

Probability Review for Final Exam  
 18.05 Spring 2014  
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**Problem 1.** We compute

$$E[X] = -2 \cdot \frac{1}{15} + -1 \cdot \frac{2}{15} + 0 \cdot \frac{3}{15} + 1 \cdot \frac{4}{15} + 2 \cdot \frac{5}{15} = \frac{2}{3}.$$

Thus

$$\text{Var}(X) = E\left(\left(X - \frac{2}{3}\right)^2\right) = \frac{14}{9}.$$

**Problem 2.** We first compute

$$E[X] = \int_0^1 x \cdot 2x dx = \frac{2}{3}$$

$$E[X^2] = \int_0^1 x^2 \cdot 2x dx = \frac{1}{2}$$

$$E[X^4] = \int_0^1 x^4 \cdot 2x dx = \frac{1}{3}.$$

Thus,

$$\text{Var}(X) = E[X^2] - (E[X])^2 = \frac{1}{2} - \frac{4}{9} = \frac{1}{18}$$

and

$$\text{Var}(X^2) = E[X^4] - (E[X^2])^2 = \frac{1}{3} - \frac{1}{4} = \frac{1}{12}.$$

**Problem 3.** Use  $\text{Var}(X) = E(X^2) - E(X)^2 \Rightarrow 3 = E(X^2) - 4 \Rightarrow \boxed{E(X^2) = 7}.$

**Problem 4.** answer:

Make a table

$X:$	0	1
prob:	$(1-p)$	$p$
$X^2$	0	1.

From the table,  $\boxed{E(X) = 0 \cdot (1-p) + 1 \cdot p = p}.$

Since  $X$  and  $X^2$  have the same table  $E(X^2) = E(X) = p.$

Therefore,  $\boxed{\text{Var}(X) = p - p^2 = p(1-p)}.$

**Problem 5.** Let  $X$  be the number of people who get their own hat.

Following the hint: let  $X_j$  represent whether person  $j$  gets their own hat. That is,

$X_j = 1$  if person  $j$  gets their hat and 0 if not.

We have,  $X = \sum_{j=1}^{100} X_j$ , so  $E(X) = \sum_{j=1}^{100} E(X_j)$ .

Since person  $j$  is equally likely to get any hat, we have  $P(X_j = 1) = 1/100$ . Thus,  $X_j \sim \text{Bernoulli}(1/100) \Rightarrow E(X_j) = 1/100 \Rightarrow \boxed{E(X) = 1}$ .

**Problem 6.** (a) There are a number of ways to present this.

$X \sim 3 \text{ binomial}(25, 1/6)$ , so

$$P(X = 3k) = \binom{25}{k} \left(\frac{1}{6}\right)^k \left(\frac{5}{6}\right)^{25-k}, \quad \text{for } k = 0, 1, 2, \dots, 25.$$

(b)  $X \sim 3 \text{ binomial}(25, 1/6)$ .

Recall that the mean and variance of  $\text{binomial}(n, p)$  are  $np$  and  $np(1-p)$ . So,

$$E(X) = 3 E(\text{textbinomial}(25, 1/6)) = 3 \cdot 25/6 = 75/6, \quad \text{and } \text{Var}(X) = 9 \text{Var}(\text{textbinomial}(25, 1/6)) = 9 \cdot 25 \cdot \frac{5}{6} \cdot \frac{1}{6} = 25 \cdot \frac{5}{2} = 62.5$$

(c)  $E(X + Y) = E(X) + E(Y) = 150/6 = 25$ ,  $E(2X) = 2E(X) = 150/6 = 25$ .

$$\text{Var}(X + Y) = \text{Var}(X) + \text{Var}(Y) = 250/4. \quad \text{Var}(2X) = 4\text{Var}(X) = 500/4.$$

The means of  $X + Y$  and  $2X$  are the same, but  $\text{Var}(2X) > \text{Var}(X + Y)$ .

This makes sense because in  $X + Y$  sometimes  $X$  and  $Y$  will be on opposite sides from the mean so distances to the mean will tend to cancel, However in  $2X$  the distance to the mean is always doubled.

**Problem 7.** First we find the value of  $a$ :

$$\int_0^1 f(x) dx = 1 = \int_0^1 x + ax^2 dx = \frac{1}{2} + \frac{a}{3} \Rightarrow a = 3/2.$$

The CDF is  $F_X(x) = P(X \leq x)$ . We break this into cases:

(i)  $b < 0 \Rightarrow F_X(b) = 0$ .

