# 1.1

# 1.1 Matrices over Finite Fields

# Division

### Question 1

The program to find the inverse of each element of GF(p) is shown on page 7, labeled

### inverses(p)

This function takes the parameter p which is assumed to be a positive integer, the function returns an array of inverses, where the  $n^{th}$  entry represents the inverse of n in GF(p).

Testing the programs functionality

```
>> print(inverses(2))
[1]
>> print(inverses(5))
[1, 3, 2, 4]
>> print(inverses(7))
[1, 4, 5, 2, 3, 6]
```

Note if that if p is a positive composite integer then the program will have 0s where inverses do not exist

```
>> print(inverses(8))
[1, 0, 3, 0, 5, 0, 7]
```

If we assume that p is prime then we know that  $\forall a \in GF(p) \setminus \{0\}$  there exists an  $a^{-1}$  such that  $aa^{-1} \equiv 1 \mod p$  so  $(p-a)(p-a^{-1}) \equiv aa^{-1} \equiv 1 \mod p$ . Therefore, if L is our array of inverses (and assume indexing starts at 1) then  $L[a] = a^{-1}$  and  $L[p-a] = p-a^{-1}$ , so only need to calculate inverses for the first half of the list.

#### Question 2

For each element  $a \in GF(p) \setminus \{0\}$  the following operations must occur

- 1.  $a^{-1}$  comparisons must be made to find  $a^{-1}$
- 2. assignment of  $a^{-1}$  in the array

The assignment of an element into an array is of order  $\Theta(1)$  so doing this for each  $a \in GF(p) \setminus \{0\}$  means step 2 has order  $\Theta(p)$  over the whole algorithm.

We know that each comparison has complexity  $\Theta(1)$ , and the total number of comparisons is  $\sum_{a \in GF(p)\setminus\{0\}} a^{-1} = \sum_{n=1}^{p-1} n = \frac{(p-1)p}{2} = \Theta(p^2)$ , so step 1 has complexity  $\Theta(p^2)$  over the whole algorithm. Adding both together, we get the complexity of this algorithm is  $\Theta(p^2)$ 

### Gaussian Elimination

### Question 3

The function to perform Gaussian elimination in GF(p) is shown on page 7, labelled

```
gauss_elim(M, mod)
```

Where M is the matrix is to be Gaussian eliminated and mod is the size of the finite field we are working in.

So the matrix  $A_1$  has rank 4 in GF(11) and the row space has basis

$$(1,0,3,0,0),(0,1,7,0,0),(0,0,0,1,0),(0,0,0,0,1)$$

```
>> print(gauss_elim(A1,19))
[[ 1 0 0 0 13]
  [ 0 1 0 0 6]
  [ 0 0 1 0 3]
  [ 0 0 0 1 1]]
```

So the matrix  $A_1$  has rank 4 in GF(19) and the row space has basis

```
(1,0,0,0,13),(0,1,0,0,6),(0,0,1,0,3),(0,0,0,1,1)
>> A2 = np.array([[6,16,11,14,1,4],
                   [7,9,1,1,21,0],
                   [8,2,9,12,17,7],
                   [2,19,2,19,7,12]])
>> print(gauss_elim(A1,19))
[[ 1 0 0 9 11
                  9]
 0 ]
      1
         0 10 5
                  5]
 [ 0
      0
         1
            9 14
                  7]
 ΓΟ
           0
      0
         0
               0
                  0]]
```

So the matrix  $A_2$  has rank 3 in GF(23) and the row space has basis

$$(1,0,0,9,11,9),(0,1,0,10,5,5),(0,0,1,9,14,7)$$

# Kernels and Annihilators

### Question 4

Assume A an  $m \times n$  matrix is in row echelon form and let  $L = \{l(1), \ldots, l(r)\}$  where r is the rank of the matrix, then for a given row i,  $A_{ij} = 0$  for j < l(i) and  $A_{ij} = 0$  for j = l(k) such that  $i < k \le r$ . This implies that  $A_{ij} = 0$  for all  $j \in L \setminus \{l(i)\}$  So to solve the equation

$$A\mathbf{x} = \mathbf{0}$$

Then we require

$$[A\mathbf{x}]_i = 0 \ \forall i \iff \sum_{j=1}^n A_{ij}x_j = 0 \ \forall i \iff x_{l(i)} = -\sum_{GF(p) \setminus L} A_{ij}x_j \ \forall i$$

