

# Assignment 03

## 1

*Proof.*

$$AB = I \implies (AB)A = A \implies A(BA) = A \implies BA = I.$$

□

## 2

Assume  $f, g \in V$ . Then

$$\int_0^1 [cf(t)]^2 dt = c^2 \int_0^1 f^2(t) dt < \infty \implies cf(t) \in V.$$

Also,

$$\begin{aligned} \int_0^1 [f(t) + g(t)]^2 dt &= \int_0^1 f^2(t) dt + \int_0^1 g^2(t) dt + 2 \int_0^1 f(t)g(t) dt \\ &\leq \int_0^1 f^2(t) dt + \int_0^1 g^2(t) dt + 2 \int_0^1 |f(t)g(t)| dt \\ &\leq \int_0^1 f^2(t) dt + \int_0^1 g^2(t) dt + \int_0^1 f^2(t) + g^2(t) dt \\ &< \infty. \end{aligned}$$

Hence

$$f(t) + g(t) \in V.$$

$V$  is a vector space.

## 3

### (a)

Linearly independent.

### (b)

Linearly dependent.

$$\vec{v}_1 - 3\vec{v}_2 + 2\vec{v}_3 + \vec{v}_4 = \vec{0}.$$

**4**

$$\left\{ \begin{bmatrix} 1 \\ 0 \\ -2 \\ 1 \end{bmatrix}, \begin{bmatrix} -2 \\ 3 \\ -3 \\ 4 \end{bmatrix}, \begin{bmatrix} 3 \\ -6 \\ 1 \\ -3 \end{bmatrix} \right\}$$

**5**

**(a)**

Yes. The set of all  $2 \times 2$  diagonal matrices is a subset of  $\mathbb{R}^{2 \times 2}$  closed under linear combination.

**(b)**

No.  $U_{2 \times 2} + L_{2 \times 2} = M_{2 \times 2}$ , which is no longer triangular.

**(c)**

Yes. The set of all  $2 \times 2$  lower triangular matrices is a subset of  $\mathbb{R}^{2 \times 2}$  closed under linear combination.

**(d)**

No. The sum of two such matrices is a matrix  $B$  with  $b_{12} = 2$ .

**(e)**

Yes. The set of all  $2 \times 2$  matrices  $B$  with  $b_{11} = 0$  is a subset of  $\mathbb{R}^{2 \times 2}$  closed under linear combination.

**(f)**

Yes. The set of all symmetric  $2 \times 2$  matrices is a subset of  $\mathbb{R}^{2 \times 2}$  closed under linear combination.

**(g)**

No. For example,  $A = \begin{bmatrix} 1 & 1 \\ 0 & 0 \end{bmatrix}$ ,  $B = \begin{bmatrix} 0 & 0 \\ 1 & 2 \end{bmatrix}$  are two singular matrices. But their sum  $A + B = \begin{bmatrix} 1 & 1 \\ 1 & 2 \end{bmatrix}$  is non-singular.

**6**

**(a)**

$$\begin{bmatrix} 18/5 \\ 4/5 \\ 7/5 \end{bmatrix}$$

**(b)**

$$x_3 \begin{bmatrix} 2 \\ -1 \\ 1 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -5 \\ 1 \\ 0 \\ 1 \end{bmatrix} + \begin{bmatrix} -5 \\ 1 \\ 0 \\ 0 \end{bmatrix}, (x_3, x_4) \in \mathbb{R}^2.$$

**7**

**(a)**

$$\begin{bmatrix} 1 & 2 & -3 \\ -2 & -2 & 3 \\ 2 & 4 & 6 \end{bmatrix} \xrightarrow{E_{12} E_{13} E_{23} E_{31} E_{21}} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 12 \end{bmatrix}.$$

