

一、解：属单因素方差分析问题： $r=4$, $n_1=n_2=n_3=n_4=5$, $n=20$.

对于 $A_1 \rightarrow T_{1.} = \sum_{j=1}^5 X_{1j} = 5.2 + 6.3 + 4.9 + 6.8 + 3.2 = 26.4$

$T_{1.}^2 = (\sum_{j=1}^5 X_{1j})^2 = (26.4)^2 = 696.96$

$\sum_{j=1}^5 X_{1j}^2 = 5.2^2 + 6.3^2 + 4.9^2 + 6.8^2 + 3.2^2 = 147.22$

对于 $A_2 \rightarrow T_{2.} = \sum_{j=1}^5 X_{2j} = 7.4 + 8.3 + 5.9 + 4.9 + 6.5 = 33$

$T_{2.}^2 = (\sum_{j=1}^5 X_{2j})^2 = (33)^2 = 1089$

$\sum_{j=1}^5 X_{2j}^2 = 7.4^2 + 8.3^2 + 5.9^2 + 4.9^2 + 6.5^2 = 224.72$

对于 $A_3 \rightarrow T_{3.} = \sum_{j=1}^5 X_{3j} = 12.3 + 9.4 + 7.8 + 8.5 + 10.8 = 48.8$

$T_{3.}^2 = (\sum_{j=1}^5 X_{3j})^2 = (48.8)^2 = 2381.44$

$\sum_{j=1}^5 X_{3j}^2 = 12.3^2 + 9.4^2 + 7.8^2 + 8.5^2 + 10.8^2 = 489.38$

	$T_{i.} = \sum_{j=1}^5 X_{ij}$	$T_{i.}^2$	$\sum_{j=1}^5 X_{ij}^2$
A_1	26.4	696.96	147.22
A_2	33	1089	224.72
A_3	48.8	2381.44	489.38
A_4	48.8	2381.44	489.38
Σ	$\sum_{i=1}^4 T_{i.} = 139.7$	$\sum_{i=1}^4 T_{i.}^2 = 5159.65$	$\sum_{i=1}^4 \sum_{j=1}^5 X_{ij}^2 = 1081.35$

$T_{..} = 26.4 + 33 + 48.8 = 108.2$

对于 $A_4 \rightarrow T_{4.} = \sum_{j=1}^5 X_{4j} = 3.9 + 6.4 + 7.9 + 4.1 + 9.2 = 31.5$, $T_{4.}^2 = 992.25$, $\sum_{j=1}^5 X_{4j}^2 = 220.03$

$T_{..} = \sum_{i=1}^4 T_{i.} = 26.4 + 33 + 31.5 + 48.8 = 139.7$

$T_{..}^2 = \sum_{i=1}^4 T_{i.}^2 = 696.96 + 1089 + 992.25 + 2381.44 = 5159.65$

$\sum_{i=1}^4 \sum_{j=1}^5 X_{ij}^2 = 147.22 + 224.72 + 220.03 + 489.38 = 1081.35$

$\sigma_T^2 = \sum_{i=1}^r \sum_{j=1}^{n_i} (X_{ij} - \bar{X})^2 = \sum_{i=1}^r \sum_{j=1}^{n_i} X_{ij}^2 - \frac{1}{n} (\sum_{i=1}^r \sum_{j=1}^{n_i} X_{ij})^2 = 1081.35 - \frac{1}{20} \times (139.7)^2 = 1081.35 - 975.8045 = 105.5455$

$\sigma_A^2 = \sum_{i=1}^r n_i (\bar{X}_{i.} - \bar{X})^2 = \frac{1}{n} \sum_{i=1}^r (\sum_{j=1}^{n_i} X_{ij})^2 - \frac{1}{n} (\sum_{i=1}^r \sum_{j=1}^{n_i} X_{ij})^2 = \frac{1}{5} \times 5159.65 - \frac{1}{20} \times (139.7)^2 = 1031.93 - 975.8045 = 56.1255$

$\sigma_E^2 = \sigma_T^2 - \sigma_A^2 = 105.5455 - 56.1255 = 49.42$

当 $\alpha=0.01$ 时方差分析如下：

方差来源	离差平方和	自由度	均方离差平方和	$F_{\text{值}}$	$F_{0.01}(3,16) = 5.29$
组间	$\sigma_A^2 = 56.1255$	$r-1=3$	$S_A^2 = \sigma_A^2/3 = 18.7085$	$S_A^2/S_E^2 = 6.0567$	
组内	$\sigma_E^2 = 49.42$	$n-r=16$	$S_E^2 = \sigma_E^2/16 = 3.08875$		
总和	$\sigma_T^2 = 105.5455$	$n-1=19$			

由于 $F = 6.0567 > F_{0.01}(3,16) = 5.29$ 故有显著差异



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二. 解: 本题属于两因素重复试验的方差分析问题, 由条件已知 $Y=3, S=3, t=2, n=18$.