So  $\mathbf{x}$  is uniquely determined by  $x_j$  such that  $j \in GF(p) \setminus L$  So to find a basis of ker A, iterate through basis  $\{\mathbf{x}'_1, \dots, \mathbf{x}'_{n-r}\}$  of the vector space  $\{\mathbf{x} \in GF(p)^n : x_j = 0 \ \forall j \in L\}$  and use  $[\mathbf{x}]_{l(i)} = -[A\mathbf{x}']_i$  to find a basis of ker A

The function to perform this algorithm is shown on pages 8 and 9, labeled

```
kernel_basis(M, mod)
```

Where M is the matrix for which the kernel is to be found and mod is the size of the finite field we are working in.

```
\Rightarrow B1 = np.array([[4,6,5,2,3],
                    [5,0,3,0,1],
                    [1,5,7,1,0],
                    [5,5,0,3,1],
                    [2,1,2,4,0]])
>> print(kernel_basis(B1, 13))
[array([[7],
        [2],
        [1],
        [2],
        [1]])]
>> print(kernel_basis(B1, 17))
[array([[0],
        [0],
        [0],
        [0],
        [0]])]
\Rightarrow B2 = np.array([[3,7,19,3,9,6],
                    [10,2,20,15,3,0]
                    [14,1,3,14,11,3],
                    [26,1,21,6,3,5],
                    [0,1,3,19,0,3]
>> print(kernel_basis(B2, 23))
[array([[6],
        [6],
        [9],
        [9],
```

[9], [1]])]

### Question 5

If U is a subspace of  $F^n$  then  $\dim U + \dim U^{\circ} = \dim F^n = n$ 

### Question 6

For this I defined a function to find a basis of  $U^{\circ}$  as shown on page 9, labeled

# row\_annihilator(U, mod)

Where U is a matrix where the rows are the row basis of a subspace U and mod is the size of the finite field we are working in. we also define a function  $col_annihilator(U, mod)$  similarly for column spaces

```
>> U_annihilate = row_annihilator(A1, 19)
>> U_annihilate_annihilate = col_annihilator(U_annihilate, 19)
>> print(U_annihilate)
[[ 6]
 [13]
 [16]
 [18]
 [ 1]]
>> print(U_annihilate_annihilate)
     1
        0 0 0]
 [10
     0
        1 0 0]
 [16
     0
        0 1 0]
 [ 3
        0 0 1]]
     0
```

Now from Question 4, we know that in GF(19) U has basis

$$(1,0,0,0,13),(0,1,0,0,6),(0,0,1,0,3),(0,0,0,1,1)$$

and from above we know that in GF(19)  $(U^{\circ})^{\circ}$  has basis

$$(1, 1, 0, 0, 0), (10, 0, 1, 0, 0), (16, 0, 0, 1, 0), (3, 0, 0, 0, 1)$$

Now note that in the field GF(19)

$$(1,0,0,0,13) = 13*(3,0,0,0,1)$$

$$(0,1,0,0,6) = 6*(3,0,0,0,1) + (1,1,0,0,0)$$

$$(0,0,1,0,3) = 3*(3,0,0,0,1) + (10,0,1,0,0)$$

$$(0,0,0,1,1) = (3,0,0,0,1) + (16,0,0,1,0)$$

So the 2 bases are equivalent, therefore  $U=(U^{\circ})^{\circ}$ 

# Question 7

Now if an  $m_1 \times n$  matrix A has row space U and an  $m_2 \times n$  matrix B has row space W, then the row space defined by concatenating matrices A and B, i.e by the matrix

```
\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ a_{m_11} & a_{m_12} & \dots & a_{m_1n} \\ b_{11} & b_{12} & \dots & b_{1n} \\ \vdots & \vdots & \ddots & \vdots \\ b_{m_21} & b_{m_22} & \dots & b_{m_2n} \end{pmatrix}
```

Is equivalent to U+W, so if we Gaussian eliminate the above matrix, we get a basis of the row space of U+W. We can use the fact that  $U\cap W=(U^\circ)^\circ\cap (W^\circ)^\circ=(U^\circ+W^\circ)^\circ$  to find a basis of  $U\cap W$  using only sum and annihilation of vector sub-spaces.

