

# Classical Adjoint (Adjugate) Matrix: Advanced Problems

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## 1 Adjugate and Matrix Transformations

**Problem 1.1.** *Similarity Transformation*

**Question 1.1.1.** *Prove that if  $B = P^{-1}AP$  for an invertible matrix  $P$ , then  $\text{adj}(B) = P^{-1} \text{adj}(A)P$  [cite: 104].*

**Hint 1.1.1.1.** Start with the definition  $B = P^{-1}AP$ . Apply the adjugate operation to both sides:  $\text{adj}(B) = \text{adj}(P^{-1}AP)$ .

**Hint 1.1.1.2.** Recall the property  $\text{adj}(XYZ) = \text{adj}(Z) \text{adj}(Y) \text{adj}(X)$  [derived by applying  $\text{adj}(MN) = \text{adj}(N) \text{adj}(M)$  twice, see cite: 49]. Apply this to  $\text{adj}(P^{-1}AP)$ .

**Hint 1.1.1.3.** You should have  $\text{adj}(B) = \text{adj}(P) \text{adj}(A) \text{adj}(P^{-1})$ . Now we need to relate  $\text{adj}(P)$  and  $\text{adj}(P^{-1})$ .

**Hint 1.1.1.4.** Recall the inverse formula:  $M^{-1} = \frac{1}{\det(M)} \text{adj}(M)$ . This implies  $\text{adj}(M) = \det(M)M^{-1}$ . Apply this to  $M = P$  and  $M = P^{-1}$ .

**Hint 1.1.1.5.** Express  $\text{adj}(P)$  in terms of  $\det(P)$  and  $P^{-1}$ . Express  $\text{adj}(P^{-1})$  in terms of  $\det(P^{-1})$  and  $(P^{-1})^{-1} = P$ . Remember  $\det(P^{-1}) = 1/\det(P)$ .

**Hint 1.1.1.6.** Substitute the expressions for  $\text{adj}(P)$  and  $\text{adj}(P^{-1})$  from the previous hint back into the equation from Hint 3:  $\text{adj}(B) = (\det(P)P^{-1}) \text{adj}(A)((\det P)^{-1}P)$ .

**Hint 1.1.1.7.** Simplify the expression by canceling the determinant factors. Does the result match the desired  $P^{-1} \text{adj}(A)P$ ?

## 2 Connections to Characteristic Polynomial and Eigenvectors

**Problem 2.1.** *Cayley-Hamilton Theorem and Adjugate*

**Question 2.1.1.** *Let  $p(\lambda) = \det(\lambda I - A) = \lambda^n + c_{n-1}\lambda^{n-1} + \cdots + c_1\lambda + c_0$  be the characteristic polynomial of  $A$ . Show that  $\text{adj}(A) = -(A^{n-1} + c_{n-1}A^{n-2} + \cdots + c_2A + c_1I)$  [cite: 148]. (Note: Requires careful handling of polynomial definition and potentially the identity  $\text{adj}(\lambda I - A)(\lambda I - A) = p(\lambda)I$ ).*

**Hint 2.1.1.1.** Recall the Cayley-Hamilton Theorem:  $p(A) = A^n + c_{n-1}A^{n-1} + \cdots + c_1A + c_0I = 0$  [cite: 147].

**Hint 2.1.1.2.** Consider the matrix polynomial  $B(\lambda) = \text{adj}(\lambda I - A)$ . The entries of  $B(\lambda)$  are cofactors of  $\lambda I - A$ , which are polynomials in  $\lambda$  of degree at most  $n-1$ . We can write  $B(\lambda) = B_{n-1}\lambda^{n-1} + \cdots + B_1\lambda + B_0$ , where  $B_k$  are matrix coefficients.

**Hint 2.1.1.3.** Use the fundamental identity:  $(\lambda I - A) \text{adj}(\lambda I - A) = \det(\lambda I - A)I = p(\lambda)I$ . Substitute the polynomial forms:  $(\lambda I - A)(B_{n-1}\lambda^{n-1} + \cdots + B_0) = (\lambda^n + c_{n-1}\lambda^{n-1} + \cdots + c_0)I$ .

**Hint 2.1.1.4.** Expand the left side and collect terms by powers of  $\lambda$ .  $\lambda^n B_{n-1} + \lambda^{n-1}(B_{n-2} - AB_{n-1}) + \cdots + \lambda(B_0 - AB_1) - AB_0$ .