$A \backslash B$	B_1	B_2	B_3	$T_{t..} = \sum_{j=1}^3 \sum_{k=1}^2 X_{tjk}$	$T_{t..}^2 = (\sum_{j=1}^3 \sum_{k=1}^2 X_{tjk})^2$	$\sum_{j=1}^3 \sum_{k=1}^2 X_{tjk}^2$
A_1	43.39 (82), 6724	37.29 (66), 4356	58.42 (100), 10000	248	61504	10708
A_2	47.53 (100), 10000	41.30 (71), 5041	46.60 (106), 11236	277	76729	13315
A_3	38.42 (80), 6400	48.47 (95), 9025	56.41 (97), 9409	272	73984	12538
$T_{.j} = \sum_{t=1}^3 \sum_{k=1}^2 X_{tjk}$	282	232	303	$\sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 X_{tjk} = 797$	$\sum_{t=1}^3 T_{t..}^2 = 212217$	$\sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 X_{tjk}^2 = 36561$
$T_{.j}^2 = (\sum_{t=1}^3 \sum_{k=1}^2 X_{tjk})^2$	68644	53824	91809	$\sum_{j=1}^3 T_{.j}^2 = 214277$		
$\sum_{t=1}^3 (\sum_{k=1}^2 X_{tjk})^2$	23124	18422	30645	$\sum_{j=1}^3 \sum_{t=1}^3 (\sum_{k=1}^2 X_{tjk})^2 = 12191$		

其中, $T_{.j} = \sum_{t=1}^3 \sum_{k=1}^2 X_{tjk}$, $T_{t..} = \sum_{j=1}^3 \sum_{k=1}^2 X_{tjk}$, $T = \sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 X_{tjk}$,

$$\sigma_T^2 = \sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 (X_{tjk} - \bar{X})^2 = \sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 X_{tjk}^2 - \frac{T^2}{n} = 36561 - \frac{1}{18} \times 797^2 = 1271.61$$

$$\sigma_A^2 = \sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 (\bar{X}_{t..} - \bar{X})^2 = st \sum_{t=1}^3 (\bar{X}_{t..} - \bar{X})^2 = \frac{1}{st} \sum_{t=1}^3 T_{t..}^2 - \frac{T^2}{n} = \frac{1}{6} \times 212217 - \frac{1}{18} \times 797^2 = 80.11$$

$$\sigma_B^2 = \sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 (\bar{X}_{.j} - \bar{X})^2 = rt \sum_{j=1}^3 (\bar{X}_{.j} - \bar{X})^2 = \frac{1}{rt} \sum_{j=1}^3 T_{.j}^2 - \frac{T^2}{n} = \frac{1}{6} \times 214277 - \frac{1}{18} \times 797^2 = 423.44$$

$$\sigma_E^2 = \sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 (X_{tjk} - \bar{X}_{tjk})^2 = \sum_{t=1}^3 \sum_{j=1}^3 \sum_{k=1}^2 X_{tjk}^2 - \frac{1}{t} \sum_{t=1}^3 \sum_{j=1}^3 (\sum_{k=1}^2 X_{tjk})^2 = 36561 - \frac{1}{3} \times 12191 = 465.5$$

$$\sigma_{AXB}^2 = \sigma_T^2 - \sigma_A^2 - \sigma_B^2 - \sigma_E^2 = 1271.61 - 80.11 - 423.44 - 465.5 = 302.56$$

方差来源	离差平方和	自由度	均方离差平方和	F值	显著性
因素A	$\sigma_A^2 = 80.11$	$r-1=2$	$S_A^2 = \sigma_A^2/2 = 40.055$	$S_A^2/S_E^2 = 0.774$	不显著
因素B	$\sigma_B^2 = 423.44$	$s-1=2$	$S_B^2 = \sigma_B^2/2 = 211.72$	$S_B^2/S_E^2 = 4.094$	不显著
交互作用AXB	$\sigma_{AXB}^2 = 302.56$	$(r-1)(s-1)=4$	$S_{AXB}^2 = \sigma_{AXB}^2/4 = 75.64$	$S_{AXB}^2/S_E^2 = 1.4625$	不显著
误差E	$\sigma_E^2 = 465.5$	$rst-1=9$	$S_E^2 = \sigma_E^2/9 = 51.72$		
总和	$\sigma_T^2 = 1271.61$	$rst-1=17$			