To find the basis of U+W and  $U\cap W$  I defined the functions vector\_row\_space\_sum(U, W, mod) and vector\_col\_space\_sum(U, W, mod) which given matrices U and W finds the basis of the respective row and column space. I also defined the function bases(U, W, mod) to quickly find and return the bases of U, W, U+W,  $U\cap W$ 

```
>> U1, W1, U1pW1, U1aW1 = bases(A1, B1, 11)
>> print(U1)
[[1 0 3 0 0]
 [0 1 7 0 0]
 [0 0 0 1 0]
 [0 0 0 0 1]]
>> print(W1)
[[1 0 0 2 0]
 [0 1 0 3 0]
 [0 0 1 4 0]
 [0 0 0 0 1]]
>> print(U1pW1)
[[1 0 0 0 0]
 [0 1 0 0 0]
 [0 0 1 0 0]
 [0 0 0 1 0]
 [0 0 0 0 1]]
>> print(U1aW1)
[[7 5 1 0 0]
 [8 6 0 1 0]
 [0 0 0 0 1]]
\Rightarrow A3 = np.array([[1,0,0,0,3,0,0],
                   [0,5,0,1,6,3,0],
                   [0,0,5,0,2,0,0],
                   [2,4,0,0,0,5,1],
                   [4,3,0,0,6,2,6]])
>> B3 = np.transpose(row_annihilator(A3, 19))
>> U2, W2, U2pW2, U2aW2 = bases(A3, B3, 19)
>> print(U2)
[[ 1
     0 0 0
               0 6 1]
 [ 0
         0 0
      1
              0 3 14]
 0 0
        1 0 0 16 9]
 ΓΟ
           1
               0 0
                     81
     0
        0 0
              1 17 6]]
>> print(W2)
```

```
[[ 1 0 9 18 6
                 1 12]
 [ 0
           2
     1
        0
                 4 14]]
>> print(U2pW2)
[[1 0 0 0 0 0 0]
 [0 1 0 0 0 0 0]
 [0 0 1 0 0 0 0]
 [0 0 0 1 0 0 0]
 [0 0 0 0 1 0 0]
 [0 0 0 0 0 1 0]
 [0 0 0 0 0 0 1]]
>> print(U2aW2)
[[0 0 0 0 0 0 0]]
>> B4 = np.transpose(row_annihilator(A3, 23))
>> U3, W3, U3pW3, U3aW3 = bases(A3, B4, 23)
>> print(U3)
[[ 1
     0
         0
               0
                  8 22]
 [ 0
     1
           0
              0 3 18]
         0
 [ 0
         1
            0
               0 21
                     6]
 [ 0
     0
         0
           1
               0
                 4
                     0]
 [ 0
     0
        0
            0
              1 5
                     8]]
>> print(W3)
[[ 1 0 17 8 15 21 8]
 [0 1 0 3 0 5 17]]
>> print(U3pW3)
[[ 1
     0
         0
           0
              0
                  0 12]
 [ 0
                  0 20]
            0
 [ 0
                  0 20]
     0
        1
           0
               0
 0 ]
     0
         0
            1
               0
                  0 18]
 [ 0
     0
        0 0
              1 0 19]
 [ 0
     0
        0
           0
               0
                 1
                    7]]
>> print(U3aW3)
[[11 3 3 5 4 16
\dim U + \dim W = \dim(U + W) + \dim(U \cap W)
```

### Question 8

In the reals for the case where U is the row space of  $A_3$  and  $W = \{x^T | x \in \ker A_3\}$  we would find that  $\dim(U+W) = \dim(\mathbb{R}^7) = 7$  which implies  $\dim(U\cap W) = 0$  but in the very last example, we saw that in GF(23),  $\dim(U+W) = 6$  and  $\dim(U\cap W) = 1$ .