(ii)  $0 \leq b \leq 1 \Rightarrow F_X(b) = \int_0^b x + \frac{3}{2}x^2 dx = \frac{b^2}{2} + \frac{b^3}{2}$ .

(iii)  $1 < x \Rightarrow F_X(b) = 1$ .

Using  $F_X$  we get

$$P(.5 < X < 1) = F_X(1) - F_X(.5) = 1 - \left(\frac{.5^2 + .5^3}{2}\right) = \frac{13}{16}.$$

**Problem 8.** (i) yes, discrete, (ii) no, (iii) no, (iv) no, (v) yes, continuous

(vi) no (vii) yes, continuous, (viii) yes, continuous.

**Problem 9.** (a) We compute

$$P(X \geq 5) = 1 - P(X < 5) = 1 - \int_0^5 \lambda e^{-\lambda x} dx = 1 - (1 - e^{-5\lambda}) = e^{-5\lambda}.$$

(b) We want  $P(X \geq 15|X \geq 10)$ . First observe that  $P(X \geq 15, X \geq 10) = P(X \geq 15)$ . From similar computations in (a), we know

$$P(X \geq 15) = e^{-15\lambda} \qquad P(X \geq 10) = e^{-10\lambda}.$$

From the definition of conditional probability,

$$P(X \geq 15|X \geq 10) = \frac{P(X \geq 15, X \geq 10)}{P(X \geq 10)} = \frac{P(X \geq 15)}{P(X \geq 10)} = e^{-5\lambda}$$

**Note:** This is an illustration of the memorylessness property of the exponential distribution.

**Problem 10.**

(a) We did this in class. Let  $\phi(z)$  and  $\Phi(z)$  be the PDF and CDF of  $Z$ .

$$F_Y(y) = P(Y \leq y) = P(aZ + b \leq y) = P(Z \leq (y - b)/a) = \Phi((y - b)/a).$$

Differentiating:

$$f_Y(y) = \frac{d}{dy} F_Y(y) = \frac{d}{dy} \Phi((y - b)/a) = \frac{1}{a} \phi((y - b)/a) = \frac{1}{\sqrt{2\pi} a} e^{-(y-b)^2/2a^2}.$$

Since this is the density for  $N(b, a^2)$  we have shown  $Y \sim N(b, a^2)$ .

(b) By part (a),  $Y \sim N(\mu, \sigma^2) \Rightarrow Y = \sigma Z + \mu$ .

But, this implies  $(Y - \mu)/\sigma = Z \sim N(0, 1)$ . QED

**Problem 11.** (a)  $E(W) = 3E(X) - 2E(Y) + 1 = 6 - 10 + 1 = \boxed{-3}$

$$\text{Var}(W) = 9\text{Var}(X) + 4\text{Var}(Y) = 45 + 36 = \boxed{81}$$

(b) Since the sum of independent normal is normal part (a) shows:  $W \sim N(-3, 81)$ .

$$\text{Let } Z \sim N(0, 1). \text{ We standardize } W: P(W \leq 6) = P\left(\frac{W + 3}{9} \leq \frac{9}{9}\right) = P(Z \leq 1) \approx \boxed{.84}.$$

**Problem 12.**

**Method 1**

$U(a, b)$  has density  $f(x) = \frac{1}{b - a}$  on  $[a, b]$ . So,

$$E(X) = \int_a^b xf(x) dx = \frac{1}{b-a} \int_a^b x dx = \frac{x^2}{2(b-a)} \Big|_a^b = \frac{b^2 - a^2}{2(b-a)} = \boxed{\frac{a+b}{2}}.$$

$$E(X^2) = \int_a^b x^2 f(x) dx = \frac{1}{b-a} \int_a^b x^2 dx = \frac{x^3}{3(b-a)} \Big|_a^b = \frac{b^3 - a^3}{3(b-a)}.$$

Finding  $\text{Var}(X)$  now requires a little algebra,

$$\begin{aligned} \text{Var}(X) &= E(X^2) - E(X)^2 = \frac{b^3 - a^3}{3(b-a)} - \frac{(b+a)^2}{4} \\ &= \frac{4(b^3 - a^3) - 3(b-a)(b+a)^2}{12(b-a)} = \frac{b^3 - 3ab^2 + 3a^2b - a^3}{12(b-a)} = \frac{(b-a)^3}{12(b-a)} = \boxed{\frac{(b-a)^2}{12}}. \end{aligned}$$

## Method 2

There is an easier way to find  $E(X)$  and  $\text{Var}(X)$ .