**Hint 2.1.1.5.** Equate the matrix coefficients of powers of  $\lambda$  on both sides of the identity from Hint 3.  $\lambda^n$ :  $B_{n-1} = I$   $\lambda^{n-1}$ :  $B_{n-2} - AB_{n-1} = c_{n-1}I$  ...  $\lambda^1$ :  $B_0 - AB_1 = c_1I$   $\lambda^0$ :  $-AB_0 = c_0I$

**Hint 2.1.1.6.** From the coefficient equations, express  $B_{n-2}, B_{n-3}, \dots, B_0$  recursively in terms of  $A$  and the coefficients  $c_k$ .  $B_{n-1} = I$   $B_{n-2} = AB_{n-1} + c_{n-1}I = A + c_{n-1}I$   $B_{n-3} = AB_{n-2} + c_{n-2}I = A(A + c_{n-1}I) + c_{n-2}I = A^2 + c_{n-1}A + c_{n-2}I$  ... Continue this pattern. What is the general expression for  $B_k$ ?

**Hint 2.1.1.7.** The constant term of the polynomial  $B(\lambda) = \text{adj}(\lambda I - A)$  is  $B_0$ . Find  $B_0$  by setting  $\lambda = 0$ :  $B(0) = \text{adj}(-A)$ .

**Hint 2.1.1.8.** We know  $\text{adj}(-A) = (-1)^{n-1} \text{adj}(A)$ . So,  $B_0 = (-1)^{n-1} \text{adj}(A)$ .

**Hint 2.1.1.9.** From the recursive definition in Hint 6, what expression do you get for  $B_0$ ? It should be  $B_0 = A^{n-1} + c_{n-1}A^{n-2} + \cdots + c_1I$ . (Verify this also satisfies  $-AB_0 = c_0I$  using Cayley-Hamilton).

**Hint 2.1.1.10.** Equate the two expressions for  $B_0$ :  $(-1)^{n-1} \text{adj}(A) = A^{n-1} + c_{n-1}A^{n-2} + \cdots + c_1I$ . Does this match the target expression  $\text{adj}(A) = -(A^{n-1} + \cdots + c_1I)$ ? It matches if  $n$  is even. If  $n$  is odd,  $(-1)^{n-1} = 1$ , giving  $\text{adj}(A) = A^{n-1} + \cdots + c_1I$ . There might be a sign convention difference in the polynomial definition or the source formula [cite: 148]. Let's re-check the source: it uses  $-(\dots)$ . The standard derivation often leads to  $(-1)^{n-1}(\dots)$ . Let's assume the source formula [cite: 148] is the target. How can we reconcile this? Check the relation  $c_0 = \det(-A) = (-1)^n \det(A)$  [cite: 149]. The equation  $-AB_0 = c_0I$  is key. If  $B_0 = (-1)^{n-1} \text{adj}(A)$ , then  $-A((-1)^{n-1} \text{adj}(A)) = c_0I$ .  $(-1)^n A \text{adj}(A) = c_0I$ .  $(-1)^n \det(A)I = c_0I$ . This matches  $c_0 = (-1)^n \det(A)$ . Now equate  $B_0 = A^{n-1} + \cdots + c_1I$  with  $B_0 = (-1)^{n-1} \text{adj}(A)$  to solve for  $\text{adj}(A)$ .  $\text{adj}(A) = (-1)^{n-1}(A^{n-1} + c_{n-1}A^{n-2} + \cdots + c_1I)$ . Compare with [cite: 148]. The formula in [cite: 148] seems to be off by a factor of  $(-1)^{n-1}$  based on this standard derivation unless their  $c_k$  definition differs. Assuming the derivation is correct, the result is  $\text{adj}(A) = (-1)^{n-1}(A^{n-1} + \cdots + c_1I)$ .

**Question 2.1.2.** Let  $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$ . Find the eigenvalues  $\lambda$  of  $A$ . For each eigenvalue, compute  $B = A - \lambda I$  and find a non-zero column of  $\text{adj}(B)$ . Verify that this column is an eigenvector of  $A$  corresponding to  $\lambda$ .

**Hint 2.1.2.1.** First, find the eigenvalues of  $A$  by solving the characteristic equation  $\det(A - \lambda I) = 0$ .