Three column vectors are linearly independent. Hence

$$\dim(\cdot) = 3.$$

**(b)**

Let

$$\alpha(x^2 - 4) + \beta x^2(x^4 - 2) + \theta(x^6 - 8) = 0.$$

Simplifying the equation, we obtain

$$(\beta + \theta)x^6 + (\alpha - 2\beta)x^2 - 4(\alpha + 2\theta) = 0$$

which yields

$$\alpha = 2\beta = -2\theta.$$

This shows that

$$x^2(x^4 - 2) \in \text{span}(x^2 - 4, x^6 - 8).$$

Hence  $\{x^2 - 4, x^6 - 8\}$  is a basis for the original span.

$$\dim(\cdot) = 2.$$

**8**

**(a)**

*Proof.* Let  $S = \{\vec{x}_1, \dots, \vec{x}_n\}$  be any set of  $n$  linearly independent vectors in  $V$ . We need to prove that every vector in  $V$  is in the span of vectors in  $S$ , by contradiction.

Suppose  $\exists \vec{v} \in V$  s.t.  $\vec{v} \notin \text{span}(\vec{x}_1, \dots, \vec{x}_n)$ , then  $\vec{v}, \vec{x}_1, \dots, \vec{x}_n$  are  $(n + 1)$  L.I. vectors in  $V$ . We now use induction to show this is impossible to happen with  $\dim(V) = n$ , since any  $(n + 1)$  vectors in  $n$ -dimensional vector

space  $V$  must be L.D.

**Base case.** If  $n = 1$ , a basis for  $V$  contains exactly one vector  $\vec{b}_1$ . Suppose  $\vec{v}_1 = c_1 \vec{b}_1$  and  $\vec{v}_2 = c_2 \vec{b}_1$  are two non-zero vectors in  $V$  (otherwise they are already L.D.). Clearly we may write  $\vec{v}_1/c_1 - \vec{v}_2/c_2 = \vec{0}$ , showing  $\vec{v}_1, \vec{v}_2$  as L.D.

**Inductive step.** Suppose in any  $(n - 1)$ -dimensional vector space  $V$ , any  $n$  vectors are L.D. We want to show this implies that in any  $n$ -dimensional  $V$ , any  $(n + 1)$  vectors  $\vec{v}_1, \dots, \vec{v}_{n+1}$  are L.D.

Let  $B = \{\vec{b}_1, \dots, \vec{b}_n\}$  be a basis for  $V$ . Then we may write

$$\vec{v}_1 = a_{11}\vec{b}_1 + a_{12}\vec{b}_2 + \dots + a_{1n}\vec{b}_n \quad (1)$$

$$\vec{v}_2 = a_{21}\vec{b}_1 + a_{22}\vec{b}_2 + \dots + a_{2n}\vec{b}_n \quad (2)$$

$\vdots$

$$\vec{v}_n = a_{n1}\vec{b}_1 + a_{n2}\vec{b}_2 + \dots + a_{nn}\vec{b}_n \quad (n)$$

$$\vec{v}_{n+1} = a_{n+1,1}\vec{b}_1 + a_{n+1,2}\vec{b}_2 + \dots + a_{n+1,n}\vec{b}_n \quad (n+1)$$

Note that if  $a_{11}, a_{21}, \dots, a_{n+1,1}$  are all 0, then  $\vec{v}_1, \dots, \vec{v}_{n+1}$  are L.D. by the hypothesis. Hence we may assume, WLOG, that  $a_{11} \neq 0$ . Multiply (1) by  $\frac{a_{i1}}{a_{11}}$  and obtain

$$\frac{a_{i1}\vec{v}_1}{a_{11}} = a_{i1}\vec{b}_1 + L_i(\vec{b}_2, \vec{b}_3, \dots, \vec{b}_n) \quad (1^*)$$

where  $L_i(\cdot)$  is some linear combination of  $\vec{b}_2, \vec{b}_3, \dots, \vec{b}_n$ . Now subtract (1\*) from equations (2), ..., (n+1):

$$\vec{v}_2^* := \vec{v}_2 - \frac{a_{21}\vec{v}_1}{a_{11}} = L_2^*(\vec{b}_2, \vec{b}_3, \dots, \vec{b}_n) \quad (2^*)$$

