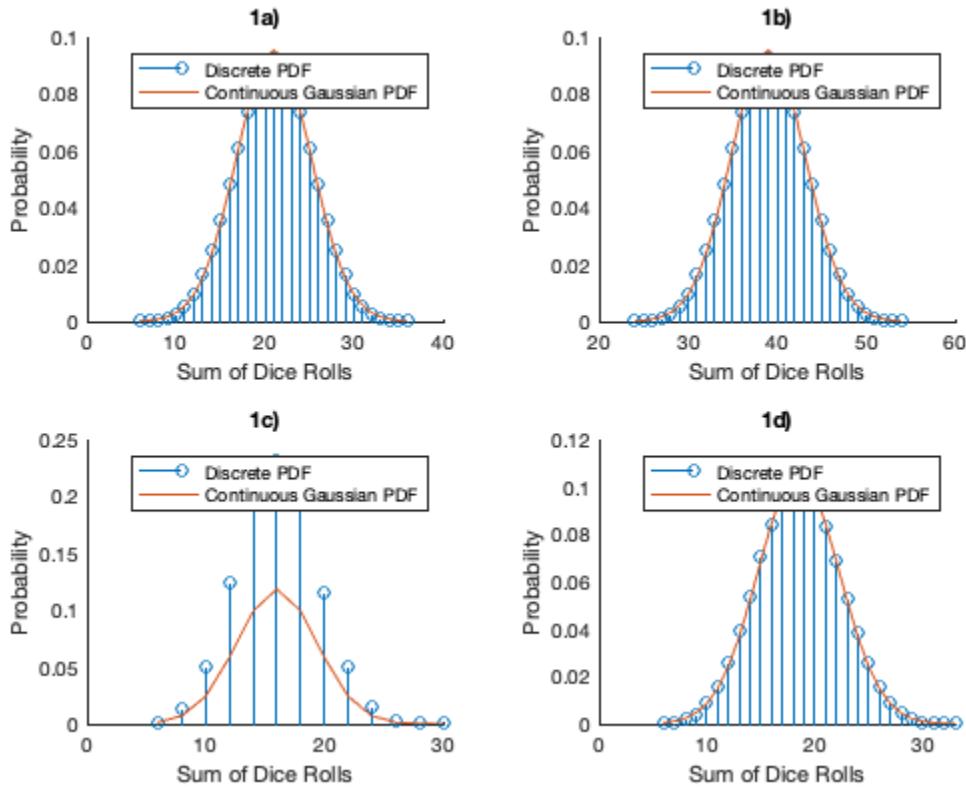
% a)
[pmf_a, shift_a, mu_a, sigma_a] = simDice(a, N);
fprintf('1a) Mean & Standard Deviation: [%0.3g %0.3g]\n', mu_a, sigma_a);
fprintf('1a) Sum of the PDF: %0.3g\n', sum(pmf_a));
% b)
[pmf_b, shift_b, mu_b, sigma_b] = simDice(b, N);
fprintf('1b) Mean & Standard Deviation: [%0.3g %0.3g]\n', mu_b, sigma_b);
fprintf('1b) Sum of the PDF: %0.3g\n', sum(pmf_b));
% c)
[pmf_c, shift_c, mu_c, sigma_c] = simDice(c, N);
fprintf('1c) Mean & Standard Deviation: [%0.3g %0.3g]\n', mu_c, sigma_c);
fprintf('1c) Sum of the PDF: %0.3g\n', sum(pmf_c));
% d)
[pmf_d, shift_d, mu_d, sigma_d] = sim2Dice(d1, d2, N/2);
fprintf('1d) Mean & Standard Deviation: [%0.3g %0.3g]\n', mu_d, sigma_d);
fprintf('1d) Sum of the PDF: %0.3g\n\n', sum(pmf_d));

figure();
hold('on');
title('Discrete PDFs vs. Continuous Gaussian PDF');
tiledlayout(2,2);
nexttile();
hold('on');
title('1a)');
stem(shift_a, pmf_a);
plot(shift_a, normpdf(shift_a, mu_a, sigma_a));
xlabel('Sum of Dice Rolls');
ylabel('Probability');
legend('Discrete PDF', 'Continuous Gaussian PDF');
```