**Hint 2.1.2.2.** For the first eigenvalue  $\lambda_1$ , calculate the matrix  $B_1 = A - \lambda_1 I$ .

**Hint 2.1.2.3.** Calculate  $\text{adj}(B_1)$ . Since  $A - \lambda_1 I$  is singular,  $\det(B_1) = 0$ . What does this imply about  $B_1 \text{adj}(B_1)$ ?

**Hint 2.1.2.4.** Find  $\text{adj}(B_1)$  using the  $2 \times 2$  shortcut. Is it the zero matrix? (It shouldn't be if the eigenvalue has geometric multiplicity 1, which is true here).

**Hint 2.1.2.5.** Let  $\mathbf{c}_1$  be any non-zero column of  $\text{adj}(B_1)$ . Verify that  $B_1 \mathbf{c}_1 = \mathbf{0}$ . (This follows from  $B_1 \text{adj}(B_1) = 0$ ).

**Hint 2.1.2.6.** Since  $B_1 \mathbf{c}_1 = (A - \lambda_1 I) \mathbf{c}_1 = \mathbf{0}$ , rewrite this as  $A \mathbf{c}_1 = \lambda_1 \mathbf{c}_1$ . Does this confirm  $\mathbf{c}_1$  is an eigenvector for  $\lambda_1$ ?

**Hint 2.1.2.7.** Repeat steps 2-6 for the second eigenvalue  $\lambda_2$ .

### 3 Further Exploration of Identities and Properties

#### Problem 3.1. *Adjugate Iteration and Determinants*

**Question 3.1.1.** Let  $A$  be an  $n \times n$  matrix. Find an expression for  $\text{adj}(\text{adj}(\text{adj}(A)))$  in terms of  $A$  and  $\det(A)$ . Assume  $n \geq 2$ .

**Hint 3.1.1.1.** Let  $B = \text{adj}(A)$ . We know  $\text{adj}(B) = \text{adj}(\text{adj}(A)) = (\det A)^{n-2} A$  [cite: 67].

**Hint 3.1.1.2.** Let  $C = \text{adj}(B) = \text{adj}(\text{adj}(A))$ . We want to find  $\text{adj}(C)$ .

**Hint 3.1.1.3.** Use the formula  $\text{adj}(C) = \det(C)C^{-1}$  if  $C$  is invertible. When is  $C = (\det A)^{n-2} A$  invertible? (Assume  $A$  is invertible for now).

**Hint 3.1.1.4.** Calculate  $\det(C) = \det((\det A)^{n-2} A)$ . Use  $\det(kA) = k^n \det(A)$ .  $\det(C) = ((\det A)^{n-2})^n \det(A) = (\det A)^{n(n-2)+1}$ .

**Hint 3.1.1.5.** Calculate  $C^{-1} = ((\det A)^{n-2} A)^{-1}$ . Use  $(kA)^{-1} = k^{-1} A^{-1}$ .  $C^{-1} = ((\det A)^{n-2})^{-1} A^{-1} = (\det A)^{-(n-2)} A^{-1}$ .

**Hint 3.1.1.6.** Substitute  $\det(C)$  and  $C^{-1}$  into  $\text{adj}(C) = \det(C)C^{-1}$ . Simplify the expression involving powers of  $\det(A)$ . Recall  $A^{-1} = (1/\det A) \text{adj}(A)$ .

**Hint 3.1.1.7.**  $\text{adj}(C) = (\det A)^{n(n-2)+1} (\det A)^{-(n-2)} A^{-1} = (\det A)^{n^2-2n+1-n+2} A^{-1} = (\det A)^{n^2-3n+3} A^{-1}$ . This is one form. Can we relate it back to  $A$  or  $\text{adj}(A)$ ?

**Hint 3.1.1.8.** Alternative approach: Apply the  $\text{adj}(\text{adj}(X))$  formula to  $X = \text{adj}(A)$ .  $\text{adj}(\text{adj}(\text{adj}(A))) = (\det(\text{adj } A))^{n-2} \text{adj}(A)$ .

**Hint 3.1.1.9.** Substitute  $\det(\text{adj } A) = (\det A)^{n-1}$ .  $\text{adj}(\text{adj}(\text{adj}(A))) = ((\det A)^{n-1})^{n-2} \text{adj}(A) = (\det A)^{(n-1)(n-2)} \text{adj}(A)$ .

