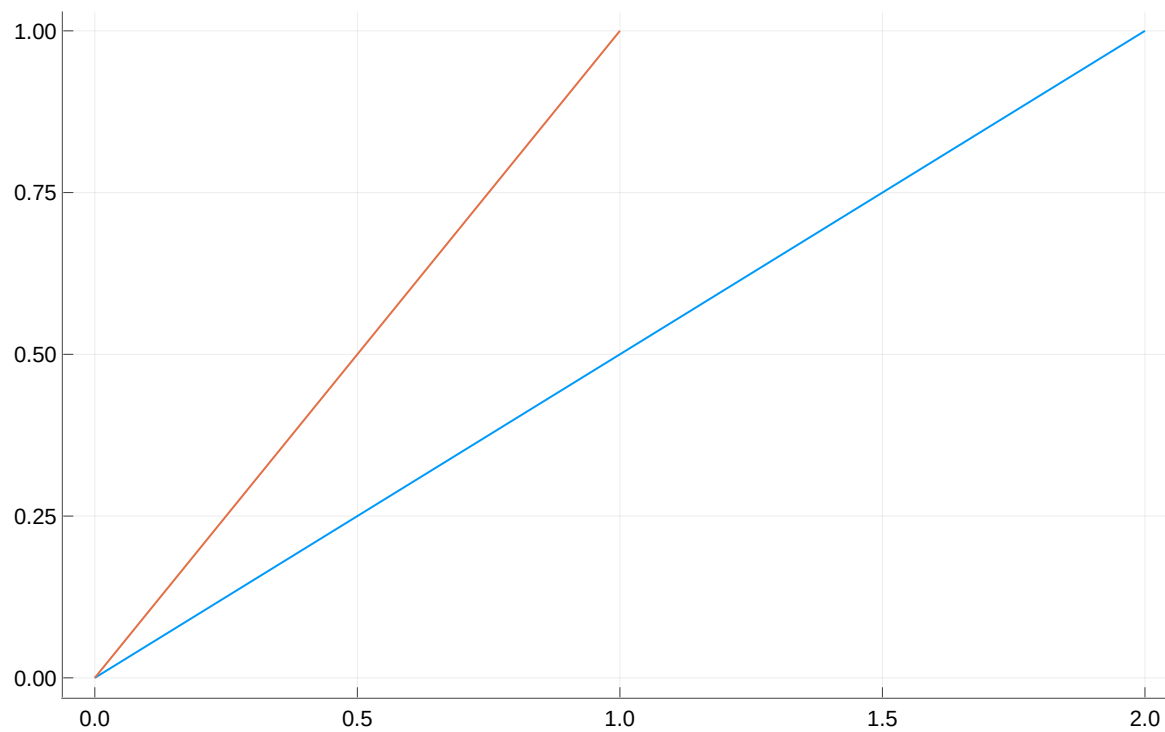


PlotlyBackend()

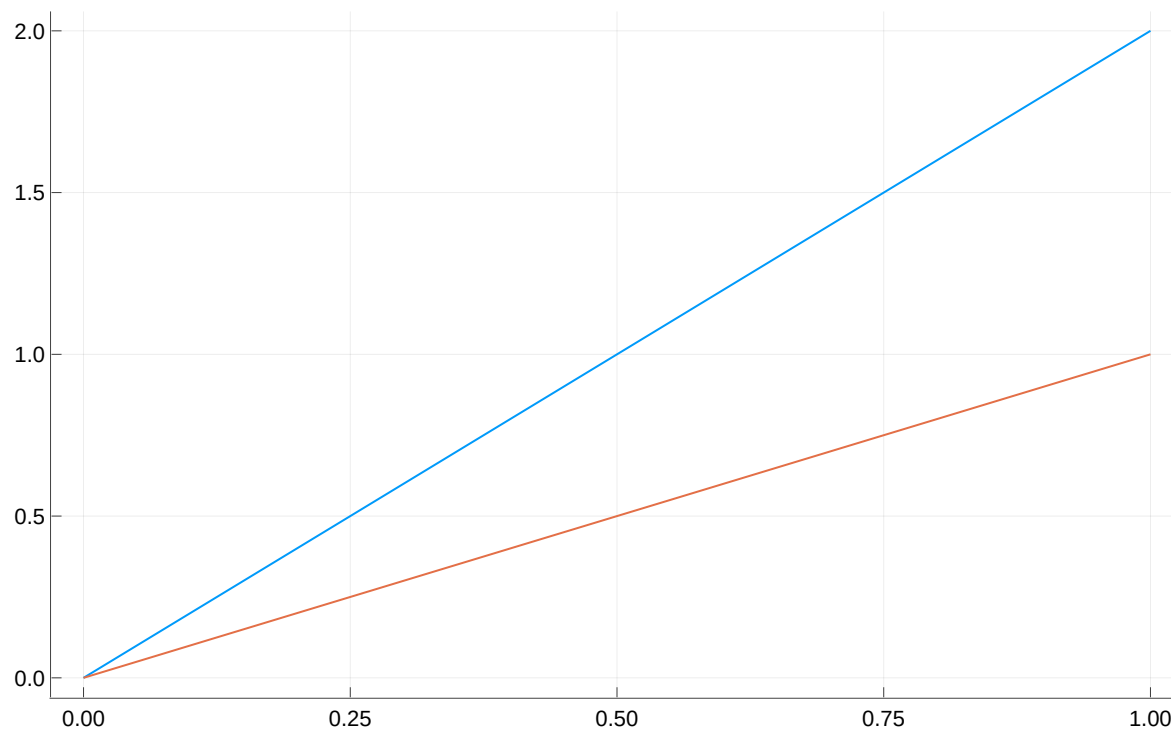
```
1 begin
2     using LinearAlgebra
3     include("./LA.jl")
4     using .LA
5     using RowEchelon
6     using InvertedIndices
7
8     using Plots
9     using PlutoUI
10    using Roots
11    plotly();
12 end
```