$\vdots$

$$\vec{v}_n^* := \vec{v}_n - \frac{a_{n1}\vec{v}_1}{a_{11}} = L_n^*(\vec{b}_2, \vec{b}_3, \dots, \vec{b}_n) \quad (n^*)$$

$$\vec{v}_{n+1}^* := \vec{v}_{n+1} - \frac{a_{n+1,1}\vec{v}_1}{a_{11}} = L_{n+1}^*(\vec{b}_2, \vec{b}_3, \dots, \vec{b}_n) \quad (n+1^*)$$

Note that  $\vec{v}_2^*, \dots, \vec{v}_n^*, \vec{v}_{n+1}^*$  are  $n$  vectors in  $(n - 1)$ -dimensional vector space. By the hypothesis, they are L.D. i.e.  $\exists$  non-trivial  $\{\lambda_2, \dots, \lambda_n, \lambda_{n+1}\}$  s.t.

$$\lambda_2 \vec{v}_2^* + \dots + \lambda_n \vec{v}_n^* + \lambda_{n+1} \vec{v}_{n+1}^* = \vec{0}.$$

Equivalently

$$\lambda_2(\vec{v}_2 - \frac{a_{21}\vec{v}_1}{a_{11}}) + \dots + \lambda_n(\vec{v}_n - \frac{a_{n1}\vec{v}_1}{a_{11}}) + \lambda_{n+1}(\vec{v}_{n+1} - \frac{a_{n+1,1}\vec{v}_1}{a_{11}}) = \vec{0}.$$

Rearranging, we obtain

$$-(\frac{\lambda_2 a_{21}}{a_{11}} + \dots + \frac{\lambda_n a_{n1}}{a_{11}} + \frac{\lambda_{n+1} a_{n+1,1}}{a_{11}}) \vec{v}_1 + \lambda_2 \vec{v}_2 + \dots + \lambda_n \vec{v}_n + \lambda_{n+1} \vec{v}_{n+1} = \vec{0}.$$

showing  $\vec{v}_1, \dots, \vec{v}_{n+1}$  as L.D. This completes the induction and our proof.

**(b)**

□

*Proof.* Suppose  $V = \text{span}(\vec{x}_1, \dots, \vec{x}_n)$ . We want to show  $S = \{\vec{x}_1, \dots, \vec{x}_n\}$  are L.I. Then by **(a)**  $S$  forms a basis for  $V$ .

We shall prove by contradiction. Assume  $S$  is L.D. Further assume, WLOG, that  $S' = \{\vec{x}_1, \dots, \vec{x}_j\}$  is a maximal L.I. subset of  $S$ . Then  $V$  can be spanned solely by  $S'$ . But by **(a)** this creates a basis for  $V$  with only  $j$  vectors, implying  $\dim(V) = j \neq n$ , the desired contradiction.

□

**9**

$$\forall \vec{x} \in V, \vec{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = x_1 \begin{bmatrix} 1 \\ 0 \\ 0 \\ -1 \end{bmatrix} + x_2 \begin{bmatrix} 0 \\ 1 \\ 0 \\ -1 \end{bmatrix} + x_3 \begin{bmatrix} 0 \\ 0 \\ 1 \\ -1 \end{bmatrix}.$$

Hence

$$V = \text{span}\left(\begin{bmatrix} 1 \\ 0 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 1 \\ 0 \\ -1 \end{bmatrix}, \begin{bmatrix} 0 \\ 0 \\ 1 \\ -1 \end{bmatrix}\right). \quad (*)$$