```
nexttile();
hold('on');
title('1b)');
stem(shift_b, pmf_b);
plot(shift_b, normpdf(shift_b, mu_b, sigma_b));
xlabel('Sum of Dice Rolls');
ylabel('Probability');
legend('Discrete PDF', 'Continuous Gaussian PDF');

nexttile();
hold('on');
title('1c)');
stem(shift_c, pmf_c);
plot(shift_c, normpdf(shift_c, mu_c, sigma_c));
xlabel('Sum of Dice Rolls');
ylabel('Probability');
legend('Discrete PDF', 'Continuous Gaussian PDF');

nexttile();
hold('on');
title('1d)');
stem(shift_d, pmf_d);
plot(shift_d, normpdf(shift_d, mu_d, sigma_d));
xlabel('Sum of Dice Rolls');
ylabel('Probability');
legend('Discrete PDF', 'Continuous Gaussian PDF');
```



QUESTION 2

```

clear;

% a)
x1 = 1:6;
x2 = 1:6;
fx1 = simDice(x1', 1)';
fx2 = simDice(x2', 1)';
fx1x2 = fx1*fx2';
fprintf('2a) fx1x2:\n');
fprintf('\t %0.3g %0.3g %0.3g %0.3g %0.3g %0.3g\n', fx1x2);
Ex1 = x1*fx1x2*ones(1,length(x1))'; % E[x1]
fprintf('\n2a) E[x1]: %0.3g\n', Ex1);
Ex2 = x2*fx1x2*ones(1,length(x2))'; % E[x2]
fprintf('2a) E[x2]: %0.3g\n', Ex2);
Ex1_Ex1 = round((x1 - Ex1)*fx1); % E[(x1-E[x1])]
fprintf('2a) E[(x1 - E[x1])]: %0.3g\n', Ex1_Ex1);
Ex12 = (x1.^2)*fx1; % E[x1^2]
fprintf('2a) E[x1^2]: %0.3g\n', Ex12);
Px1 = ((x1 - Ex1).^2)*fx1; % E[(x1 - E[x1])^2]
fprintf('2a) E[(x1 - E[x1])^2]: %0.3g\n', Px1);
Px1x2 = ((x1 - Ex1)*(x2 - Ex2))*fx1x2; % E[(x1 - E[x1])(x2 - E[x2])]
fprintf('2a) Px1x2:\n')
fprintf('\t %0.3g %0.3g %0.3g %0.3g %0.3g %0.3g\n', Px1x2');

```

```

% b)
Px1x2 = ((x1 - Ex1)*(x2 - Ex2)')*fx1x2;
fprintf ('\n2b) Px1x2:\n')
fprintf ('\t %0.3g %0.3g %0.3g %0.3g %0.3g\n', Px1x2');

% c)
v1 = x1;
fv1 = fx1;
[fv2, v2] = simDice(x1', 2);
fv2 = fv2';
fv1v2 = fv1*fv2';
fprintf ('\n2c) fv1v2:\n');
fprintf(['\t %0.3g %0.3g %0.3g %0.3g %0.3g %0.3g ' ...
    '%0.3g %0.3g %0.3g\n'], fv1v2');

% d)
Ev1 = v1*fv1;
fprintf ('\n2d) E[v1]: %0.3g\n', Ev1);
v1RMS = sqrt((v1.^2)*fv1);
fprintf ('2d) RMS(v1): %0.3g\n', v1RMS);
Pv1 = ((v1 - Ev1).^2) * fv1;
fprintf ('2d) E[(v1 - E[v1])^2]: %0.3g\n', Pv1);

% e)
Ev2 = v2*fv2;
fprintf ('2e) E[v2]: %0.3g\n', Ev2);
v2RMS = sqrt((v2.^2)*fv2);
fprintf ('2e) RMS(v2): %0.3g\n', v2RMS);
Pv2 = ((v2 - Ev2).^2) * fv2;
fprintf ('2e) E[(v2 - E[v2])^2]: %0.3g\n', Pv2);

% f)
Pv1v2 = ((v1 - Ev1)*(v2(1:6) - Ev2)')*fv1*fv2';
fprintf ('2f) Pv1v2:\n');
fprintf(['\t %0.3g %0.3g %0.3g %0.3g %0.3g %0.3g %0.3g ' ...
    '%0.3g %0.3g %0.3g\n'], Pv1v2');