Scaling

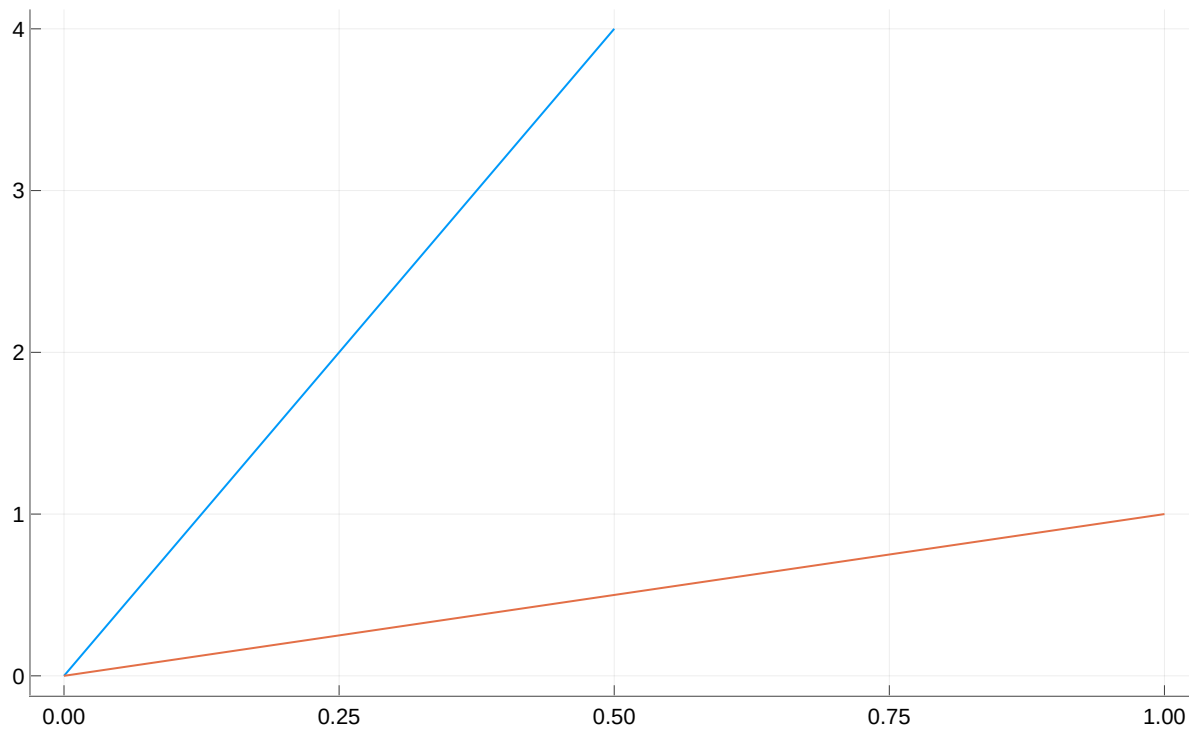
```
1 md"# Scaling"
```



```
1 begin
2     k = 2 #scale
3     A_x = [
4         k 0;
5         0 1;
6     ]
7
8     x = [
9         1;
10        1
11    ]
12
13     LA.graph_vectors([(A_x * x) x])
14 end
```



```
1 begin
2     k_y = 2 #scale
3     A_y = [
4         1 0;
5         0 k_y;
6     ]
7
8     x_y = [
9         1;
10        1
11    ]
12
13    LA.graph_vectors([(A_y * x_y) x_y])
14 end
```



```

1 begin
2     k_y_ = 4 #xscale
3     k_x_ = 1//2 #yscale
4     A_xy = [
5         k_x_ 0;
6         0 k_y_;
7     ]
8
9     X_xy = [
10        1;
11        1
12    ]
13
14    LA.graph_vectors([(A_xy * X_xy) X_xy])
15 end

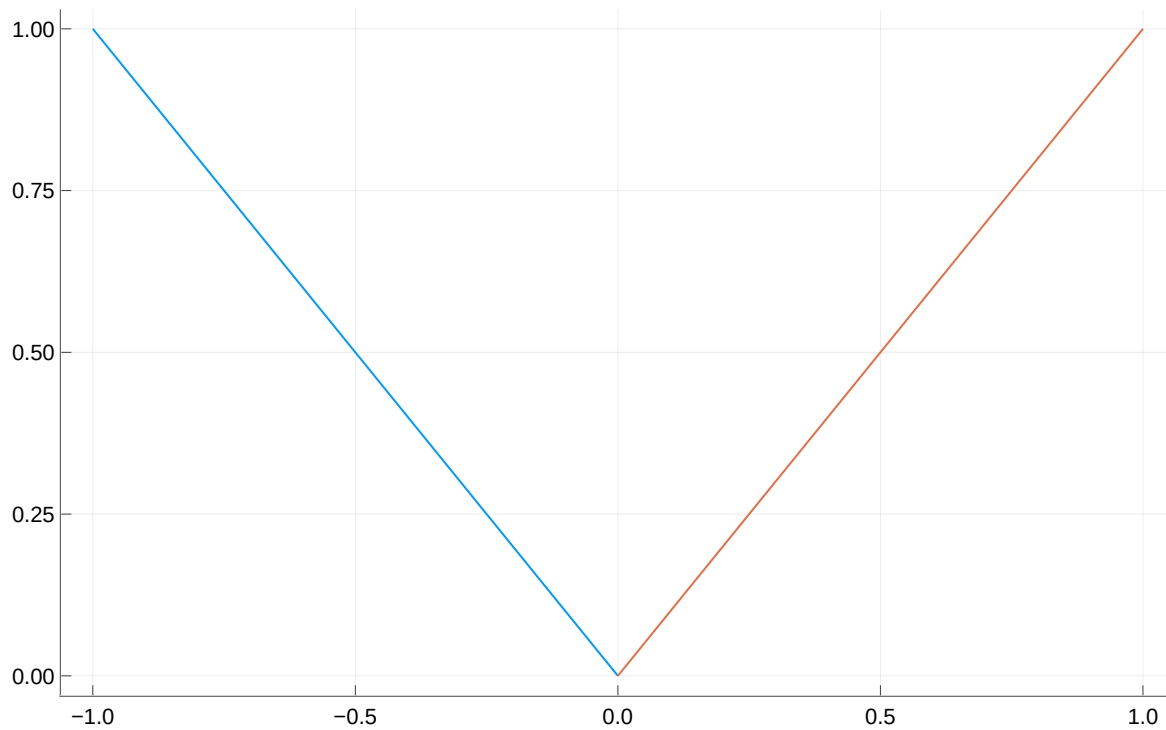
```

Rotating

```
1 md"# Rotating"
```

Always counter clockwise

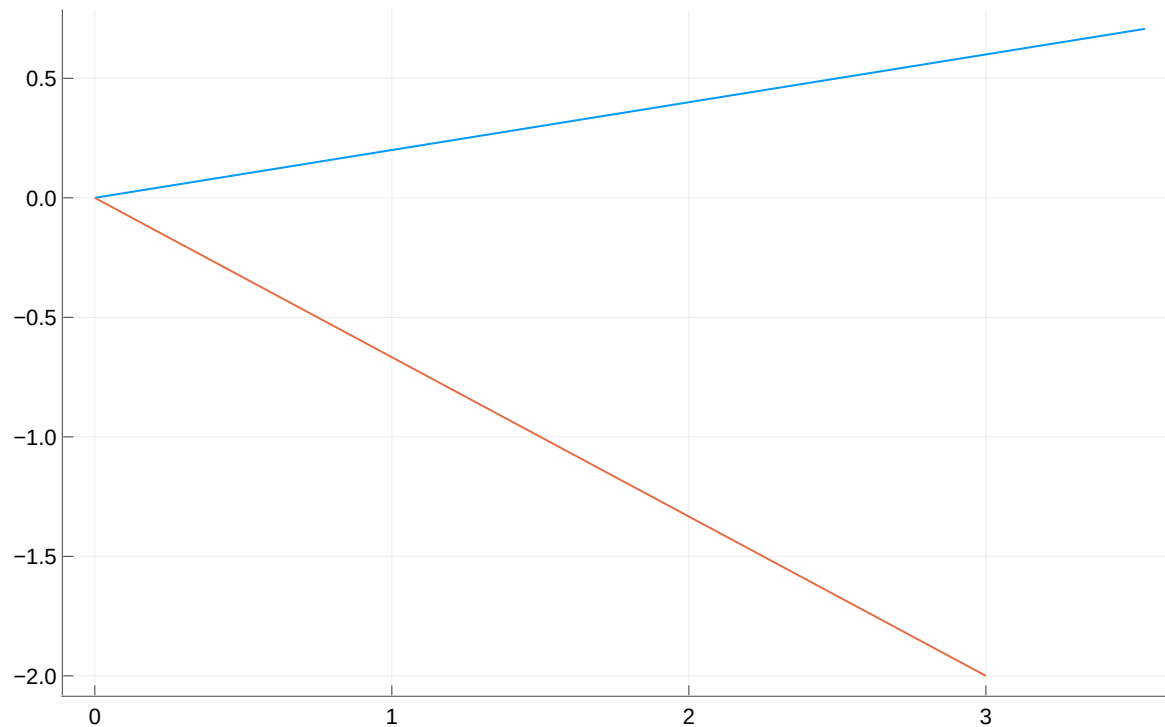
```
1 md"### Always counter clockwise"
```



```
1 begin
2   θ = pi/2
3   B = [
4       cos(θ) -sin(θ);
5       sin(θ) cos(θ)
6   ]
7
8   X_rot = [
9       1;
10      1
11   ]
12
13   println(B)
14
15   LA.graph_vectors([(B * X_rot) X_rot])
16 end
```

```
[6.123233995736766e-17 -1.0; 1.0 6.123233995736766e-17]
```