The linear independence of  $S$  together with  $(*)$  implies that  $S$  is a basis for  $V$ . Moreover,

$$\dim(V) = |S| = 3.$$

**10**

**(a)**

From  $B$ , we observe the pivot variables at the first, second, and fifth columns. Hence a basis for  $\text{Col}(A)$

$$S = \{a_1, a_2, a_5\} = \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 7 \\ 2 \end{bmatrix} \right\}$$

and

$$\text{rank}(A) = |S| = 3.$$

**(b)**

First note that

$$N(A) = N(B).$$

Solving  $Bx = 0$ ,

$$x = \begin{bmatrix} -x_3 - x_4 + 14x_6 - 14x_7 \\ -x_3 - 2x_4 - 8x_6 + 5x_7 \\ x_3 \\ x_4 \\ -4x_6 + 2x_7 \\ x_6 \\ x_7 \end{bmatrix} = x_3 \begin{bmatrix} -1 \\ -1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} + x_4 \begin{bmatrix} -1 \\ -2 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix} + x_6 \begin{bmatrix} 14 \\ -8 \\ 0 \\ 0 \\ -4 \\ 1 \\ 0 \end{bmatrix} + x_7 \begin{bmatrix} -14 \\ 5 \\ 0 \\ 0 \\ 2 \\ 0 \\ 1 \end{bmatrix}.$$

A basis for  $N(B)$ , therefore also a basis for  $N(A)$ , is then

$$S = \left\{ \begin{bmatrix} -1 \\ -1 \\ 1 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} -1 \\ -2 \\ 0 \\ 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 14 \\ -8 \\ 0 \\ 0 \\ -4 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} -14 \\ 5 \\ 0 \\ 0 \\ 2 \\ 0 \\ 1 \end{bmatrix} \right\}.$$

Moreover,

$$\dim[N(A)] = |S| = 4.$$

## 11

Let

$$\alpha y_1 + \beta y_2 + \theta y_3 = 0.$$

Equivalently

$$\alpha(x_1 + x_2) + \beta(x_2 + x_3) + \theta(x_3 + x_1) = 0.$$

Rearranging,

$$(\alpha + \theta)x_1 + (\alpha + \beta)x_2 + (\beta + \theta)x_3 = 0.$$

From L.I. of  $\{x_i\}$  it follows that

$$\alpha + \theta = \alpha + \beta = \beta + \theta = 0,$$

which implies

$$\alpha = \beta = \theta = 0.$$

Hence  $\{y_i\}$  is also L.I.

## 12

(1)

$q = 3$ .

(2)

All  $q \neq 3$  will do.

(3)

Impossible;  $\text{rank}(A) \leq 2$ .

13

*Proof.* Denote  $\text{rank}(A) = r \leq m$ , which is the dimension of the row space of  $A$ . WLOG assume (through row permutation)  $\{\vec{a}_1, \vec{a}_2, \dots, \vec{a}_r\}$  is a maximal L.I. subset of  $\{\vec{a}_i\}$ , where  $\vec{a}_i$  denotes the  $i$ -th row of  $A$ . Note that  $AB = [\vec{a}_i B]$ . Then since  $B$  is invertible,  $\{\vec{a}_1 B, \vec{a}_2 B, \dots, \vec{a}_r B\}$  remains a maximal L.I. subset of  $\{(\vec{a}b)_i\}$ , where  $(\vec{a}b)_i$  denotes the  $i$ -th row of  $AB$ . Hence the dimension of the row space of  $AB$  remains to be  $r$ , i.e.