Let  $U \sim U(a, b)$ . Then the calculations above show  $E(U) = 1/2$  and  $(E(U^2) = 1/3$   
 $\Rightarrow \text{Var}(U) = 1/3 - 1/4 = 1/12$ .

Now, we know  $X = (b-a)U + a$ , so  $E(X) = (b-a)E(U) + a = (b-a)/2 + a = (b+a)/2$   
and  $\text{Var}(X) = (b-a)^2 \text{Var}(U) = (b-a)^2/12$ .

## Problem 13.

(a)  $S_n \sim \text{Binomial}(n, p)$ , since it is the number of successes in  $n$  independent Bernoulli trials.

(b)  $T_m \sim \text{Binomial}(m, p)$ , since it is the number of successes in  $m$  independent Bernoulli trials.

(c)  $S_n + T_m \sim \text{Binomial}(n + m, p)$ , since it is the number of successes in  $n + m$  independent Bernoulli trials.

(d) Yes,  $S_n$  and  $T_m$  are independent. We haven't given a formal definition of independent random variables yet. But, we know it means that knowing  $S_n$  gives no information about  $T_m$ . This is clear since the first  $n$  trials are independent of the last  $m$ .

**Problem 14.** Compute the median for the exponential distribution with parameter  $\lambda$ . The density for this distribution is  $f(x) = \lambda e^{-\lambda x}$ . We know (or can compute) that the distribution function is  $F(a) = 1 - e^{-\lambda a}$ . The median is the value of  $a$  such that  $F(a) = .5$ . Thus,  $1 - e^{-\lambda a} = 0.5 \Rightarrow 0.5 = e^{-\lambda a} \Rightarrow \log(0.5) = -\lambda a \Rightarrow$   
 $a = \log(2)/\lambda$ .

**Problem 15.** (a) The joint distribution is given by

$Y \setminus X$	1	2	3	
1	$\frac{1168}{5383}$	$\frac{825}{5383}$	$\frac{305}{5383}$	$\frac{2298}{5383}$
2	$\frac{573}{5383}$	$\frac{1312}{5383}$	$\frac{1200}{5383}$	$\frac{3085}{5383}$
	$\frac{1741}{5383}$	$\frac{2137}{5383}$	$\frac{1505}{5383}$	1

with the marginal distribution of  $X$  at right and of  $Y$  at bottom.

(b)  $X$  and  $Y$  are dependent because, for example,

$$P(X = 1 \text{ and } Y = 1) = \frac{1168}{5383}$$

is not equal to

$$P(X = 1)P(Y = 1) = \frac{1741}{5383} \cdot \frac{2298}{5383}.$$

**Problem 16.** (a) Here we have two continuous random variables  $X$  and  $Y$  with joint probability density function

$$f(x, y) = \frac{12}{5}xy(1 + y) \quad \text{for } 0 \leq x \leq 1 \text{ and } 0 \leq y \leq 1,$$

and  $f(x, y) = 0$  otherwise. So

$$P\left(\frac{1}{4} \leq X \leq \frac{1}{2}, \frac{1}{3} \leq Y \leq \frac{2}{3}\right) = \int_{\frac{1}{4}}^{\frac{1}{2}} \int_{\frac{1}{3}}^{\frac{2}{3}} f(x, y) dy dx = \frac{41}{720}.$$

(b)  $F(a, b) = \int_0^a \int_0^b f(x, y) dy dx = \frac{3}{5}a^2b^2 + \frac{2}{5}a^2b^3$  for  $0 \leq a \leq 1$  and  $0 \leq b \leq 1$ .

(c) Since  $f(x, y) = 0$  for  $y > 1$ , we have

$$F_X(a) = \lim_{b \rightarrow \infty} F(a, b) = F(a, 1) = a^2.$$

(d) For  $0 \leq x \leq 1$ , we have

$$f_X(x) = \int_{-\infty}^{\infty} f(x, y) dy = \int_0^1 f(x, y) dy = 2x.$$

This is consistent with (c) because  $\frac{d}{dx}(x^2) = 2x$ .