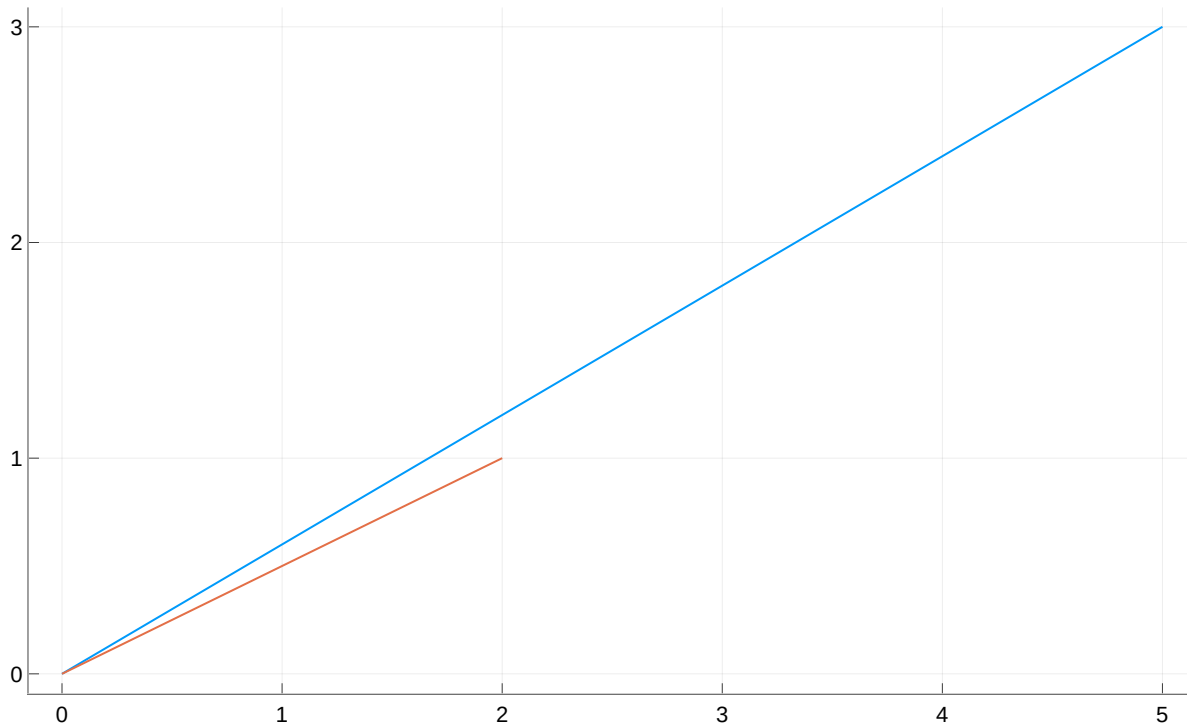




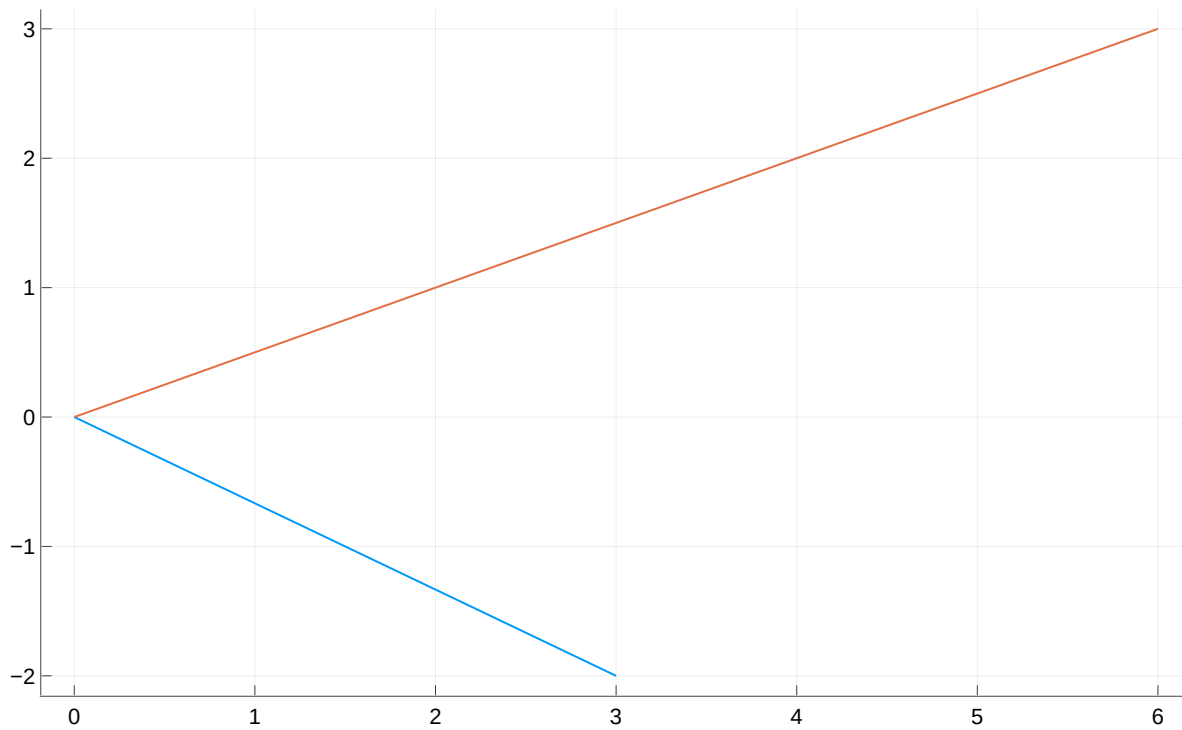
```
1 begin
2     θ_ = pi/4
3     X_rot_2 = [3 -2]'
4     B_ = [
5         cos(θ_) -sin(θ_);
6         sin(θ_) cos(θ_)
7     ]
8
9     LA.graph_vectors([(B_ * X_rot_2) X_rot_2])
10 end
```