# **Programs**

### Division

```
def inverses(p):
    inv = []
    for a in range(1,p):
        a_{inv} = 1
        for a_inv in range(1,p):
            if a*a_inv % p == 1:
                inv.append(a_inv)
                break
        else:
            inv.append(0)
    return inv
Gaussian Elimination
def transpose(M, i, j ,mod):
    # swap rows i and j
    M = M.copy()
    y = M[i].copy()
    M[i] = M[j].copy()
    M[j] = y
    return M
def multiply(M, i, a, mod):
    # multiply row i by a
    M = M.copy()
    M[i] = a*M[i] % mod
    return M
def subtract(M, i, j, a, mod):
    # subtract a * row j from row i
    M = M.copy()
    M[i] = (M[i] - a*M[j]) \% mod
    return M
def gauss_elim(M, mod):
    M = M.copy()
    M = M \% mod
    rows = M.shape[0]
    cols = M.shape[1]
    inv = inverses(mod)
    cur_col = 0
    cur_row = 0
    while cur_col < cols and cur_row < rows:
        if M[cur_row][cur_col] == 0:
            for row in range(cur_row, rows):
                if M[row][cur_col]!=0:
                    M=transpose(M, row, cur_col, mod)
                    break
            else:
                cur_col += 1
                continue
```

```
lead_coef = M[cur_row][cur_col]
M = multiply(M, cur_row, inv[lead_coef-1], mod)
for row in range(0, rows):
    if row == cur_row:
        continue
    row_lead_coef = M[row][cur_col]
    M = subtract(M, row, cur_row, row_lead_coef, mod)
    cur_row+=1
    cur_col+=1
return M
```

### Kernels and Annihilators

```
def rank(M, mod):
    M = gauss_elim(M,mod)
    r=0
    rows = M.shape[0]
    cols = M.shape[1]
    cur_col = 0
    cur_row = 0
    while cur_col < cols and cur_row < rows:
        if M[cur_row][cur_col] > 0:
            r+=1
            cur_row+=1
        else:
            cur_col+=1
    return r
def undetermined_x(M, mod):
   M = gauss_elim(M,mod)
    rows = M.shape[0]
    cols = M.shape[1]
    cur_col = 0
    cur\_row = 0
    undet_x = []
    while cur_col < cols and cur_row < rows:
        if M[cur_row][cur_col] > 0:
            cur_row+=1
            cur_col+=1
        else:
            undet_x.append(cur_col)
            cur_col+=1
    if cur_col != cols:
        undet_x = undet_x + [x for x in range(cur_col, cols)]
    return undet_x
def kernel_basis(M, mod):
    M = gauss_elim(M ,mod)
    r = rank(M, mod)
    cols = M.shape[1]
    undet_x = undetermined_x(M, mod)
    basis = []
    for i in undet_x:
        x = np.zeros((cols,1), dtype=int)
```

```
x[i][0]=1
        j=cols-1
        cur\_row = r-1
        determined_x = np.matmul(M, x)
        while j>=0:
            if j not in undet_x:
                x[j] = -determined_x[cur_row] % mod
                cur_row -= 1
            j-=1
        basis.append(x)
    if len(basis) == 0:
        basis.append(np.zeros((cols,1), dtype=int))
    return basis
def get_rid_of_empty(U):
    rows = U.shape[0]
    cols = U.shape[1]
    non_empty = []
    for i in range(rows):
        for j in range(cols):
            if U[i][j]!=0:
                non_empty.append(i)
                break
    U=U[non_empty]
    return U
def vector_row_space_sum(U, W, mod):
    \# U and W given in the form of a matrix where U has row space U
    # and W has row space W
    U = gauss_elim(U,mod)
    W = gauss_elim(W,mod)
    full_space = np.concatenate((U, W))
    full_space = gauss_elim(full_space,mod)
    full_space=get_rid_of_empty(full_space)
    return full_space
def vector_col_space_sum(U, W, mod):
    row_U = np.transpose(U)
    row_W = np.transpose(W)
    row_full_space = vector_row_space_sum(row_U, row_W, mod)
    full_space = np.transpose(row_full_space)
    return full_space
def row_annihilator(U, mod):
    # U given in the form of a matrix with row space U
    basis = kernel_basis(U, mod)
    matrix_basis = np.concatenate(basis, axis=1)
    return matrix_basis
def col_annihilator(U, mod):
    return np.transpose(row_annihilator(np.transpose(U),mod))
def bases(U, W, mod):
```

return U, W, UpW, UaW