当 $\alpha=0.05$ 时, 查表得 $F_{0.05}(2,9)=4.26 > 0.774$, $F_{0.05}(2,9)=4.26 > 4.094$, $F_{0.05}(4,9)=3.63 > 1.4625$



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三. 解: (1)

可以看作是一元线性回归模型

(2)	x_i	y_i	x_i^2	y_i^2	$x_i y_i$
	1	1.3	1	1.69	1.3
	2	2.5	4	6.25	5
	3	3.7	9	13.69	11.1
	4	5.3	16	28.09	21.2
	5	6.4	25	40.96	32
	6	7.2	36	51.84	43.2
Σ	21	26.4	91	142.52	113.8

$$\bar{x} = \frac{1}{n} \sum_{i=1}^6 x_i = \frac{1}{6} \times 21 = 3.5$$

$$\bar{y} = \frac{1}{n} \sum_{i=1}^6 y_i = \frac{1}{6} \times 26.4 = 4.4$$

$$L_{xx} = \sum_{i=1}^6 (x_i - \bar{x})^2 = \sum_{i=1}^6 x_i^2 - n\bar{x}^2 = 91 - 6 \times (3.5)^2 = 17.5$$

$$L_{xy} = \sum_{i=1}^6 x_i y_i - 6\bar{x}\bar{y} = 113.8 - 6 \times 3.5 \times 4.4 = 21.4$$

$$\hat{\beta}_1 = \frac{L_{xy}}{L_{xx}} = \frac{21.4}{17.5} = 1.22, \quad \hat{\beta}_0 = \bar{y} - \hat{\beta}_1 \bar{x} = 4.4 - 1.22 \times 3.5 = 0.1$$

$$\therefore \hat{y} = \hat{\beta}_0 + \hat{\beta}_1 x = 0.1 + 1.22x$$

(3) 当 $x=10$ 时, y 的预测值 $\hat{y} = 0.1 + 1.22 \times 10 = 12.3$

$$L_{yy} = \sum_{i=1}^6 y_i^2 - 6\bar{y}^2 = 142.52 - 6 \times 4.4^2 = 26.36$$

$$Q_E^2 = L_{yy} - \hat{\beta}_1 L_{xy} = 26.36 - 1.22 \times 21.4 = 0.252, \quad \hat{\sigma}^2 = \frac{Q_E^2}{n-2} = \frac{0.252}{4} = 0.063$$

$$T = \frac{y - \hat{y}}{\sqrt{\hat{\sigma}^2 [1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{L_{xx}}]}} \sim t_{(n-2)} = t_{(4)}$$

给定置信度 $1-\alpha=0.95$ 时, 令 $P\{|T| \leq t_{0.025}(4)\} = 1-\alpha = 0.95$

$$\therefore t_{\frac{\alpha}{2}(n-2)} \sqrt{\hat{\sigma}^2 [1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{L_{xx}}]} = t_{0.025}(4) \sqrt{0.063 \times [1 + \frac{1}{6} + \frac{(10-3.5)^2}{17.5}]} = t_{0.025}(4) \times 0.47 = 2.7764 \times 0.47 = 1.305$$

$$\therefore \text{当 } x=10 \text{ 时, 置信度为 } 95\% \text{ 的 } y \text{ 的预测区间为: } [\hat{y} - t_{0.025}(4) \sqrt{\hat{\sigma}^2 [1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{L_{xx}}]}, \hat{y} + t_{0.025}(4) \sqrt{\hat{\sigma}^2 [1 + \frac{1}{n} + \frac{(x - \bar{x})^2}{L_{xx}}]}]$$

$$= [11.025, 13.635]$$



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四.解: (1) 构造函数 $X(t) = \cos(t - \Phi)$, $-\infty < t < \infty$

Φ	0	$\frac{\pi}{2}$
P	$\frac{1}{2}$	$\frac{1}{2}$

$$X(\omega) = \cos(t - \Phi) = \cos \Phi \cos t + \sin \Phi \sin t$$

$X(\omega)$ 的分布律为:

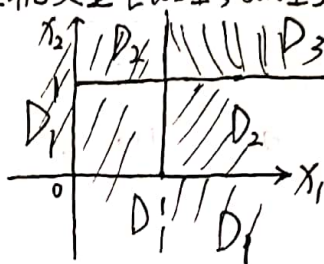
$X(\omega)$	1	0
P	$\frac{1}{2}$	$\frac{1}{2}$

当 $\Phi = 0$ 时, $X(\omega) = 1$, $P\{\Phi = 0\} = \frac{1}{2}$
 当 $\Phi = \frac{\pi}{2}$ 时, $X(\omega) = 0$, $P\{\Phi = \frac{\pi}{2}\} = \frac{1}{2}$

一维分布函数: $F(x, 0) = \begin{cases} 0 & , x < 0 \\ \frac{1}{2} & , 0 \leq x < 1 \\ 1 & , 1 \leq x \end{cases}$

$X(\omega) = \cos \Phi$, $X(\frac{\pi}{2}) = \sin \Phi$, 二维随机变量 $[\cos \Phi, \sin \Phi]$ 的分布律为:

$(\cos \Phi, \sin \Phi)$	(1, 0)	(0, 1)
P	$\frac{1}{2}$	$\frac{1}{2}$



$$F(x_1, x_2; 0, \frac{\pi}{2}) = P\{\cos \Phi \leq x_1, \sin \Phi \leq x_2\} = \begin{cases} 0, & (x_1, x_2) \in D_1 \\ \frac{1}{2}, & (x_1, x_2) \in D_2 \\ 1, & (x_1, x_2) \in D_3 \end{cases}$$

(2) $M_X(t) = E[X(t)] = E[\cos(t - \Phi)] = \frac{1}{2} \cos t + \frac{1}{2} \sin t$
 $= \frac{1}{2} (\sin t + \cos t)$

Φ	0	$\frac{\pi}{2}$
$X(t)$	$\cos t$	$\sin t$
P	$\frac{1}{2}$	$\frac{1}{2}$

$$R_X(t_1, t_2) = \frac{1}{2} \cos t_1 \cdot \cos t_2 + \frac{1}{2} \sin t_1 \cdot \sin t_2 = \frac{1}{2} \cos(t_1 - t_2)$$

注: $\cos(a \pm b) = \cos a \cos b \mp \sin a \sin b$

$\sin(a \pm b) = \sin a \cos b \pm \cos a \sin b$



五. 解: (1) $E(u) = \frac{b-a}{2} = \frac{1-0}{2} = \frac{1}{2}$, $D(u) = \frac{(b-a)^2}{12} = \frac{1}{12}$, $E(u^2) = D(u) + E(u)^2 = \frac{1}{12} + \frac{1}{4} = \frac{1}{3}$

$$m_{X(t)} = E[X(t)] = E[u + \cos(2\pi t + \varphi)] = E(u) + E[\cos(2\pi t + \varphi)] = \frac{1}{2} + \int_0^{2\pi} \frac{1}{2\pi} \cos(2\pi t + \varphi) d\varphi = \frac{1}{2}$$

$$R_{X(t_1, t_2)} = E[X(t_1) \cdot X(t_2)] = E[(u + \cos(2\pi t_1 + \varphi)) \cdot (u + \cos(2\pi t_2 + \varphi))]$$

$$= E[u^2 + u \cdot \cos(2\pi t_2 + \varphi) + u \cdot \cos(2\pi t_1 + \varphi) + \underbrace{\cos(2\pi t_1 + \varphi) \cdot \cos(2\pi t_2 + \varphi)}_{\text{trig identity}}]$$

$$= E(u^2) + E[u \cos(2\pi t_2 + \varphi)] + E[u \cos(2\pi t_1 + \varphi)] + E[\cos(2\pi t_1 + \varphi) \cos(2\pi t_2 + \varphi)]$$

$$= \frac{1}{3} + \frac{1}{2} \int_0^{2\pi} \frac{1}{2\pi} \cos(2\pi t_2 + \varphi) d\varphi + \frac{1}{2} \int_0^{2\pi} \frac{1}{2\pi} \cos(2\pi t_1 + \varphi) d\varphi + \frac{1}{2} \int_0^{2\pi} \frac{1}{2\pi} \cos(2\pi t_1 + \varphi) \cos(2\pi t_2 + \varphi) d\varphi$$

$$+ \frac{1}{4\pi} \int_0^{2\pi} \cos(2\pi t_1 - 2\pi t_2) d\varphi \quad \cos \alpha \cdot \cos \beta = \frac{1}{2} [\cos(\alpha + \beta) + \cos(\alpha - \beta)]$$