Translation

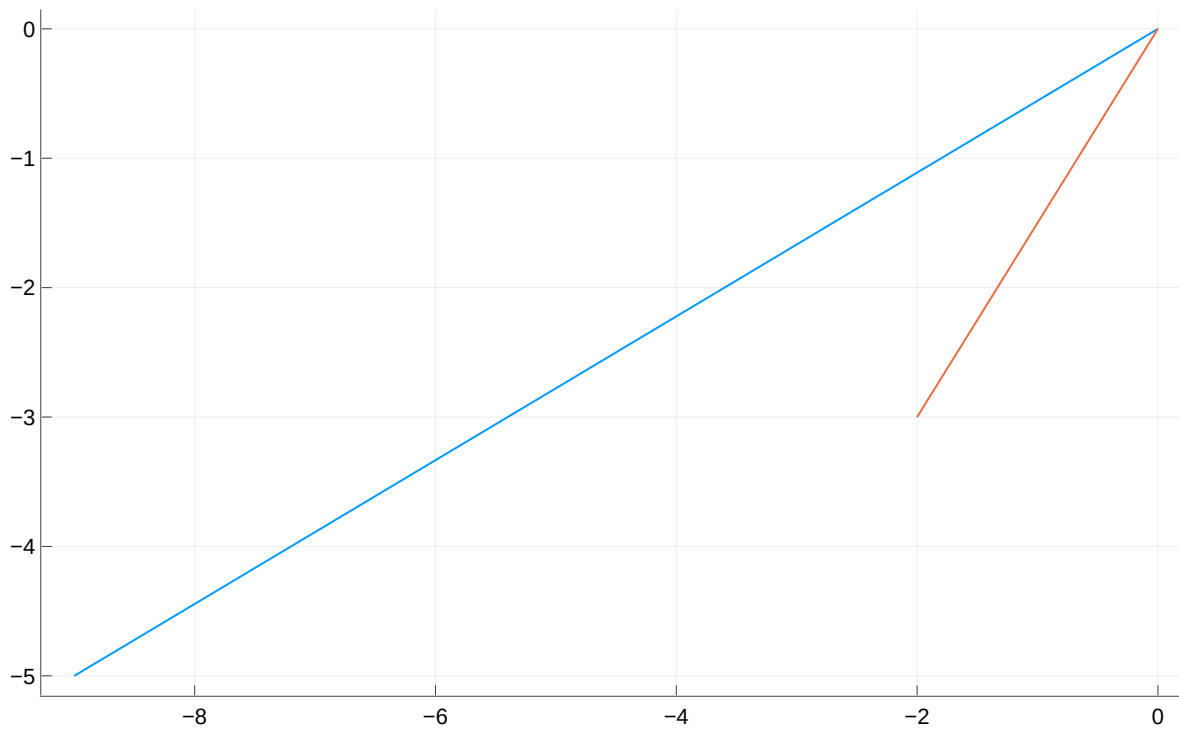
```
1 md"# Translation"
```



```
1 begin
2     # Addition
3     δ_x = 3
4     δ_y = 2
5     X_trans = [ 2 1]'
6
7     result = X_trans + [δ_x δ_y]'
8
9     LA.graph_vectors([result X_trans])
10
11 end
```



```
1 begin
2     # Multiplication
3     α_x = 3 # right 3
4     α_y = 5 # up 5
5     placeholder = [3 -2 1]'
6     A_mult = [
7         1 0 α_x;
8         0 1 α_y;
9         0 0 1
10    ]
11    result_mult = A_mult * placeholder
12    result_ = result_mult[1:end-1, :]
13    LA.graph_vectors([placeholder[1:end-1, :] result_])
14 end
```

```

1 begin
2     γ_x = -7 # left 7
3     γ_y = -2 # down 2
4     B_mult = [
5         1 0 γ_x;
6         0 1 γ_y;
7         0 0 1
8     ]
9     X_mult = [-2 -3 1]'
10    result_mult_2 = (B_mult * X_mult)[1:end-1, :]
11    LA.graph_vectors([result_mult_2 X_mult[1:end-1, :]])
12    # -9, -5
13 end

```

Rank

```
1 md"# Rank"
```

The number of nonzero rows in the row echelon form of the matrix.

The number of independent rows.

If transformation squishes matrix into a line, it is rank 1.

If transformation squishes matrix into a plane, it is rank 2.

Rank = Number of Dimensions of the output matrix.

```
1 md"""### If transformation squishes matrix into a line, it is rank 1.
2
3 ### If transformation squishes matrix into a plane, it is rank 2.
4
5 ### Rank = Number of Dimensions of the output matrix.
6
7 """
```

**Set of all possible outputs after transformation =
Column Space of A = Span of Columns in Matrix.**

Rank = Number of dimensions of column space.

Full Rank = Rank(A) == Dim(A)

Zero vector always included in column space.

For full rank, only $\langle 0,0 \rangle$ lands on $\langle 0,0 \rangle$ after transformation.

For non-full rank, many vectors can get squished into $\langle 0,0 \rangle$.

2d squished into a line (rank 1) => Full line of vectors that land on origin.

3d Transformation gets squished into a line (rank 1) => Full plane that lands on $\langle 0, 0, 0 \rangle$

3d transformation squished into plane (rank 2) => Full line of vectors that land on $\langle 0, 0, 0 \rangle$.

Set of vectors that land on origin = null space = kernel of matrix = space of all vectors that become null (land on zero vector).

When \mathbf{b} is zero vector ($\mathbf{Ax} = \mathbf{b}$), the null space gives you all the possible solutions to the equation

```

1 md"### Set of all possible outputs after transformation = Column Space of A = Span of
  Columns in Matrix.
2
3 ### Rank = Number of dimensions of column space.
4
5 ### Full Rank = Rank(A) == Dim(A)
6
7 Zero vector always included in column space.
8
9 For full rank, only  $\langle 0,0 \rangle$  lands on  $\langle 0,0 \rangle$  after transformation.
10
11 For non-full rank, many vectors can get squished into  $\langle 0,0 \rangle$ .
12
13 2d squished into a line (rank 1) => Full line of vectors that land on origin.
14
15 3d Transformation gets squished into a line (rank 1) => Full plane that lands on  $\langle 0, 0, 0 \rangle$ 
16
17 3d transformation squished into plane (rank 2) => Full line of vectors that land on
   $\langle 0, 0, 0 \rangle$ .
18
19
```

```
20 Set of vectors that land on origin = null space = kernel of matrix = space of all  
21 vectors that become null (land on zero vector).
```

```
22
```

```
### When b is zero vector ( $Ax = b$ ), the null space gives you all the possible
```

Nonsquare Matrices

3x2 matrix

3x2 matrix = transformation of 2D vectors into 3D vectors. Where i and j land.