2a) fx1x2:
0.0278 0.0278 0.0278 0.0278 0.0278 0.0278
0.0278 0.0278 0.0278 0.0278 0.0278 0.0278
0.0278 0.0278 0.0278 0.0278 0.0278 0.0278
0.0278 0.0278 0.0278 0.0278 0.0278 0.0278
0.0278 0.0278 0.0278 0.0278 0.0278 0.0278
0.0278 0.0278 0.0278 0.0278 0.0278 0.0278

2a) E[x1]: 3.5
2a) E[x2]: 3.5
2a) E[(x1 - E[x1])]: 0
2a) E[x1^2]: 15.2
2a) E[(x1 - E[x1])^2]: 2.92
2a) Px1x2:
0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486

```

```

0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486

```

2b) P_{x1x2} :

```

0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486
0.486 0.486 0.486 0.486 0.486 0.486

```

2c) f_{v1v2} :

```

0.00463 0.00926 0.0139 0.0185 0.0231 0.0278 0.0231 0.0185 0.0139 0.00926
0.00463
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0.00463
0.00463 0.00926 0.0139 0.0185 0.0231 0.0278 0.0231 0.0185 0.0139 0.00926
0.00463

```

2d) $E[v1]$: 3.5

2d) $RMS(v1)$: 3.89

2d) $E[(v1 - E[v1])^2]$: 2.92

2e) $E[v2]$: 7

2e) $RMS(v2)$: 7.4

2e) $E[(v2 - E[v2])^2]$: 5.83

2f) P_{v1v2} :

```

0.081 0.162 0.243 0.324 0.405 0.486 0.405 0.324 0.243 0.162 0.081
0.081 0.162 0.243 0.324 0.405 0.486 0.405 0.324 0.243 0.162 0.081
0.081 0.162 0.243 0.324 0.405 0.486 0.405 0.324 0.243 0.162 0.081
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0.081 0.162 0.243 0.324 0.405 0.486 0.405 0.324 0.243 0.162 0.081

```

QUESTION 4

```
clear;
```

```

Vo = [-2.5 -1.5 -0.5 0.5 1.5 2.5];
fx = groupcounts(Vo') ./ length(Vo);

```

```

figure();
hold('on');
title('4a) Vo PDF')
stem(fx);
xlabel('Value');

```

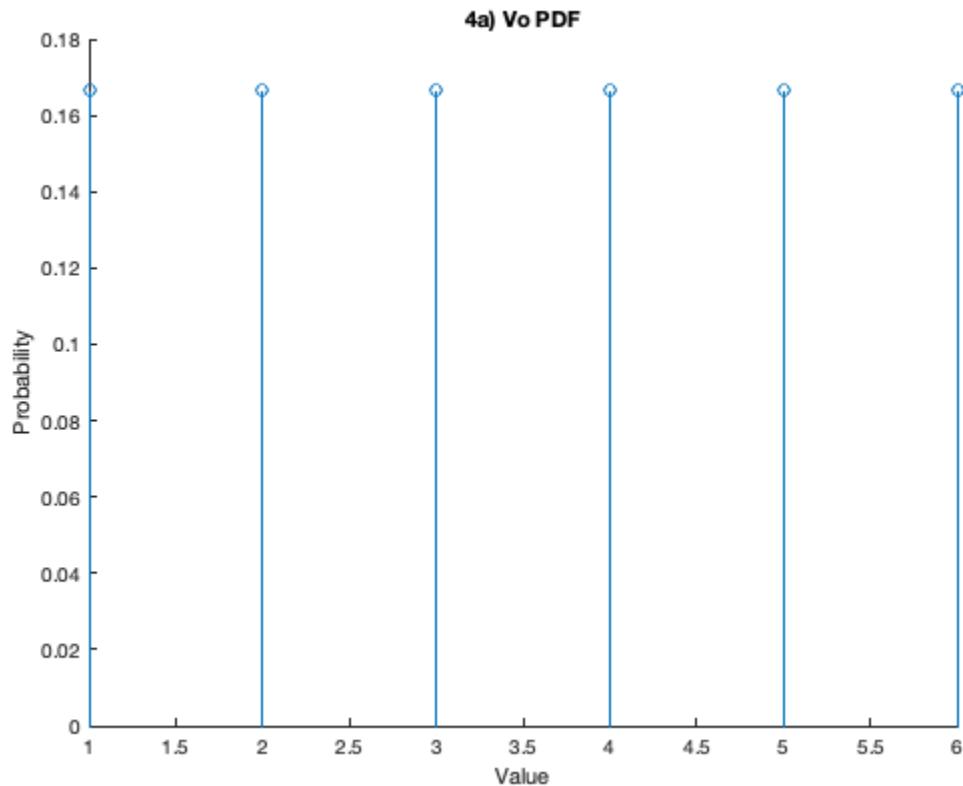
```