(e) We first compute  $f_Y(y)$  for  $0 \leq y \leq 1$  as

$$f_Y(y) = \int_0^1 f(x, y) dx = \frac{6}{5}y(y + 1).$$

Since  $f(x, y) = f_X(x)f_Y(y)$ , we conclude that  $X$  and  $Y$  are independent.

**Problem 17.** (a) The marginal probability  $P_Y(1) = 1/2$

$$\Rightarrow P(X = 0, Y = 1) = P(X = 2, Y = 1) = 0.$$

Now each column has one empty entry. This can be computed by making the column add up to the given marginal probability.

$Y \setminus X$	0	1	2	$P_Y$
-1	1/6	1/6	1/6	1/2
1	0	1/2	0	1/2
$P_X$	1/6	2/3	1/6	1

(b) No,  $X$  and  $Y$  are not independent.

For example,  $P(X = 0, Y = 1) = 0 \neq 1/12 = P(X = 0) \cdot P(Y = 1)$ .

**Problem 18.** For shorthand, let  $P(X = a, Y = b) = p(a, b)$ .

(a)  $P(X = Y) = p(1, 1) + p(2, 2) + p(3, 3) + p(4, 4) = \boxed{34/136}$ .

(b)  $P(X + Y = 5) = p(1, 4) + p(2, 3) + p(3, 2) + p(4, 1) = \boxed{34/136}$ .

(c)  $P(1 < X \leq 3, 1 < Y \leq 3) = \text{sum of middle 4 probabilities in table} = \boxed{34/136}$ .

(d)  $\{1, 4\} \times \{1, 4\} = \{(1, 1), (1, 4), (4, 1), (4, 4)\} \Rightarrow \text{prob.} = \boxed{34/136}$ .

**Problem 19.** (a)  $X$  and  $Y$  are independent, so the table is computed from the product of the known marginal probabilities. Since they are independent,  $\text{Cov}(X, Y) = 0$ .

$Y \setminus X$	0	1	$P_Y$
0	1/8	1/8	1/4
1	1/4	1/4	1/2
2	1/8	1/8	1/4
$P_X$	1/2	1/2	1

(b) The sample space is  $\Omega = \{\text{HHH, HHT, HTH, HTT, THH, THT, TTH, TTT}\}$ .

$$P(X = 0, Z = 0) = P(\{\text{TTH, TTT}\}) = 1/4.$$

$$P(X = 0, Z = 1) = P(\{\text{TTH, THT}\}) = 1/4.$$

$$P(X = 0, Z = 2) = 0.$$

$$P(X = 1, Z = 0) = 0.$$

$$P(X = 1, Z = 1) = P(\{\text{HTH, HTT}\}) = 1/4.$$

$$P(X = 1, Z = 2) = P(\{\text{HHH, HHT}\}) = 1/4.$$

$Z \setminus X$	0	1	$P_Z$
0	1/4	0	1/4
1	1/4	1/4	1/2
2	0	1/4	1/4
$P_X$	1/2	1/2	1

$$\text{Cov}(X, Z) = E(XZ) - E(X)E(Z).$$

$$E(X) = 1/2, \quad E(Z) = 1, \quad E(XZ) = \sum x_i y_j p(x_i, y_j) = 3/4.$$

$$\Rightarrow \text{Cov}(X, Z) = 3/4 - 1/2 = \boxed{1/4}.$$

**Problem 20.** (a)  $F(a, b) = P(X \leq a, Y \leq b) = \int_0^a \int_0^b (x + y) dy dx.$

Inner integral:  $xy + \frac{y^2}{2} \Big|_0^b = xb + \frac{b^2}{2}$ . Outer integral:  $\frac{x^2}{2}b + \frac{b^2}{2}x \Big|_0^a = \frac{a^2b + ab^2}{2}$ .

So  $F(x, y) = \frac{x^2y + xy^2}{2}$  and  $F(1, 1) = 1$ .

(b)  $f_X(x) = \int_0^1 f(x, y) dy = \int_0^1 (x + y) dy = xy + \frac{y^2}{2} \Big|_0^1 = x + \frac{1}{2}$ .

By symmetry,  $f_Y(y) = y + 1/2$ .