Column space of 3x2 matrix (space where all vectors land after transformation) = 2D plane slicing through origin of 3D space.

But matrix is still full rank since $\dim(\text{Column Space}) == \dim(\text{input space})$.

2x3 Matrix

2x3 matrix = transformation of 3d Vectors into 2 dimensions. Where i, j, and k land.

3 columns == starting in space with 3 basis vectors (i, j, k) (or 3D).

2 rows == landing spot of 3d Vectors only land on space described by 2 coordinates => 2 dimensions.

1x2 Matrix

2D => 1D (number line).

2 columns == starting in space with 2 basis vectors (i, j) (or 2D)

1 row == landing space of 2d vectors only land on space described by 1 coordinate => number line.

Line of evenly spaced dots before transformation (not necessarily passing through origin) == line of evenly spaced dots after transformation (visual understand of what linearity means)

Same as dot product.

```

1 md" ### Nonsquare Matrices
2
3 #### 3x2 matrix
4
5 3x2 matrix = transformation of 2D vectors into 3D vectors. Where i and j land.
6
7 Column space of 3x2 matrix (space where all vectors land after transformation) = 2D
  plane slicing through origin of 3D space.
8
9 But matrix is still full rank since  $\dim(\text{Column Space}) == \dim(\text{input space})$ .
10
11 ----
12
13
14 #### 2x3 Matrix
15
16 2x3 matrix = transformation of 3d Vectors into 2 dimensions. Where i, j, and k land.
```

```

17
18 3 columns == starting in space with 3 basis vectors (i, j, k) (or 3D).
19
20 2 rows == landing spot of 3d Vectors only land on space described by 2 coordinates =>
    2 dimensions.
21
22
23 #### 1x2 Matrix
24
25 2D => 1D (number line).
26
27 2 columns == starting in space with 2 basis vectors (i, j) (or 2D)
28
29 1 row == landing space of 2d vectors only land on space described by 1 coordinate =>
    number line.
30
31 Line of evenly spaced dots before transformation (not necessarily passing through
    origin) == line of evenly spaced dots after transformation (visual understand of what
    linearity means)
32
33 Same as dot product.
34
35
36
37
38
39 "
40

```

1 Enter cell code...

```

2×2 Matrix{Float64}:
1.0  0.0
0.0  1.0

```

```

1 begin
2     C = [
3         1 2;
4         3 4
5     ]
6     rref(C)
7 end

```

2

1 rank(C)

```

D = 2×2 Matrix{Int64}:
 1  2
 2  4

```

```

1 D = [
2     1 2;
3     2 4
4 ]

```

```
2×2 Matrix{Float64}:  
 1.0  2.0  
 0.0  0.0
```

```
1 rref(D)
```

```
1
```

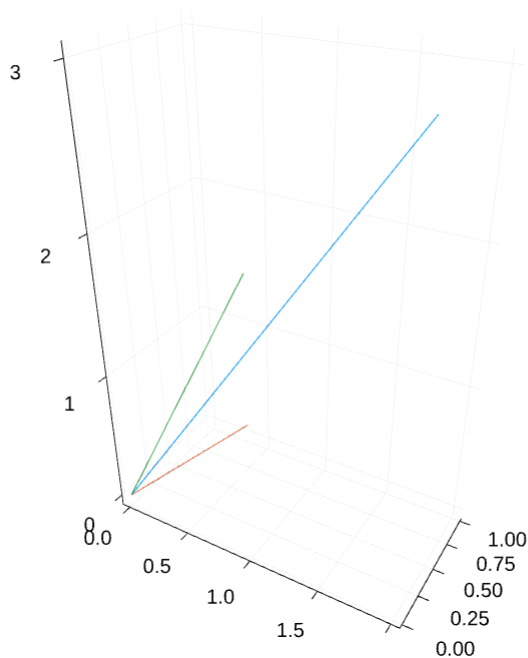
```
1 rank(D)
```

```
E = 3×3 Matrix{Int64}:  
 2  1  1  
 0  0  0  
 3  1  2
```

```
1 E = [  
2     2 1 1;  
3     0 0 0;  
4     3 1 2  
5  ]
```

```
3×3 Matrix{Float64}:  
 1.0  0.0  1.0  
 0.0  1.0 -1.0  
 0.0  0.0  0.0
```

```
1 rref(E)
```



```
1 LA.graph_vectors(E)
```

```
2
```

```
1 rank(E)
```

```
F = 3×2 Matrix{Int64}:
```

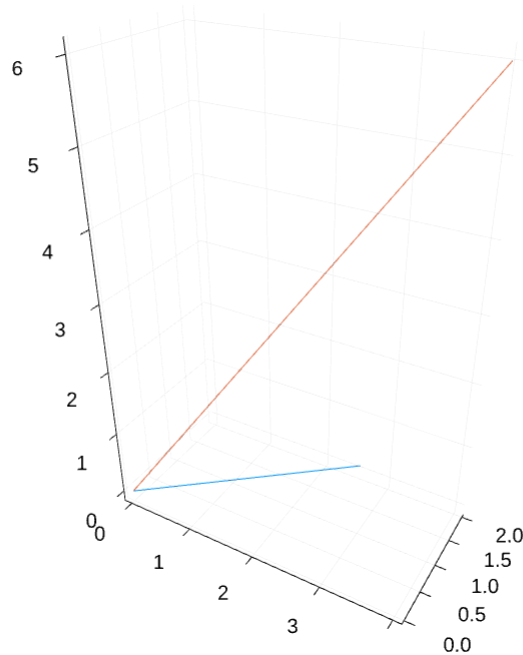
```
 3  4
 1  2
 1  6
```

```
1 F = [
2     3 4;
3     1 2;
4     1 6
5 ]
```

```
3×2 Matrix{Float64}:
```

```
1.0  0.0
0.0  1.0
0.0  0.0
```

```
1 rref(F)
```



```
1 LA.graph_vectors(F)
```

```
2
```

```
1 rank(F)
```

```
G = 2×3 Matrix{Int64}:
```