$$= \frac{1}{3} + \frac{2\pi}{4\pi} \cdot \cos(2\pi t_1 - 2\pi t_2) = \frac{1}{3} + \frac{1}{2} \cos 2\pi(t_1 - t_2)$$

$$\therefore R_X(\tau) = \frac{1}{3} + \frac{1}{2} \cos(2\pi\tau)$$

又: 数学期望是常数, 相关函数仅与时间 τ 有关

$\therefore X(t)$ 具有平稳性

$$-\lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T (1 - \frac{\tau}{T}) [R_X(\tau) - m_X^2] d\tau = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T (1 - \frac{\tau}{T}) [\frac{1}{3} + \frac{1}{2} \cos(2\pi\tau) - \frac{1}{4}] d\tau$$

$$= \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T (1 - \frac{\tau}{T}) [\frac{1}{12} + \frac{1}{2} \cos(2\pi\tau)] d\tau = \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T (1 - \frac{\tau}{T}) \times \frac{1}{12} d\tau + \lim_{T \rightarrow \infty} \frac{1}{T} \int_0^T (1 - \frac{\tau}{T}) \cdot \frac{1}{2} \cos(2\pi\tau) d\tau$$

$$= 0 + 0 = 0$$

$\therefore \langle X(t) \rangle = m_X \Rightarrow X(t)$ 具有数学期望的各态历经性.

$$Y(t) = \int_0^t X(t-\lambda) \cdot h(\lambda) d\lambda, \quad R_Y(t, t+\tau) = E[Y(t) \cdot Y(t+\tau)] = \int_0^t \int_0^{t+\tau} R_X(\alpha_2 - \alpha_1 - \tau) h(\alpha_1) h(\alpha_2) d\alpha_1 d\alpha_2$$

$\therefore Y(t)$ 具有平稳性.

$$(2) S_X(\omega) = \int_{-\infty}^{\infty} e^{-j\omega\tau} \cdot R_X(\tau) d\tau = \int_{-\infty}^{\infty} e^{-j\omega\tau} [\frac{1}{3} + \frac{1}{2} \cos(2\pi\tau)] d\tau$$

$$\therefore S_X(\omega) = \frac{1}{3} \times 2\pi \delta(\omega) + \frac{1}{2} \cdot \pi [\delta(\omega - 2\pi) + \delta(\omega + 2\pi)]$$

$$\text{又 频率响应函数为: } H(j\omega) = \frac{\alpha}{j\omega + \alpha}, \quad \alpha = \frac{1}{RC}$$

$$\therefore S_Y(\omega) = |H(j\omega)|^2 \cdot S_X(\omega) = \frac{\alpha^2}{\alpha^2 + \omega^2} \left[\frac{2\pi}{3} \delta(\omega) + \frac{\pi}{2} \delta(\omega - 2\pi) + \frac{\pi}{2} \delta(\omega + 2\pi) \right]$$

$$= \frac{2\pi}{3} \delta(\omega) + \frac{\alpha^2}{\alpha^2 + 4\pi^2} \left[\frac{\pi}{2} \delta(\omega - 2\pi) + \frac{\pi}{2} \delta(\omega + 2\pi) \right]$$

$$f(\omega) \cdot \delta(\omega) = f(\omega) \cdot \delta(\omega)$$

$$f(\omega) \cdot \delta(\omega - \omega_0) = f(\omega_0) \cdot \delta(\omega - \omega_0)$$



六.解: (1) 一步转移概率矩阵为:

$$P = \begin{bmatrix} 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ \frac{1}{3} & \frac{1}{3} & 0 & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \end{bmatrix}$$

二步转移概率矩阵为:

$$P_{(2)} = P^2 = \begin{bmatrix} 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ \frac{1}{3} & \frac{1}{3} & 0 & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \end{bmatrix} \cdot \begin{bmatrix} 0 & \frac{1}{3} & \frac{1}{3} & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \\ \frac{1}{3} & \frac{1}{3} & 0 & \frac{1}{3} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 \end{bmatrix} = \begin{bmatrix} \frac{4}{9} & \frac{1}{9} & \frac{1}{3} & \frac{1}{9} \\ \frac{1}{6} & \frac{1}{3} & \frac{1}{6} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{9} & \frac{4}{9} & \frac{1}{9} \\ \frac{1}{6} & \frac{1}{3} & \frac{1}{6} & \frac{1}{3} \end{bmatrix}$$