**Hint 3.1.1.10.** Does this formula hold even if  $A$  is singular? Consider the ranks. If  $\text{rank}(A) \leq n-2$ , then  $\text{adj}(A) = 0$ ,  $\text{adj}(\text{adj}(A)) = 0$ ,  $\text{adj}(\text{adj}(\text{adj}(A))) = 0$ . The formula gives  $(\det A)^k \text{adj}(A) = 0 \cdot 0 = 0$ . OK. If  $\text{rank}(A) = n-1$ , then  $\det A = 0$ .  $\text{adj}(A)$  has rank 1.  $\text{adj}(\text{adj}(A)) = 0$  (if  $n \geq 3$ ). Then  $\text{adj}(\text{adj}(\text{adj}(A))) = \text{adj}(0) = 0$  (if  $n \geq 3$ ). The formula gives  $0^k \text{adj}(A) = 0$ . OK for  $n \geq 3$ . Check  $n = 2$ :  $\text{adj}(\text{adj}(\text{adj}(A))) = (\det A)^{(2-1)(2-2)} \text{adj}(A) = (\det A)^0 \text{adj}(A) = \text{adj}(A)$ . Let's verify.  $\text{adj}(A)$  is  $\begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$ .  $\text{adj}(\text{adj}(A)) = A$ .  $\text{adj}(\text{adj}(\text{adj}(A))) = \text{adj}(A)$ . It holds. So the formula is  $(\det A)^{(n-1)(n-2)} \text{adj}(A)$ .

### Problem 3.2. Rank Relationships Nuances

**Question 3.2.1.** Let  $A$  be an  $n \times n$  matrix with  $n \geq 3$ . Suppose  $\text{rank}(A) = n - 1$ . We know  $\text{rank}(\text{adj}(A)) = 1$ . Let  $\mathbf{u}$  span the 1D nullspace of  $A^T$  (left nullspace of  $A$ ) and  $\mathbf{v}$  span the 1D nullspace of  $A$  (right nullspace). Show that  $\text{adj}(A)$  can be written in the form  $\text{adj}(A) = c\mathbf{u}\mathbf{v}^T$  for some non-zero scalar  $c$ .

**Hint 3.2.1.1.** Since  $\text{rank}(\text{adj}(A)) = 1$ , all columns of  $\text{adj}(A)$  are scalar multiples of a single non-zero vector, say  $\mathbf{w}_1$ . Similarly, all rows of  $\text{adj}(A)$  are scalar multiples of a single non-zero vector, say  $\mathbf{w}_2^T$ . This is a property of rank 1 matrices, which can always be written as an outer product  $\mathbf{a}\mathbf{b}^T$ . So  $\text{adj}(A) = \mathbf{a}\mathbf{b}^T$ .

**Hint 3.2.1.2.** We know that  $A\text{adj}(A) = 0$ . Substituting  $\text{adj}(A) = \mathbf{a}\mathbf{b}^T$ , we get  $A(\mathbf{a}\mathbf{b}^T) = (A\mathbf{a})\mathbf{b}^T = 0$ . Since  $\mathbf{b}^T$  is non-zero (otherwise  $\text{adj}(A) = 0$ ), this implies  $A\mathbf{a} = \mathbf{0}$ . What does this tell us about the vector  $\mathbf{a}$  in relation to the nullspace of  $A$ ?

**Hint 3.2.1.3.** Since the nullspace of  $A$  is spanned by  $\mathbf{v}$ ,  $\mathbf{a}$  must be a scalar multiple of  $\mathbf{v}$ . So,  $\mathbf{a} = k_1\mathbf{v}$  for some  $k_1 \neq 0$ .

**Hint 3.2.1.4.** Now consider  $\text{adj}(A)A = 0$ . Substituting  $\text{adj}(A) = \mathbf{a}\mathbf{b}^T$ , we get  $(\mathbf{a}\mathbf{b}^T)A = \mathbf{a}(\mathbf{b}^T A) = 0$ . Since  $\mathbf{a}$  is non-zero, this implies  $\mathbf{b}^T A = \mathbf{0}^T$ .

**Hint 3.2.1.5.** Taking the transpose,  $A^T\mathbf{b} = \mathbf{0}$ . What does this tell us about the vector  $\mathbf{b}$  in relation to the nullspace of  $A^T$ ?

**Hint 3.2.1.6.** Since the nullspace of  $A^T$  is spanned by  $\mathbf{u}$ ,  $\mathbf{b}$  must be a scalar multiple of  $\mathbf{u}$ . So,  $\mathbf{b} = k_2\mathbf{u}$  for some  $k_2 \neq 0$ .