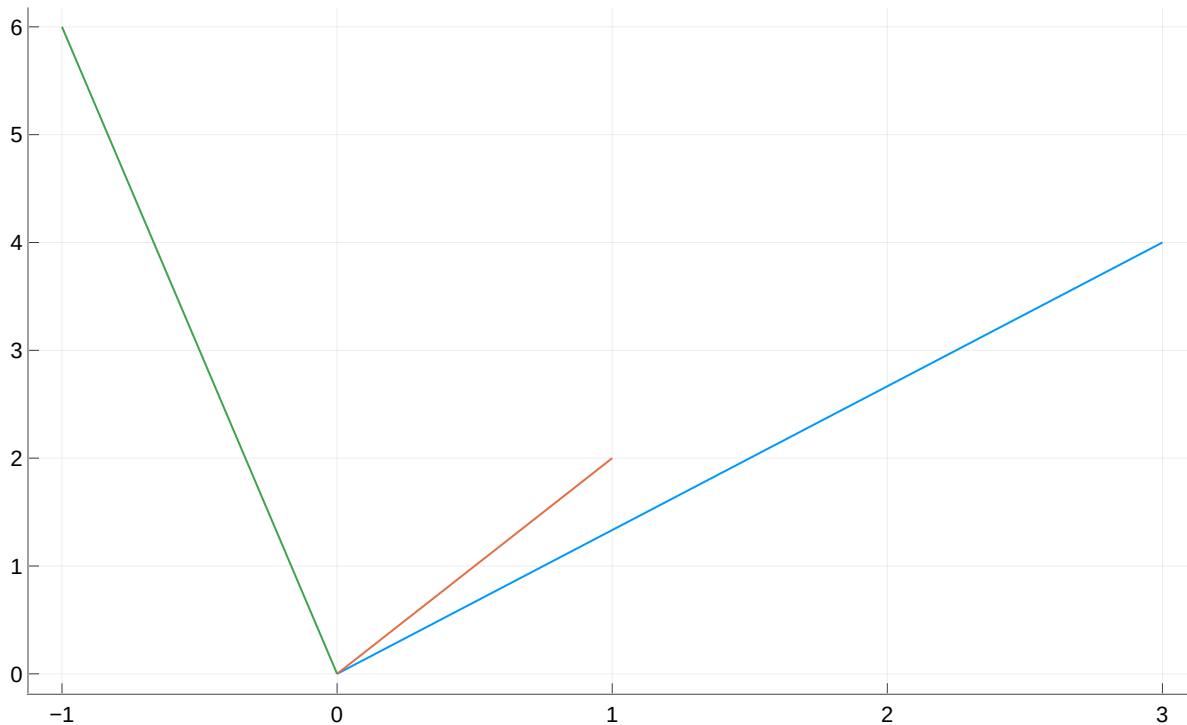
```
 3  1 -1
 4  2  6
```

```
1 G = [
2     3 1 -1;
3     4 2  6
4 ]
```



```
2×3 Matrix{Float64}:  
 1.0  0.0 -4.0  
 0.0  1.0 11.0
```

```
1 rref(G)
```



```
1 LA.graph_vectors(G)
```

```
2
```

```
1 rank(G)
```

Determinants using row operations

```
1 md"# Determinants using row operations"
```

1. If a multiple of row A is added to another row to make matrix B:

$$\det(A) = \det(B)$$

$$R_2 = 3R_3 + R_2$$

2. If the matrix B is formed by interchanging rows of A:

$$\det(A) = -\det(B)$$

$$R_1 \leftrightarrow R_2$$

3. If matrix B is formed by multiplying a row in matrix A by a number k:

$$\det(A) = (1/k)\det(B)$$

$$R_3 = (1/2)R_3$$

GOAL:

Get matrix to triangular form, then det of matrix A is equal to:

$$\det(A) = a_{11} * a_{22} * a_{33}$$

true

```
1 begin
2     I = [
3         1 2 1;
4         0 2 1;
5         0 0 4
6     ]
7
8     # in triangular form
9     # multiply diagonals
10    1 * 2 * 4 == det(I)
11 end
```

true

```
1 begin
2     J = [
3         2 3;
4         0 9
5     ]
6
7     2 * 9 == det(J)
8 end
```

8

```
1 prod(diag(UpperTriangular(I)))
```

18

```
1 prod(diag(UpperTriangular(J)))
```

Cofactors

There is a cofactor located at every position in a matrix

$$C_{ij} = (-1)^{i+j} \det(A_{ij})$$

(A_{ij} is submatrix (row i and column j crossed out))

Finding det with cofactors

about i th row:

$$\det(A) = a_{i1}C_{i1} + a_{i2}C_{i2} + \dots$$

about j th col:

$$\det(A) = a_{1j}C_{1j} + a_{2j}C_{2j} + \dots$$

```
1 md"""
2 about ith row:
3
4 $det(A) = a_{i1}C_{i1} + a_{i2}C_{i2} + \dots$
5
6 about jth col:
7
8 $det(A) = a_{1j}C_{1j} + a_{2j}C_{2j} + \dots$
9 """
```

Adjoint and Inverses

```
K = 3×3 Matrix{Int64}:
  -1  1  2
   0  6  3
   4  7  5
```

```
1 K = [
2     -1 1 2;
3     0 6 3;
4     4 7 5
5  ]
```

```
3×3 Matrix{Float64}:
 9.0  12.0 -24.0
 9.0 -13.0  11.0
-9.0   3.0  -6.0
```

```
1 LA.cofactor(K)
```

```
adj = 3×3 transpose(::Matrix{Float64}) with eltype Float64:
 9.0  9.0 -9.0
12.0 -13.0 3.0
-24.0 11.0 -6.0
```

```
1 adj = transpose(LA.cofactor(K))
```

```
3×3 transpose(::Matrix{Float64}) with eltype Float64:
 9.0  9.0 -9.0
12.0 -13.0 3.0
-24.0 11.0 -6.0
```

```
1 LA.adjugate(K)
```

```
true
```

```
1 LA.adjugate(K) == adj
2
```

Properties of adjoints

$$A * \text{adjugate}(A) = \det(A) * I$$

```
3×3 Matrix{Float64}:
-45.0  0.0  0.0
 0.0 -45.0  0.0
 0.0  0.0 -45.0
```

```
1 K * LA.adjugate(K)
```

$$\text{inv}(A) = (1/\det(A)) * \text{adjugate}(A)$$

```
inv_ = 3×3 Matrix{Float64}:
  -0.2      -0.2      0.2
 -0.266667  0.288889 -0.0666667
  0.533333 -0.244444  0.133333
```

```
1 inv_ = 1/det(K) * LA.adjugate(K)
```

```
3×3 Matrix{Float64}:
 -0.2      -0.2      0.2
 -0.266667  0.288889 -0.0666667
  0.533333 -0.244444  0.133333
```

```
1 inv(K)
```

```
3×3 Matrix{Rational{Int64}}:
 -1//5  -1//5  1//5
 -4//15 13//45 -1//15
  8//15 -11//45 2//15
```

```
1 inv(Rational.(K))
```

```
3×3 transpose(::Matrix{Float64}) with eltype Float64:
  9.0  9.0 -9.0
 12.0 -13.0 3.0
-24.0 11.0 -6.0
```

```
1 LA.adjugate(K)
```

As long as $\det(M) \neq 0$, there will be an $\text{inv}(M)$

```
1 md"### As long as det(M) != 0, there will be an inv(M)"
```

If $\det(M) = 0$, then there may be an $\text{inv}(M)$, but you have to be lucky. Solution must live on that line.

```
1 md"### If det(M) == 0, then there may be an inv(M), but you have to be lucky.
  Solution must live on that line."
```

Properties of Determinants

If any row or col of a matrix is all zeros, then the determinant is also zero

```
L = 3×3 Matrix{Int64}:
 1  2  3
 0  0  0
 4  5  6
```

```
1 L = [
2     1 2 3;
3     0 0 0;
4     4 5 6
5 ]
```

0.0

```
1 det(L)
```

If any two rows or cols are proportional to each other, then the determinant is also zero

```
1 md"""### If any two rows or cols are proportional to each other, then the determinant
   is also zero"
```

```
M = 2×2 Matrix{Int64}:
 1  2
 2  4
```

```
1 M = [
2     1 2;
3     2 4
4 ]
```