ylabel('Probability');

muVo = mean(Vo);
fprintf('\n4b) The mean of Vo: %0.3g\n', muVo);
sigmaVo = std(Vo);
fprintf('4b) The standard deviation of Vo: %0.3g\n', sigmaVo);
varVo = var(Vo);
fprintf('4b) The variance of Vo: %0.3g\n\n', varVo);

```

- 4b) The mean of Vo: 0
 4b) The standard deviation of Vo: 1.87
 4b) The variance of Vo: 3.5

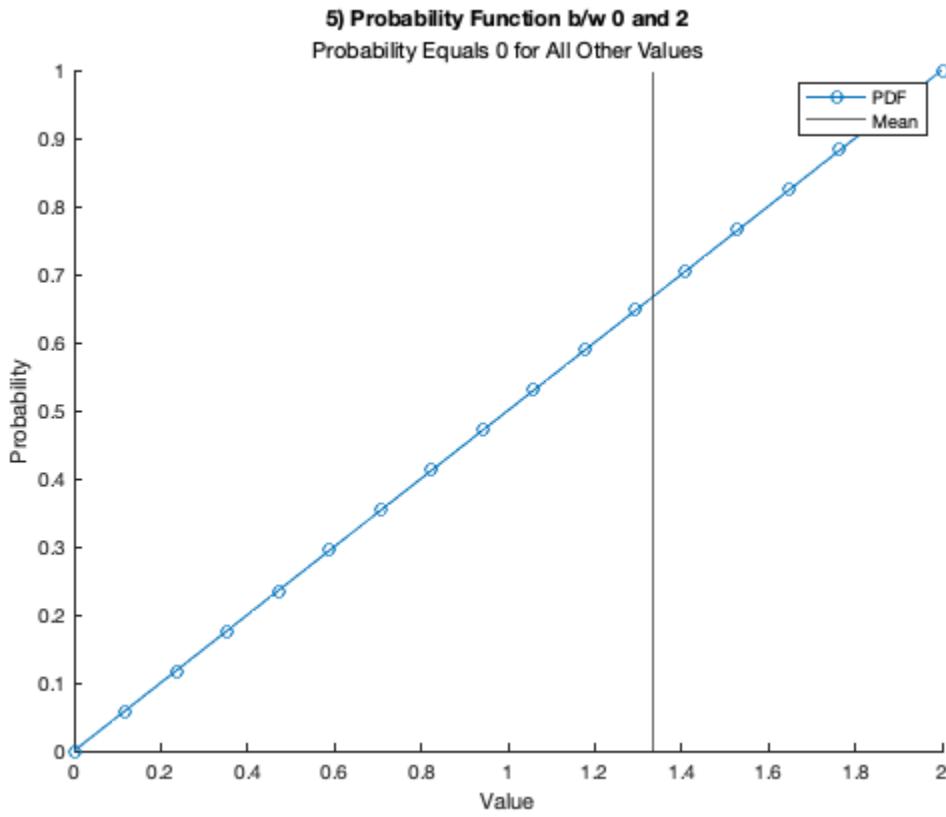


QUESTION 5

```

figure();
hold('on');
title('5) Probability Function b/w 0 and 2');
subtitle('Probability Equals 0 for All Other Values');
fplot(@(x) x/2, [0 2], '-o');
xline(4/3);
xlabel('Value');
ylabel('Probability');
legend('PDF', 'Mean');

```



QUESTION 6

```

clear;