(2) 由于二步转移概率矩阵 $P_{(2)}$ 的元素全大于零, 所以此马氏链是遍历的. 又因为该马氏链是有限马氏链, 所以存在极限分布 $\pi_1, \pi_2, \pi_3, \pi_4$.

$$\begin{bmatrix} \frac{4}{9}-1 & \frac{1}{9} & \frac{1}{3} & \frac{1}{9} \\ \frac{1}{6} & \frac{1}{3}-1 & \frac{1}{6} & \frac{1}{3} \\ \frac{1}{3} & \frac{1}{9} & \frac{4}{9}-1 & \frac{1}{9} \\ \frac{1}{6} & \frac{1}{3} & \frac{1}{6} & \frac{1}{3}-1 \end{bmatrix} \cdot \begin{bmatrix} \pi_1 \\ \pi_2 \\ \pi_3 \\ \pi_4 \end{bmatrix} = 0 \Rightarrow \begin{cases} -\frac{5}{9}\pi_1 + \frac{1}{9}\pi_2 + \frac{1}{3}\pi_3 + \frac{1}{9}\pi_4 = 0 & ① \\ \frac{1}{6}\pi_1 + (-\frac{2}{3})\pi_2 + \frac{1}{6}\pi_3 + \frac{1}{3}\pi_4 = 0 & ② \\ \frac{1}{3}\pi_1 + \frac{1}{9}\pi_2 - \frac{5}{9}\pi_3 + \frac{1}{9}\pi_4 = 0 & ③ \\ \frac{1}{6}\pi_1 + \frac{1}{3}\pi_2 + \frac{1}{6}\pi_3 + (-\frac{2}{3})\pi_4 = 0 & ④ \end{cases} \Rightarrow \begin{cases} ③-① \Rightarrow \pi_1 = \pi_3 \\ ④-② \Rightarrow \pi_2 = \pi_4 \end{cases}$$

满足 $\pi_1 + \pi_2 + \pi_3 + \pi_4 = 1, \pi_j > 0 (j=1, 2, 3, 4)$ 的解为: $\pi_1 = \frac{3}{10} = \pi_3, \pi_2 = \pi_4 = \frac{1}{5}$.

$$(3) \text{ 绝对概率: } P\{X_2=1\} = P_1^{(2)} = \sum_{t=1}^4 P_t^{(0)} \cdot P_{t1}(2) = \sum_{t=1}^4 P\{X_0=t\} \cdot P_{t1}(2) = \frac{1}{4} \times (\frac{4}{9} + \frac{1}{6} + \frac{1}{3} + \frac{1}{6}) = \frac{5}{18}$$

$$\text{多维概率: } P\{X_1=1, X_3=2, X_5=3\} = \sum_{t=1}^4 P_t^{(0)} \cdot P_{t1}(1) \cdot P_{12}(3-1) \cdot P_{23}(5-3)$$

$$= \sum_{t=1}^4 P_t^{(0)} \cdot P_{t1}(1) \cdot P_{12}(2) \cdot P_{23}(2) = P_{20} \cdot P_{12}(2) \cdot P_{23}(2) \cdot \sum_{t=1}^4 P_{t1}(1)$$

$$= \frac{1}{4} \times \frac{1}{9} \times \frac{1}{6} \times [0 + \frac{1}{2} + \frac{1}{3} + \frac{1}{2}] = \frac{1}{4 \times 9 \times 6} \times \frac{4}{3} = \frac{1}{9 \times 6 \times 3} = \frac{1}{172}$$

证: $P\{X_{cn}=t\} = P_t^{(n)}$ $P_t^{(n)} = \sum_{t \in E} P_t^{(0)} \cdot P_{tj}(cn) \Rightarrow \text{绝对概率}$

$$P\{X_{cn_1}=t_1, X_{cn_2}=t_2, \dots, X_{cn_m}=t_m\} = \sum_{t \in E} P_t^{(0)} \cdot P_{t,t_1}(n_1) \cdot P_{t_1,t_2}(n_2-n_1) \cdots P_{t_{m-1},t_m}(n_m-n_{m-1})$$

$$= P_t^{(0)} \cdot P_{t,t_2}(n_2-n_1) \cdots P_{t_{m-1},t_m}(n_m-n_{m-1}) \cdot \sum_{t \in E} P_{t,t_1}(n_1) \Rightarrow \text{多维概率}$$