0.0

```
1 det(M)
```

false

```
1 LA.is_LI(M)
```

```
2×2 Matrix{Float64}:
 1.0  2.0
 0.0  0.0
```

```
1 rref(M)
```

0.0

```
1 det(rref(M))
```

If row/col is linear combo of other rows/cols, then determinant is zero

Det of transpose of matrix is equal to the determinant of the matrix

$$\det(A^T) = \det(A)$$

The determinant of two matrices multiplied together is equal to the $\det(A) * \det(B)$

$$\det(AB) = \det(A) * \det(B)$$

Homogeneous System of Equations (square matrices)

$$AX = 0$$

Solution is non-trivial iff $\det(A) == 0$

If $\det(A) \neq 0$, then $X = 0$ is only solution (trivial case)

Linear Independence and Basis

- Solution to homogeneous equation's free variables' basis vectors are linearly independent

Vectors are LI if:

$$c_1u_1 + c_2u_2 + c_3u_3 + \dots + c_ku_k = 0$$

only if

$$c_1, c_2, c_3, \dots, c_k$$

are all zero.

Not LI example: $u_1 = 3u_2 + 2u_3$

Test:

$$\langle c_1 u_1, c_2 u_2, \dots, c_k u_k \rangle$$

as matrix A.

If $\det(A) \neq 0$, then LI.

$$u_1 = \langle 1, 0, 0 \rangle, u_2 = \langle 0, 1, 0 \rangle, u_3 = \langle 1, 1, 0 \rangle$$

```
1 md"$u_1 = \langle 1, 0, 0 \rangle, u_2 = \langle 0, 1, 0 \rangle, u_3 = \langle 1, 1, 0 \rangle"
```

But $u_3 = u_1 + u_2 \Rightarrow$ Not LI

```
1 md"But $u_3 = u_1 + u_2$ => Not LI"
```

0 = 3×3 Matrix{Int64}:

```
1  0  1
2  0  1  1
3  0  0  0
```

```
1 0 = [
2    1 0 1;
3    0 1 1;
4    0 0 0
5  ]
```

0.0

```
1 det(0)
```

false

```
1 LA.is_LI(0)
```

$$u_3 = u_1 + u_2$$

New vectors:

$$u_1 = \langle 1, 1, -1 \rangle, u_2 = \langle 0, 1, 1 \rangle, u_3 = \langle 1, 2, 3 \rangle$$


```
P = 3×3 Matrix{Int64}:
```

```
 1  0  1
 1  1  2
-1  1  3
```

```
1 P = [
2     1 0 1;
3     1 1 2;
4     -1 1 3
5 ]
```

```
3.0
```

```
1 det(P)
```

```
true
```

```
1 LA.is_LI(P)
```

Eigenvalues, Eigenvectors

```
1 md"# Eigenvalues, Eigenvectors"
```

The Eigenvalue Problem

```
1 md"## The Eigenvalue Problem"
```

$$Ax = \lambda x$$

λ = eigenvalue (scalar)

x = eigenvector

For every eigenvalue λ there is a set of nontrivial vectors x called the eigenvectors

```
2×1 Matrix{Int64}:
-24
 20
```

```
1 begin
2     Q = [
3         1 6;
4         5 2
5     ]
6
7     x_1 = [6 -5]
8
9     # Is x an eigenvector?
10
11     #Ax = λx
12
13     result_eig = Q * x_1
14 end
```

```
2×1 Matrix{Float64}:
-4.0
-4.0
```

```
1 result_eig ./ x_1
```

$$\lambda = -4$$

```
1 md"\lambda = -4"
```

```
[-4.0, 7.0]
```

```
1 eigvals(Q)
```

```
2×2 Matrix{Float64}:
-0.768221 -0.707107
 0.640184 -0.707107
```

```
1 eigvecs(Q)
```

```
val = 2×1 Matrix{Float64}:
-7.810249675906654
-7.810249675906655
```

```
1 val = [6 -5]' ./ eigvecs(Q)[:, 1]
```

Example 2

```
2×1 Matrix{Int64}:
-9
11
```

```
1 begin
2     R = [
3         1 6;
4         5 2
5     ]
6
7     x_R = [3 -2]
8
9     result_R = R * x_R
10 end
```

```
2×1 Matrix{Float64}:
-3.0
-5.5
```

```
1 result_R ./ x_R
```

NOT an eigenvector!

```
1 md"NOT an eigenvector!"
```

```
[-4.0, 7.0]
```

```
1 eigvals(R)
```

```
2×2 Matrix{Float64}:
-0.768221 -0.707107
 0.640184 -0.707107
```

```
1 eigvecs(R)
```

Example 3: 3x3 matrix

```
1 md"Example 3: 3x3 matrix"
```

```
3×1 Matrix{Int64}:
3
5
7
```

```
1 begin
2     S = [
3         2 2 1;
4         1 3 1;
5         1 2 2
6     ]
7
8     x_S = [-1 1 3]
9
10    result_S = S * x_S
11 end
```

```
3×1 Matrix{Float64}:
-3.0
 5.0
2.3333333333333335
```

```
1 result_S ./x_S
```

NOT an eigenvector!

```
1 md"NOT an eigenvector!"
```

```
3×1 Matrix{Int64}:
 5
 5
 5
```

```
1 begin
2     T = [
3         2 2 1;
4         1 3 1;
5         1 2 2
6     ]
7
8     x_T = [1 1 1]
9
10    result_T = T * x_T
11 end
```

```
3×1 Matrix{Float64}:
 5.0
 5.0
 5.0
```

```
1 result_T ./ x_T
```

$$\lambda = 5$$

Find Eigenvalues and eigenvectors from scratch

Find eigenvalue

$$Ax = \lambda x$$

$$Ax = \lambda Ix$$

$$Ax - \lambda Ix = 0$$

$$(A - \lambda I)x = 0$$

This equation has nontrivial solns when $\det(A - \lambda I) = 0$

Example 1

```
1 md"Example 1"
```

```
U = 2x2 Matrix{Int64}:
 1  2
 2  1
```

```
1 U = [
2     1 2;
3     2 1
4     ]
```

```
char_U = #1 (generic function with 1 method)
```

```
1 char_U = λ -> (1 - λ)*(1 - λ) - (2)*(2)
```

```
[-1.0, 3.0]
```

```
1 find_zeros(char_U, -100, 100)
```

$A - \lambda I$ is the characteristic matrix

```
1 md"### $A-\lambda I$ is the characteristic matrix"
```

```
[-1.0, 3.0]
```

```
1 eigvals(U)
```

Example 2

```
1 md"Example 2"
```

```
V = 2x2 Matrix{Int64}:
 2  7
 7  2
```

```
1 V = [
2     2 7;
3     7 2
4     ]
```

```
char_V = #3 (generic function with 1 method)
```

```
1 char_V = λ -> (2 - λ)*(2 - λ) - (7)(7)
```

```
[-5.0, 9.0]
```

```
1 find_zeros(char_V, -100, 100)
```

```
[-5.0, 9.0]
```

```
1 eigvals(V)
```