**Hint 3.2.1.7.** Substitute  $\mathbf{a} = k_1\mathbf{v}$  and  $\mathbf{b} = k_2\mathbf{u}$  back into  $\text{adj}(A) = \mathbf{a}\mathbf{b}^T$ .  $\text{adj}(A) = (k_1\mathbf{v})(k_2\mathbf{u})^T = (k_1k_2)\mathbf{v}\mathbf{u}^T$ . Wait, the target was  $c\mathbf{u}\mathbf{v}^T$ . Let's re-check the standard definition.

**Hint 3.2.1.8.** Rethink: The columns of  $\text{adj}(A)$  are in the nullspace of  $A$  (spanned by  $\mathbf{v}$ ). So  $\text{adj}(A)$  must have the form  $\mathbf{v}\mathbf{y}^T$  for some vector  $\mathbf{y}$ . The rows of  $\text{adj}(A)$  are related to the nullspace of  $A^T$ .  $(\text{adj}(A))^T = \text{adj}(A^T)$ . The columns of  $\text{adj}(A^T)$  are in the nullspace of  $A^T$  (spanned by  $\mathbf{u}$ ). So  $\text{adj}(A^T) = \mathbf{u}\mathbf{z}^T$ . Then  $\text{adj}(A) = (\text{adj}(A^T))^T = (\mathbf{u}\mathbf{z}^T)^T = \mathbf{z}\mathbf{u}^T$ .

**Hint 3.2.1.9.** Comparing the two forms:  $\text{adj}(A) = \mathbf{v}\mathbf{y}^T = \mathbf{z}\mathbf{u}^T$ . This implies  $\mathbf{z}$  must be proportional to  $\mathbf{v}$  and  $\mathbf{y}$  must be proportional to  $\mathbf{u}$ . Let  $\mathbf{z} = c\mathbf{v}$  and  $\mathbf{y} = d\mathbf{u}$ . Then  $\text{adj}(A) = c\mathbf{v}\mathbf{u}^T$ . This seems consistent. Where did the target  $c\mathbf{u}\mathbf{v}^T$  come from? Let's re-read the question statement. Ah, it asks to show  $\text{adj}(A) = c\mathbf{u}\mathbf{v}^T$ . Let's test this form.

**Hint 3.2.1.10.** If  $\text{adj}(A) = c\mathbf{u}\mathbf{v}^T$ . Check  $A\text{adj}(A) = A(c\mathbf{u}\mathbf{v}^T) = c(A\mathbf{u})\mathbf{v}^T$ . Is  $A\mathbf{u} = 0$ ? No,  $\mathbf{u}$  is in the left nullspace ( $A^T\mathbf{u} = 0$ ). Check  $\text{adj}(A)A = (c\mathbf{u}\mathbf{v}^T)A = c\mathbf{u}(\mathbf{v}^T A)$ . Is  $\mathbf{v}^T A = \mathbf{0}^T$ ? Yes, since  $A^T\mathbf{v} = \mathbf{0}$  is not necessarily true. We know  $A\mathbf{v} = 0$ . So  $(\mathbf{v}^T A)^T = A^T(\mathbf{v}^T)^T = A^T\mathbf{v}$ . No, this doesn't help. We know  $A\mathbf{v} = 0$ .

**Hint 3.2.1.11.** Let's reconsider  $A\text{adj}(A) = 0$ . Columns of  $\text{adj}(A)$  must be in nullspace( $A$ ), spanned by  $\mathbf{v}$ . So  $\text{adj}(A) = \mathbf{v}\mathbf{y}^T$ . Let's reconsider  $\text{adj}(A)A = 0$ . Rows of  $\text{adj}(A)$  must be in the left nullspace of  $A$  (nullspace of  $A^T$ ), spanned by  $\mathbf{u}^T$ . So  $\text{adj}(A) = \mathbf{z}\mathbf{u}^T$ . Equating these gives  $\mathbf{v}\mathbf{y}^T = \mathbf{z}\mathbf{u}^T$ . This holds if  $\mathbf{z} = k\mathbf{v}$  and  $\mathbf{y}^T = l\mathbf{u}^T$ . So  $\text{adj}(A) = (k\mathbf{v})(l\mathbf{u})^T = (kl)\mathbf{v}\mathbf{u}^T$ . This form seems correct. Let  $c = kl$ .  $\text{adj}(A) = c\mathbf{v}\mathbf{u}^T$ . The question might have swapped  $\mathbf{u}$  and  $\mathbf{v}$  roles or their definitions (row vs column). Assuming  $\mathbf{u}$  spans  $\text{Null}(A^T)$  and  $\mathbf{v}$  spans  $\text{Null}(A)$ , the result is  $\text{adj}(A) = c\mathbf{v}\mathbf{u}^T$ . If the question intended  $\mathbf{u}$  for  $\text{Null}(A)$  and  $\mathbf{v}$  for  $\text{Null}(A^T)$ , the result follows. Let's stick to the derivation  $\text{adj}(A) = c\mathbf{v}\mathbf{u}^T$ .