$$\text{rank}(AB) = \text{rank}(A).$$

□

14

*Proof.* Denote  $\text{rank}(A)$  as  $\alpha$ ,  $\text{rank}(B)$  as  $\beta$ , the  $i$ -th row of a matrix  $M$  as  $\vec{m}_i$ . WLOG assume  $\{\vec{a}_i\}_{i=1}^\alpha$  is a maximal L.I. subset of  $\{\vec{a}_i\}_{i=1}^m$ . Consider  $\vec{a}_1$ . Since  $AB = O$ ,

$$\vec{a}_1 B = [a_{11} \ a_{12} \ \dots \ a_{1n}] \begin{bmatrix} \vec{b}_1 \\ \vec{b}_2 \\ \vdots \\ \vec{b}_n \end{bmatrix} = \vec{0}, \quad (1)$$

with at least one entry in  $\vec{a}_1$  being non-zero. WLOG assume  $a_{1n} \neq 0$ , we may rewrite (1) as

$$\vec{b}_n = -\frac{1}{a_{1n}}(a_{11}\vec{b}_1 + a_{12}\vec{b}_2 + \dots + a_{1,n-1}\vec{b}_{n-1}).$$

Hence

$$\vec{b}_n \in \text{span}(\vec{b}_1, \dots, \vec{b}_{n-1}) \implies \text{Row}(B) = \text{span}(\vec{b}_1, \dots, \vec{b}_{n-1}). \quad (1^*)$$

It follows that

$$\beta \leq |\{\vec{b}_i\}_{i=1}^{n-1}| = n - 1.$$

Similarly, consider  $\vec{a}_2$ ,

$$\vec{a}_2 B = [a_{21} \ a_{22} \ \dots \ a_{2n}] \begin{bmatrix} \vec{b}_1 \\ \vec{b}_2 \\ \vdots \\ \vec{b}_n \end{bmatrix} = \vec{0}.$$

Combined with (1) a little rearrangement yields

$$(a_{1n}a_{21} - a_{2n}a_{11})\vec{b}_1 + \dots + (a_{1n}a_{2,n-2} - a_{2n}a_{1,n-2})\vec{b}_{n-2} + (a_{1n}a_{2,n-1} - a_{2n}a_{1,n-1})\vec{b}_{n-1} = \vec{0} \quad (2)$$

From L.I. of  $\{\vec{a}_i\}_{i=1}^\alpha$ , the coefficients  $(a_{1n}a_{2i} - a_{2n}a_{1i})_{i=1}^{n-1}$  cannot all be zero. It follows that  $\exists \vec{b}_u \in \{\vec{b}_i\}_{i=1}^{n-1}$  that we may take out without affecting the row space, thus

$$\beta \leq (n-1) - 1 = n-2.$$

Continuing the same process for every row in  $\{\vec{a}_i\}_{i=1}^\alpha$ . Each time  $\vec{a}_i B = \vec{0}$  and the L.I. of  $\{\vec{a}_i\}_{i=1}^\alpha$  together implies a new redundant row in  $\{\vec{b}_i\}_{i=1}^n$  for spanning  $\text{Row}(B)$ , reducing the upper bound for  $\beta$  by 1. We obtain at last

$$\beta \leq n - |\{\vec{a}_i\}_{i=1}^\alpha| = n - \alpha. \quad \square$$

## 15

*Proof.* If  $x \in N(A)$ , then

$$A^T A x = A^T (A x) = A 0 = 0 \implies x \in N(A^T A).$$

If  $x \in N(A^T A)$ , then

$$A^T A x = 0 \implies x^T A^T A x = 0 \iff (A x)^T A x = 0 \implies A x = 0 \implies x \in N(A).$$

Therefore

$$N(A) = N(A^T A). \quad \square$$

## 16

*Proof.* Denote  $\text{rank}(A)$  as  $\alpha$ ,  $\text{rank}(B)$  as  $\beta$ ,  $i$ -th column of a matrix  $M$  as  $m_i$ . WLOG assume  $\{a_i\}_{i=1}^\alpha$  is a maximal L.I. subset of  $\{a_i\}$ ,  $\{b_i\}_{i=1}^\beta$  a maximal L.I. subset of  $\{b_i\}$ . Then  $\{a_i\}_{i=1}^\alpha$  spans  $\text{Col}(A)$ ;  $\{b_i\}_{i=1}^\beta$  spans  $\text{Col}(B)$ . Consequently  $\{a_i\}_{i=1}^\alpha \cup \{b_i\}_{i=1}^\beta$  spans  $\text{Col}(A+B)$ , and

$$\text{rank}(A+B) \leq |\{a_i\}_{i=1}^\alpha \cup \{b_i\}_{i=1}^\beta| = \alpha + \beta. \quad \square$$