Px = [ 2 1;
       1 4];
[V, D] = eigs(Px);
fprintf('6a) Eigenvalues of Px: [%0.3g %0.3g]\n', diag(D));

c = [0.25, 1, 1.5];
t = linspace(0, 2 * pi);
figure();
hold("on");
title("6c) Likelihood Ellipses for Varying Probabilities")
for k = 1:length(c)
    a = (V * sqrt(c(k)*D)) * [cos(t); sin(t)];
    plot(a(1, :), a(2, :));
end
xlabel("X");
ylabel("Y");
legend('c = 0.25', 'c = 1', 'c = 1.5');

fx = @(c) ((2*pi)^(size(Px,1)/2) * det(Px)^(1/2))^(-1) .* exp(-1/2.* (c.^2));
probs = fx(c);
fprintf('6d) The probability for c = 0.25: %0.3g\n', probs(1));

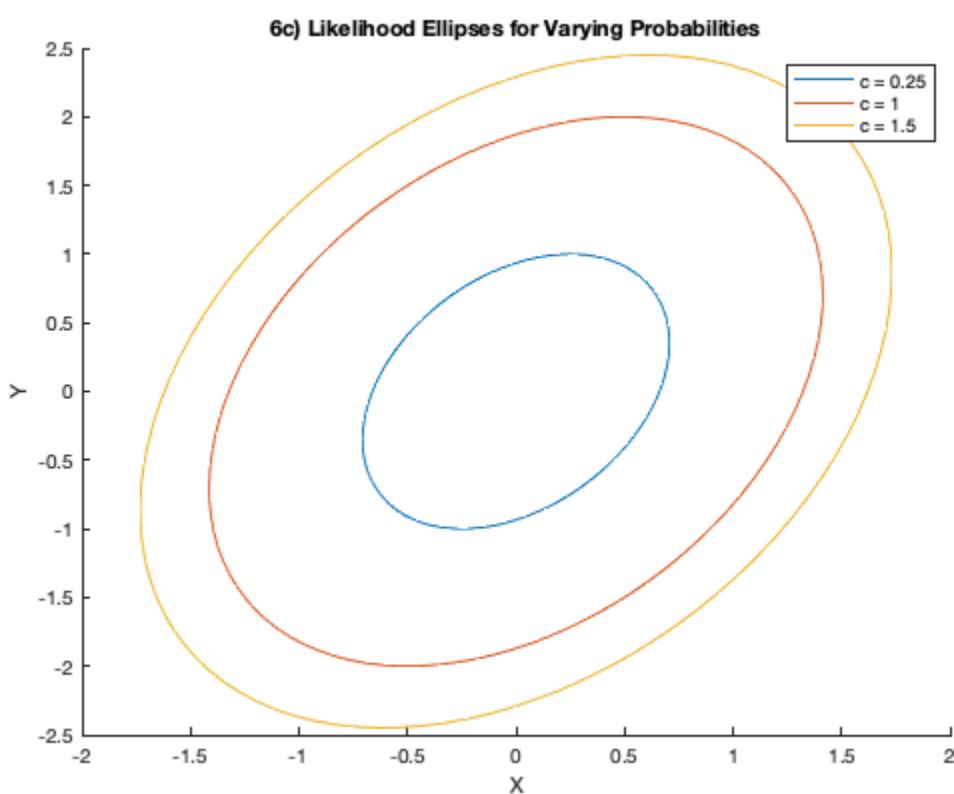
```

```

fprintf('6d) The probability for c = 1: %0.3g\n', probs(2));
fprintf('6d) The probability for c = 1.5: %0.3g\n\n', probs(3));

6a) Eigenvalues of Px: [4.41 1.59]
6d) The probability for c = 0.25: 0.0583
6d) The probability for c = 1: 0.0365
6d) The probability for c = 1.5: 0.0195

```



QUESTION 7

```

clear;
sigmax = 2.0;
varx = sigmax^2;
muy = 2*varx;
vary = 4*3*varx^2 - muy^2;
sigmay = sqrt(vary);