(c) To see if they are independent we check if the joint density is the product of the marginal densities.

$f(x, y) = x + y$ ,  $f_X(x) \cdot f_Y(y) = (x + 1/2)(y + 1/2)$ .

Since these are not equal,  $X$  and  $Y$  are not independent.

(d)  $E(X) = \int_0^1 \int_0^1 x(x + y) dy dx = \int_0^1 \left[ x^2y + x\frac{y^2}{2} \Big|_0^1 \right] dx = \int_0^1 x^2 + \frac{x}{2} dx = \frac{7}{12}$ .

(Or, using (b),  $E(X) = \int_0^1 xf_X(x) dx = \int_0^1 x(x + 1/2) dx = 7/12$ .)

By symmetry  $E(Y) = 7/12$ .

$E(X^2 + Y^2) = \int_0^1 \int_0^1 (x^2 + y^2)(x + y) dy dx = \frac{5}{6}$ .

$E(XY) = \int_0^1 \int_0^1 xy(x + y) dy dx = \frac{1}{3}$ .

$\text{Cov}(X, Y) = E(XY) - E(X)E(Y) = \frac{1}{3} - \frac{49}{144} = -\frac{1}{144}$ .

**Problem 21.**

Standardize:

$$\begin{aligned} P\left(\sum_i X_i < 30\right) &= P\left(\frac{\frac{1}{n}\sum X_i - \mu}{\sigma/\sqrt{n}} < \frac{30/n - \mu}{\sigma/\sqrt{n}}\right) \\ &\approx P\left(Z < \frac{30/100 - 1/5}{1/30}\right) \quad (\text{by the central limit theorem}) \\ &= P(Z < 3) \\ &= 1 - .0013 = .9987 \quad (\text{from the table}) \end{aligned}$$

**Problem 22.** Let  $\bar{X} = \frac{X_1 + \dots + X_{144}}{144} \Rightarrow E(\bar{X}) = 2$ , and  $\sigma_{\bar{X}} = 2/12 = 1/6$ . ( $\sqrt{n} = 12$ )

A chain of algebra gives  $P(X_1 + \dots + X_{144}) + P\left(\bar{X} > \frac{264}{144}\right) = P(\bar{X} > 1.8333)$ .

Standardization gives  $P(\bar{X} > 1.8333) = P\left(\frac{\bar{X} - 2}{1/6} > \frac{1.8333 - 2}{1/6}\right) = P\left(\frac{\bar{X} - 2}{1/6} > -1.0\right)$

Now, the Central limit theorem says  $P\left(\frac{\bar{X} - 2}{1/6} > -1.0\right) \approx P(Z > -1) = \boxed{.84}$

**Problem 23.** Let  $X_j$  be the IQ of a randomly selected person. We are given  $E(X_j) = 100$  and  $\sigma_{X_j} = 15$ .

Let  $\bar{X}$  be the average of the IQ's of 100 randomly selected people. We have  $E(\bar{X}) = 100$  and  $\sigma_{\bar{X}} = 15/\sqrt{100} = 1.5$ .

The problem asks for  $P(\bar{X} > 115)$ . Standardizing we get  $P(\bar{X} > 115) \approx P(Z > 10)$ .

This is effectively 0.

**Problem 24.** Data mean and variance  $\bar{x} = 65$ ,  $s^2 = 35.778$ . The number of degrees of freedom is 9. We look up  $t_{9,.025} = 2.262$  in the  $t$ -table The 95% confidence interval is

$$\left[\bar{x} - \frac{t_{9,.025}s}{\sqrt{n}}, \bar{x} + \frac{t_{9,.025}s}{\sqrt{n}}\right] = \left[65 - 2.262\sqrt{3.5778}, 65 + 2.262\sqrt{3.5778}\right] = [60.721, 69.279]$$

**Problem 25.** Suppose we have taken data  $x_1, \dots, x_n$  with mean  $\bar{x}$ . Remember in these probabilities  $\mu$  is a given (fixed) hypothesis.

$$P(|\bar{x} - \mu| \leq .5 \mid \mu) = .95 \Leftrightarrow P\left(\frac{|\bar{x} - \mu|}{\sigma/\sqrt{n}} < \frac{.5}{\sigma/\sqrt{n}} \mid \mu\right) = .95 \Leftrightarrow P\left(|Z| < \frac{.5\sqrt{n}}{5}\right) = .95.$$

Using the table, we have precisely that  $\frac{.5\sqrt{n}}{5} = 1.96$ . So,  $n = (19.6)^2 = \boxed{384}$ .