## 17

### (1)

$$T(A+B) = (A+B) + (A+B)^T = A + A^T + B + B^T = T(A) + T(B)$$



$$T(cA) = cA + (cA)^T = c(A + A^T) = c T(A).$$

**(2)**

Let

$$T(A) = A + A^T = O.$$

$$\begin{bmatrix} 2a_{11} & a_{12} + a_{21} \\ a_{21} + a_{12} & 2a_{22} \end{bmatrix} = \begin{bmatrix} 0 & 0 \\ 0 & 0 \end{bmatrix} \implies A = \begin{bmatrix} 0 & a_{12} \\ -a_{12} & 0 \end{bmatrix}.$$

Therefore

$$\ker(T) = \left\{ A \mid A = \begin{bmatrix} 0 & x \\ -x & 0 \end{bmatrix} \right\}.$$

**18**

*Proof.* Suppose  $L$  is one-to-one. Then

$$L(v_1) = L(v_2) \implies v_1 = v_2$$

$$\Downarrow$$

$$L(v_1) - L(v_2) = L(v_1 - v_2) = 0 \implies v_1 - v_2 = 0$$

$$\Downarrow$$

$$\ker(L) = 0.$$

Suppose  $\ker(L) = 0$ . Then

$$L(v) = 0 \implies v = 0$$

$$\Downarrow$$

$$L(v_1) - L(v_2) = L(v_1 - v_2) = 0 \implies v_1 - v_2 = 0$$

$$\Downarrow$$

$$L(v_1) = L(v_2) \implies v_1 = v_2.$$

$L$  is one-to-one. Hence

$$L \text{ is one-to-one} \iff \ker(L) = 0.$$

□

**19**

$$L : \mathbb{R}^3 \rightarrow \mathbb{R}^2.$$

This transformation is determined by

$$B = \left\{ \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 0 \\ 7 \\ 2 \end{bmatrix} \right\} \xrightarrow{L} \left\{ \begin{bmatrix} 2 \\ 1 \end{bmatrix}, \begin{bmatrix} 2 \\ 3 \end{bmatrix}, \begin{bmatrix} -1 \\ 2 \end{bmatrix} \right\} \xrightarrow{\text{Rep}_{B'}} \left\{ \begin{bmatrix} -2 \\ 1 \end{bmatrix}, \begin{bmatrix} -2 \\ 3 \end{bmatrix}, \begin{bmatrix} 1 \\ 2 \end{bmatrix} \right\}_{B'}.$$

Hence

$$\text{Rep}_{B,B'}(L) = \begin{bmatrix} -2 & -2 & 1 \\ 1 & 3 & 2 \end{bmatrix}_{B,B'}.$$

To find the  $L(\vec{u})$  and  $\text{Rep}_{B'}[L(\vec{u})]$ , first use the fact that

$$\text{Rep}_B(\vec{u}) = \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix}_B.$$

Then

$$\begin{aligned} \text{Rep}_{B'}[L(\vec{u})] &= \text{Rep}_{B,B'}(L) \text{Rep}_B(\vec{u}) \\ &= \begin{bmatrix} -2 & -2 & 1 \\ 1 & 3 & 2 \end{bmatrix} \begin{bmatrix} 2 \\ 1 \\ -1 \end{bmatrix}_B \\ &= \begin{bmatrix} -7 \\ 3 \end{bmatrix}_{B'} \end{aligned}$$