## 4 Block Matrices and Cramer's Rule

**Problem 4.1.** *Block Matrix Adjugate (Specific Entries)*

**Question 4.1.1.** Let  $A = \begin{pmatrix} B & \mathbf{u} \\ \mathbf{v}^T & \alpha \end{pmatrix}$ , where  $B$  is  $(n-1) \times (n-1)$ ,  $\mathbf{u}, \mathbf{v}$  are column vectors, and  $\alpha$  is a scalar. Find the entry in the bottom-right corner  $(n, n)$  of  $\text{adj}(A)$ , and find the column vector in the last column, positions 1 to  $n-1$  [see cite: 264-287 for context].

**Hint 4.1.1.1.** The entry  $(\text{adj } A)_{nn}$  is the cofactor  $C_{nn}(A)$ . How is this cofactor defined?

**Hint 4.1.1.2.**  $C_{nn}(A) = (-1)^{n+n} M_{nn}(A) = M_{nn}(A)$ . What is the minor  $M_{nn}(A)$ ? It's the determinant of the matrix obtained by removing row  $n$  and column  $n$  from  $A$ . What matrix block remains? Calculate its determinant.

**Hint 4.1.1.3.** Now consider the entries  $(\text{adj } A)_{in}$  for  $i = 1, \dots, n-1$ . These are the cofactors  $C_{ni}(A)$ .

**Hint 4.1.1.4.**  $C_{ni}(A) = (-1)^{n+i} M_{ni}(A)$ . What is the minor  $M_{ni}(A)$ ? It's the determinant of the matrix obtained by removing row  $n$  and column  $i$  from  $A$ . Let's call this submatrix  $S_{ni}$ .

**Hint 4.1.1.5.** Write down the structure of  $S_{ni}$ . Its rows are the first  $n-1$  rows of  $A$ . Its columns are the first  $n-1$  columns of  $A$  excluding column  $i$ , plus the last column of  $A$ .  $S_{ni} = \begin{pmatrix} B_{\text{no col } i} & \mathbf{u} \\ \mathbf{v}^T_{\text{no col } i?} & \alpha \end{pmatrix}$ . This requires careful interpretation. Let's reconsider  $M_{ni}$ . Remove row  $n$  ( $\mathbf{v}^T, \alpha$ ) and column  $i$  (column  $i$  of  $B$ , and  $v_i$ ). The remaining matrix is  $\begin{pmatrix} B_{\text{no row } n, \text{ col } i} & \mathbf{u}_{\text{no row } n?} \\ \mathbf{v}^T_{\text{no col } i?} & ?? \end{pmatrix}$ . No, this is getting confusing.

**Hint 4.1.1.6.** Let's use a known formula (e.g., from Wikipedia's "Adjugate matrix" page under block matrices, often derived alongside the block inverse formula). The block corresponding to  $(\text{adj } A)_{1:n-1, n}$  is given by  $-\text{adj}(B)\mathbf{u}$ .

**Hint 4.1.1.7.** Let's try to verify this. Recall  $A \text{adj}(A) = \det(A)I$ . The last column of this product should be  $\det(A)\mathbf{e}_n$ . This column is  $A \cdot (\text{last col of adj } A)$ . Let the last column of  $\text{adj } A$  be  $\begin{pmatrix} \mathbf{y} \\ \delta \end{pmatrix}$ . We found  $\delta = C_{nn} = \det(B)$ . Let's assume  $\mathbf{y} = -\text{adj}(B)\mathbf{u}$ .