Example 3

```
1 md"Example 3"
```

```
W = 2x2 Matrix{Int64}:
 3  -2
 1  -1
```

```
1 W = [
2     3 -2;
3     1 -1
4 ]
```

```
char = #5 (generic function with 1 method)
```

```
1 char = λ -> (3 - λ)*(-1 - λ) - (-2)
```

```
[-0.414214, 2.41421]
```

```
1 find_zeros(char, -100, 100)
```

```
[-0.414214, 2.41421]
```

```
1 eigvals(W)
```

$$\frac{-b \pm \sqrt{b^2 - 4ac}}{2a}$$

```
1 md"\frac{-b \pm \sqrt{b^2 - 4ac}}{2a}"
```

```
2.414213562373095
```

```
1 1 + sqrt(2)
```

```
-0.41421356237309515
```

```
1 1 - sqrt(2)
```

Example 4

```
1 md"Example 4"
```

```
Y = 3×3 Matrix{Int64}:
```

```
 5  4  4
-7 -3 -1
 7  4  2
```

```
1 Y = [
2     5 4 4;
3     -7 -3 -1;
4     7 4 2
5 ]
```

```
char_Y = #7 (generic function with 1 method)
```

```
1 char_Y = λ -> (5 - λ)*(λ^2 + λ - 2) - 4*(7*λ - 7) + 4*(7*λ - 7)
```

```
[-2.0, 1.0, 5.0]
```

```
1 find_zeros(char_Y, -100, 100)
```

```
[-2.0, 1.0, 5.0]
```

```
1 eigvals(Y)
```

Example 5

```
1 md"Example 5"
```

```
Z = 4×4 Matrix{Int64}:
```

```
 5 -2  6 -1
 0  3 -8  0
 0  0  5  4
 0  0  0  1
```

```
1 Z = [
2     5 -2 6 -1;
3     0 3 -8 0;
4     0 0 5 4;
5     0 0 0 1
6 ]
```

```
[1.0, 3.0, 5.0, 5.0]
```

```
1 eigvals(Z)
```

```
1 # Upper Triangular Matrix! Use shortcut of diagonal prods
```

```
char_Z = #9 (generic function with 1 method)
```

```
1 char_Z = λ -> (5 - λ)*(3 - λ)*(5 - λ)*(1-λ)
```

```
[1.0, 3.0, 5.0]
```

```
1 find_zeros(char_Z, -100, 100)
```

Finding eigenvectors

```
2×2 Matrix{Int64}:

```

```
1  2
2  1
```

```
1  U
```

```
[-1.0, 3.0]
```

```
1  eigvals(U)
```

```
2×2 Matrix{Float64}:

```

```
-0.707107  0.707107
 0.707107  0.707107
```

```
1  eigvecs(U)
```

```
2×2 Matrix{Int64}:

```

```
2  2
2  2
```

```
1  begin
2      λ_U = -1
3      matrix_U = U - λ_U * LA.identity(2)
4  end
```

Dependent system! Infinite solutions! [0 0]

```
1. error(::String) @ error.jl:35
2. aug_solve(::Matrix{Int64}, ::Float64) @ LA.jl:56
3. aug_solve(::Matrix{Int64}) @ LA.jl:33
4. top-level scope @ [Local: 1 [inlined]
```

```
1  LA.aug_solve([matrix_U [0 0]'])
```

```
2×2 Matrix{Float64}:

```

```
1.0  1.0
0.0  0.0
```

```
1  rref(matrix_U)
```

for lambda = -1

$$X = \langle -k, k \rangle$$

or

$$X = k \langle -1, 1 \rangle$$

```
1  md"""$X = \langle -k, k \rangle$
2
3  or
4
5  $X = k \langle -1, 1 \rangle$"""
```

Eigenvector: for lambda = -1: k <-1, 1>


```
2×2 Matrix{Int64}:
```

```
-2  2
 2 -2
```

```
1 begin
2     λ_U_3 = 3
3     matrix_U_3 = U - λ_U_3 * LA.identity(2)
4 end
```

```
2×2 Matrix{Float64}:
```

```
1.0 -1.0
0.0  0.0
```

```
1 rref(matrix_U_3)
```

$$x_2 = k$$

$$x_1 - k = 0, x_1 = k$$

$$X = \langle k, k \rangle = k \langle 1, 1 \rangle$$

eigenbasis: [-1, 1], [1, 1]

true

```
1 begin
2     eig_neg1 = [-1 1]'
3     eig_3 = [1 1]'
4     LA.is_LI([eig_neg1 eig_3])
5 end
```

3x3 example

eig_result (generic function with 1 method)

```
1 function eig_result(M::AbstractMatrix, lambda::Real)
2     return M - lambda * LA.identity(size(M, 1))
3 end
```

```
3×3 Matrix{Int64}:
```

```
5  4  4
-7 -3 -1
7  4  2
```

```
1 Y
```

```
result_Y = 3×3 Matrix{Int64}:
  0  4  4
 -7 -8 -1
  7  4 -3
```

```
1 result_Y = eig_result(Y, 5)
```

eig_solve (generic function with 1 method)

```
1 function eig_solve(M::AbstractMatrix)
2     return rref([M [0 for i=1:Integer(size(M, 1))]])
3 end
```

```
3×4 Matrix{Float64}:
 1.0  0.0 -1.0 -0.0
 0.0  1.0  1.0  0.0
 0.0  0.0  0.0  0.0
```

```
1 eig_solve(result_Y)
```

for lambda = 5, $X = [k, -k, k] = k[1, -1, 1]$ for $(x_3 = k)$

```
3×4 Matrix{Float64}:
 1.0  0.0 -1.0 -0.0
 0.0  1.0  2.0  0.0
 0.0  0.0  0.0  0.0
```

```
1 eig_solve(eig_result(Y, 1))
```

for lambda = 1, $X = k[1, -2, 1]$ for $x_3 = k$

```
3×4 Matrix{Float64}:
 1.0  0.0  0.0  0.0
 0.0  1.0  1.0  0.0
 0.0  0.0  0.0  0.0
```

```
1 eig_solve(eig_result(Y, -2))
```

for lambda = -2, $X = k[0, -1, 1]$ for $x_3 = k$

Basis Vectors: $\langle 1, -1, 1 \rangle, \langle 1, -2, 1 \rangle, \langle 0, -1, 1 \rangle$

bluebrown_eigvals (generic function with 1 method)

```
1 function bluebrown_eigvals(M::AbstractMatrix)
2     # Must be 2x2 matrix
3     if Integer(size(M, 1)) != 2 && Integer(size(M, 2)) != 2
4         throw("Not 2x2 Matrix")
5     end
6     # trace: sum of diags
7     m = tr(M)/2
8     p = det(M)
9     square = sqrt(m^2 - p)
10    return [m + square, m - square]
11 end
12
13
```

[7.0, -4.0]

1 `bluebrown_eigvals(Q)`

[7.0, -4.0]

1 `bluebrown_eigvals(R)`

[3.0, -1.0]

1 `bluebrown_eigvals(U)`

[9.0, -5.0]

1 `bluebrown_eigvals(V)`

[2.41421, -0.414214]

1 `bluebrown_eigvals(W)`