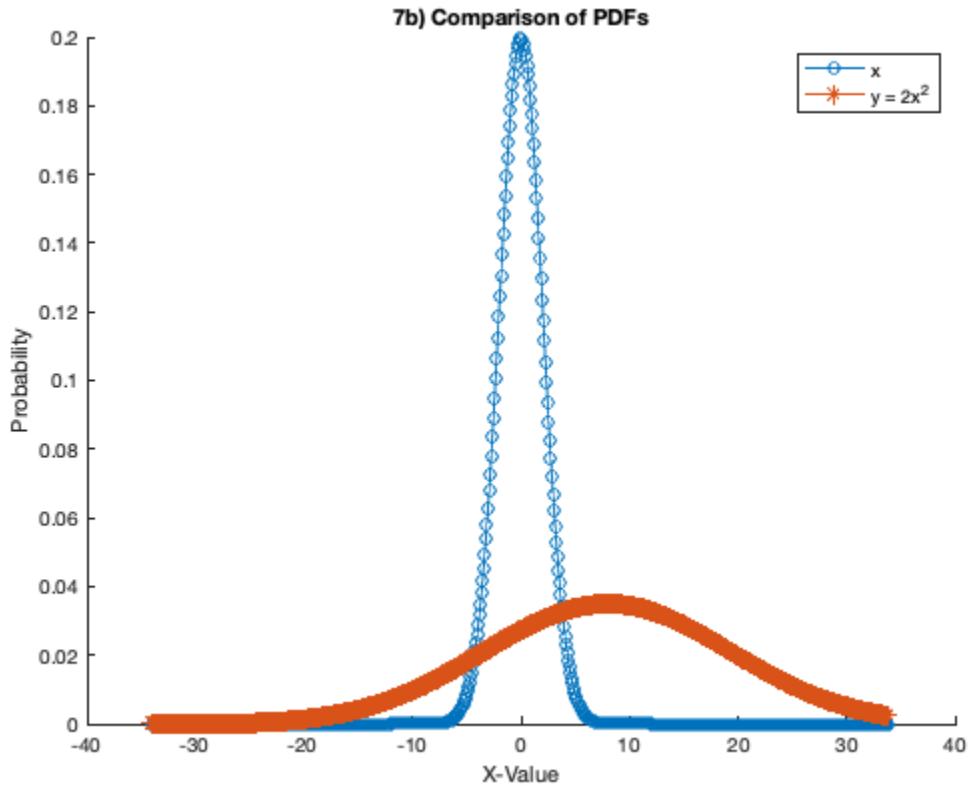
figure();
hold("on");
title('7b) Comparison of PDFs');
plot(-3*sigmay:0.1:3*sigmay, normpdf(-3*sigmay:0.1:3*sigmay, 0, sigmax), '-o');
plot(-3*sigmay:0.1:3*sigmay, normpdf(-3*sigmay:0.1:3*sigmay, muy, sigmay), '-*');
xlabel('X-Value');

```

```

ylabel('Probability');
legend('x', 'y = 2x^2');

```



FUNCTIONS

```

function [pmf, shift, mu, sigma] = simDice(die, N)
    pmf = genPMF(die, N);
    shift = linspace(min(die*N), max(die*N), length(pmf));
    mu = sum(shift.*pmf);
    sigma = sqrt(sum(((shift - mu).^2).*pmf));
end

function [pmf, shift, mu, sigma] = sim2Dice(d1, d2, N)
    pmf = genPMF(d1, N);
    [~, fx] = genPMF(d2, N);
    for i = 1:N
        pmf = conv(pmf, fx);
    end
    pmf(pmf == 0) = [];
    maxs = d1*N + d2*N;
    shift = linspace(min(d1*N + d2*N), max(d1*N + d2*N), length(pmf));
    mu = sum(shift.*pmf);
    sigma = sqrt(sum(((shift - mu).^2).*pmf));
end

function [pmf, fx] = genPMF(die, N)

```

```
[probs, vals] = groupcounts(die);
fx = zeros(length(die),1);
fx(vals) = probs./length(die);
pmf = fx';
for i = 1:N-1
    pmf = conv(pmf, fx');
end
pmf(pmf == 0) = [];
end

1a) Mean & Standard Deviation: [21 4.18]
1a) Sum of the PDF: 1
1b) Mean & Standard Deviation: [39 4.18]
1b) Sum of the PDF: 1
1c) Mean & Standard Deviation: [16 3.37]
1c) Sum of the PDF: 1
1d) Mean & Standard Deviation: [18.5 3.8]
1d) Sum of the PDF: 1
```

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