Then  $A \begin{pmatrix} -\text{adj}(B)\mathbf{u} \\ \det(B) \end{pmatrix} = \begin{pmatrix} B & \mathbf{u} \\ \mathbf{v}^T & \alpha \end{pmatrix} \begin{pmatrix} -\text{adj}(B)\mathbf{u} \\ \det(B) \end{pmatrix} = \begin{pmatrix} -B \text{adj}(B)\mathbf{u} + \mathbf{u} \det(B) \\ -\mathbf{v}^T \text{adj}(B)\mathbf{u} + \alpha \det(B) \end{pmatrix}$ .

**Hint 4.1.1.8.** Simplify the top block:  $-\det(B)I\mathbf{u} + \det(B)\mathbf{u} = -\det(B)\mathbf{u} + \det(B)\mathbf{u} = \mathbf{0}$ . This matches the top  $n-1$  entries of  $\det(A)\mathbf{e}_n$ , which are zeros.

**Hint 4.1.1.9.** Simplify the bottom block:  $\alpha \det(B) - \mathbf{v}^T \text{adj}(B)\mathbf{u}$ . Recall the formula for the determinant of a block matrix (Schur complement):  $\det(A) = \alpha \det(B) - \mathbf{v}^T \text{adj}(B)\mathbf{u}$ . This matches the required  $(n, n)$  entry of  $\det(A)I$ .

**Hint 4.1.1.10.** So, the bottom-right entry is  $\det(B)$  and the rest of the last column is the vector  $-\text{adj}(B)\mathbf{u}$ .

**Question 4.1.2.** Explain how Cramer's Rule  $x_i = \frac{\det(A_i)}{\det(A)}$  for solving  $A\mathbf{x} = \mathbf{b}$  follows directly from the adjugate inverse formula  $\mathbf{x} = \frac{1}{\det(A)} \text{adj}(A)\mathbf{b}$ .

**Hint 4.1.2.1.** Start with  $\mathbf{x} = \frac{1}{\det(A)} \operatorname{adj}(A)\mathbf{b}$ . Let's look at the  $i$ -th component,  $x_i$ .

**Hint 4.1.2.2.**  $x_i = \frac{1}{\det(A)} (\text{row } i \text{ of } \operatorname{adj}(A)) \cdot \mathbf{b}$ .

**Hint 4.1.2.3.** What are the entries of row  $i$  of  $\operatorname{adj}(A)$ ? Remember  $(\operatorname{adj} A)_{ij} = C_{ji}(A)$ . So row  $i$  is  $((\operatorname{adj} A)_{i1}, \dots, (\operatorname{adj} A)_{in}) = (C_{1i}(A), C_{2i}(A), \dots, C_{ni}(A))$ .

**Hint 4.1.2.4.** Substitute this into the dot product:  $x_i = \frac{1}{\det(A)} \sum_{j=1}^n (\operatorname{adj} A)_{ij} b_j = \frac{1}{\det(A)} \sum_{j=1}^n C_{ji}(A) b_j$ .

**Hint 4.1.2.5.** Consider the sum  $\sum_{j=1}^n b_j C_{ji}(A)$ . Recall the Laplace (cofactor) expansion for a determinant along a column  $k$ :  $\det(M) = \sum_{j=1}^n M_{jk} C_{jk}(M)$ .

**Hint 4.1.2.6.** Let  $A_i$  be the matrix  $A$  with its  $i$ -th column replaced by the vector  $\mathbf{b}$ . Consider the Laplace expansion of  $\det(A_i)$  along its  $i$ -th column. The entries of this column are  $b_1, \dots, b_n$ . The cofactors  $C_{ji}(A_i)$  for this column are calculated from the submatrices formed by removing row  $j$  and column  $i$  from  $A_i$ . Are these submatrices the same as those used to calculate  $C_{ji}(A)$ ? Yes, because column  $i$  was removed.

**Hint 4.1.2.7.** Therefore, the Laplace expansion of  $\det(A_i)$  along column  $i$  is  $\det(A_i) = \sum_{j=1}^n (A_i)_{ji} C_{ji}(A_i) = \sum_{j=1}^n b_j C_{ji}(A)$ .

**Hint 4.1.2.8.** Compare the sum in Hint 7 with the sum in Hint 4. They are identical.

**Hint 4.1.2.9.** Substitute  $\sum_{j=1}^n C_{ji}(A) b_j = \det(A_i)$  back into the expression for  $x_i$  from Hint 4. Does this yield Cramer's Rule?