Converting back,

$$L(\vec{u}) = [\vec{u}'_1 \mid \vec{u}'_2] \begin{bmatrix} -7 \\ 3 \end{bmatrix}_{B'} = \begin{bmatrix} 7 \\ 3 \end{bmatrix}.$$

## 20

```
julia> using LinearAlgebra

julia> # pairwise inner products are zero

julia> a1'a2
0.0

julia> a1'a3
0.0

julia> a2'a3
0.0

julia> # expansion of x in basis {a1, a2, a3}

julia> norm((a1'x)a1 + (a2'x)a2 + (a3'x)a3 - x)
4.577566798522237e-16
```

## 21

```
julia> using LinearAlgebra
```

(1)

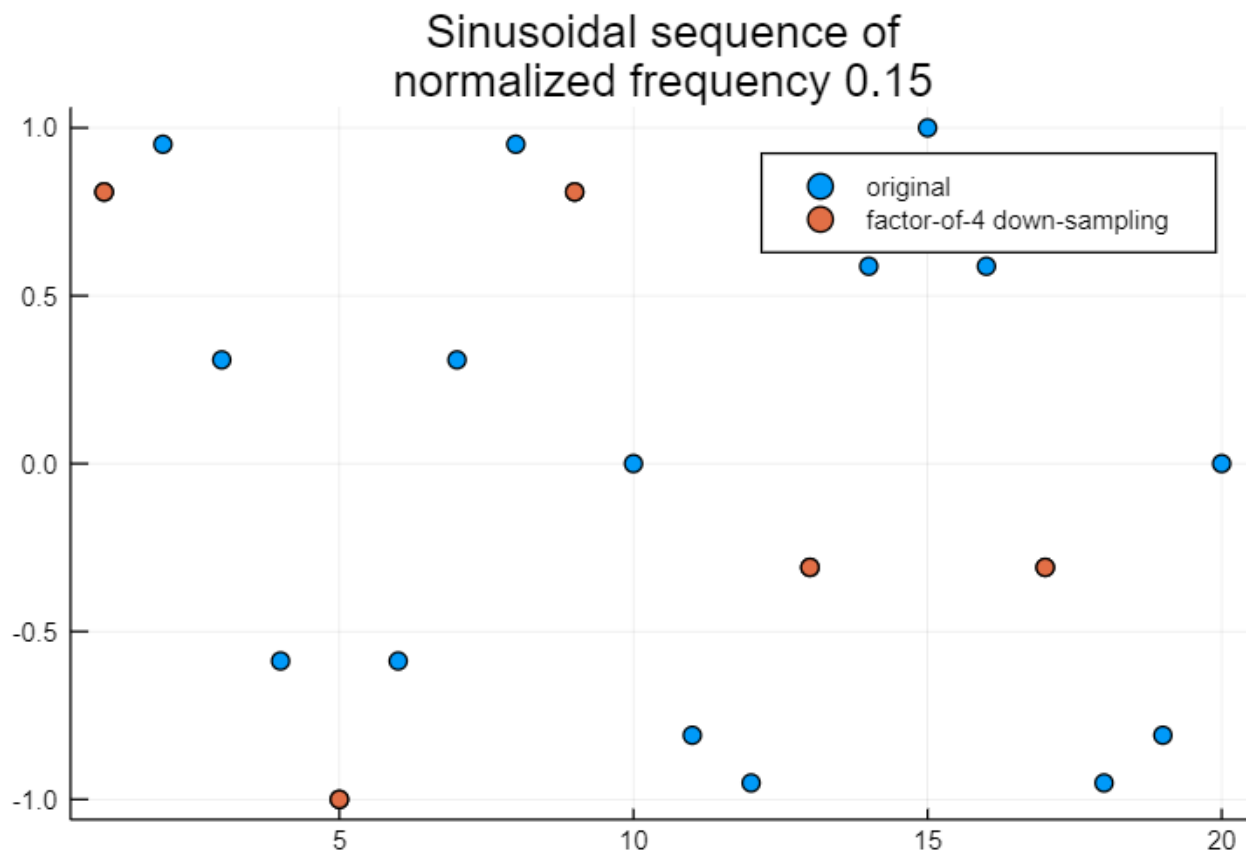
```
julia> rank(A * B) <= min(rank(A), rank(B))  
true
```