If we use our rule of thumb that the .95 interval is  $2\sigma$  we have  $\sqrt{n}/10 = 2 \Rightarrow n = 400$ .

**Problem 26.** The rule-of-thumb is that a 95% confidence interval is  $\bar{x} \pm 1/\sqrt{n}$ . To be within 1% we need

$$\frac{1}{\sqrt{n}} = .01 \Rightarrow n = 10000.$$

Using  $z_{.025} = 1.96$  instead the 95% confidence interval is

$$\bar{x} \pm \frac{z_{.025}}{2\sqrt{n}}.$$

To be within 1% we need

$$\frac{z_{.025}}{2\sqrt{n}} = .01 \Rightarrow n = 9604.$$

Note, we are using the standard Bernoulli approximation  $\sigma \leq 1/2$ .

**Problem 27.** The 90% confidence interval is

$$\bar{x} \pm \frac{z_{.05}}{2\sqrt{n}} = \bar{x} \pm \frac{1.64}{40}$$

We want  $\bar{x} - \frac{1.64}{40} > .5$ , that is  $\bar{x} > .541$ .

So  $\frac{\text{number preferring A}}{400} > .541$ . So,

$$\text{number preferring A} > 216.4$$

**Problem 28.** A 95% confidence means about 5% = 1/20 will be wrong. You'd expect about 2 to be wrong.

With a probability  $p = .05$  of being wrong, the number wrong follows a Binomial(40,  $p$ ) distribution. This has expected value 2, and standard deviation  $\sqrt{40(.05)(.95)} = 1.38$ . 10 wrong is  $(10-2)/1.38 = 5.8$  standard deviations from the mean. This would be surprising.

**Problem 29.** We have  $n = 27$  and  $s^2 = 5.86$ . If we fix a hypothesis for  $\sigma^2$  we know

$$\frac{(n-1)s^2}{\sigma^2} \sim \chi_{n-1}^2$$

We used R to find the critical values. (Or use the  $\chi^2$  table.)

$$c_{025} = \text{qchisq}(.975, 26) = 41.923$$

$$c_{975} = \text{qchisq}(.025, 26) = 13.844$$

The 95% confidence interval for  $\sigma^2$  is

$$\left[ \frac{(n-1) \cdot s^2}{c_{.025}}, \frac{(n-1) \cdot s^2}{c_{.975}} \right] = \left[ \frac{26 \cdot 5.86}{41.923}, \frac{26 \cdot 5.86}{13.844} \right] = [3.6343, 11.0056]$$

We can take square roots to find the 95% confidence interval for  $\sigma$

$$[1.9064, 3.3175]$$

**Problem 30.** (a) The model is  $y_i = a + bx_i + \varepsilon_i$ , where  $\varepsilon_i$  is random error. We assume the errors are independent with mean 0 and the same variance for each  $i$  (homoscedastic).

The total error squared is

$$E^2 = \sum (y_i - a - bx_i)^2 = (1 - a - b)^2 + (1 - a - 2b)^2 + (3 - a - 3b)^2$$

The least squares fit is given by the values of  $a$  and  $b$  which minimize  $E^2$ . We solve for them by setting the partial derivatives of  $E^2$  with respect to  $a$  and  $b$  to 0. In R we found that  $a = 1.0$ ,  $b = 0.5$

(b) This is similar to part (a). The model is

$$y_i = ax_{i,1} + bx_{i,2} + c + \varepsilon_i$$

where the errors  $\varepsilon_i$  are independent with mean 0 and the same variance for each  $i$  (homoscedastic).

The total error squared is

$$E^2 = \sum (y_i - ax_{i,1} - bx_{i,2} - c)^2 = (3 - a - 2b - c)^2 + (5 - 2a - 3b - c)^2 + (1 - 3a - c)^2$$

The least squares fit is given by the values of  $a$ ,  $b$  and  $c$  which minimize  $E^2$ . We solve for them by setting the partial derivatives of  $E^2$  with respect to  $a$ ,  $b$  and  $c$  to 0. In R we found that  $a = 0.5$ ,  $b = 1.5$ ,  $c = -0.5$

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