(2)

```
julia> rank(A + B) <= rank(A) + rank(B)  
true
```

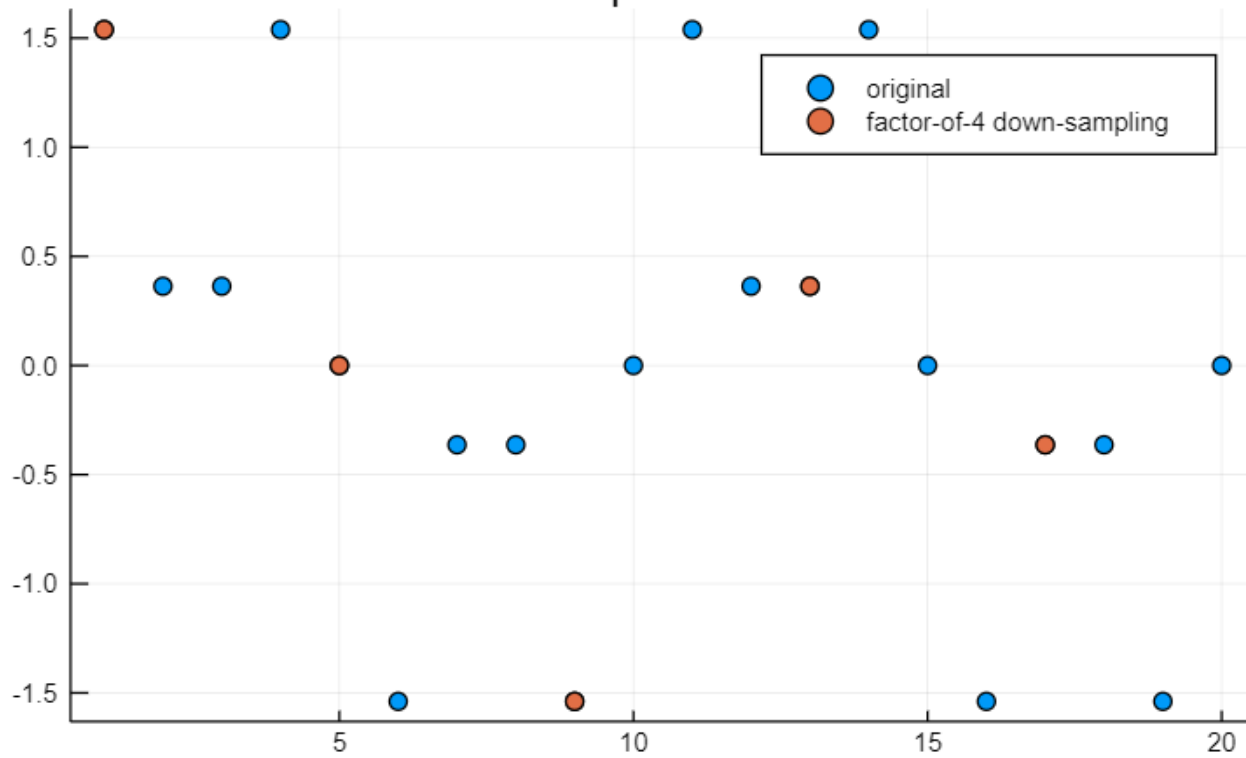
22

(i)



(ii)

Sum of sinusoidal sequences of  
normalized frequencies 0.1 and 0.3



23

```
julia> X = count(x -> (rand() < p), 1:N)
4
```

24

```
julia> total = 1e5;

julia> est_PI = 4 * hit / total
3.13388
```

Sampling distribution  
(N = 2000)

