```
Matrix: A = [12;34;56]
Vector: V= [123], V= [1;2;3]
size (A) = [r c] , length(v)
                                 Size (A, 1) : r
                                   Size (A, 2) : C
A= ones(r,c)
hist (-) // histogram
eye(n) // n = n identity matrix
 load ('file.dat')
 who(s) // shows variables
  clear ___
  save file.abc x, -ascii //text
   : ever element along row/column
     A(r,:), A(:,c), A(r,c)
     A(:) all elements into column vector
```

Computational Operations

 $A \times B$ ,  $A \times B$ matrix multiplication element wize operation max(), rand A' (transpose) Sum (a), prod (a), floor (a), ceil (a) pinv(A) // inverse
max(A,[], 1) column

```
Plotting Data Rlabel(), ylabel(), legend()

plot(x,y) print - dpng "file"

hold on; axis([x, x, y, y, ])

clf; subplot(r,c, element #)

imagesc(A), colorbar, colormap gray
```

```
Control Statements
```

$$V=zeros(10,1)$$
  $\longleftrightarrow$  indices= 1:10:  
for i=start:end, for i=indices  
 $v(i)=2^{4}i$ ;  
end;

while 
$$i = 1$$

while  $i = 1$ 

while  $i = 1$ 
 $i = 1 + 1$ 

functions : file.m

- function y = square Number(x) $y = x^2$ :
- function  $[y_1, y_2] = square And Cube Number(x)$   $y_1 = x^{\lambda};$   $y_2 = x^{\lambda};$   $\begin{bmatrix} x & 0 \\ 0 & 1 \end{bmatrix} \times \begin{bmatrix} 0 \\ 0 & 1 \end{bmatrix}$

Cost function: 
$$x_0$$
  $x_1$   $x_2$   $x_3$   $x_4$   $x_5$   $x_6$   $x$ 

function T = cost Function J(X, y, theta) m = size(X, 1) // 3 rows / training examplespredictions = X \* theta;  $sqr Errors = (predictions - y).^2;$  J = 1/(2\*m) \* sum(sqr Errors);

Vectorization

$$h_{\theta}(x) = \sum_{j=0}^{n} \theta_{j} x_{j}$$
$$= \theta^{T} x$$

unvectorized:

prediction = 0.0;

for j = 1: n+1,

prediction = prediction

+ theta(j) \* x(j)

end;

vectorized
prediction = theta \* x;

Gradient Descent

$$\theta_{j} := \theta_{j} - \alpha \frac{1}{m} \sum_{i=1}^{m} (h_{\theta}(x^{(i)}) - y^{(i)}) \chi_{j}^{(i)}$$

$$\frac{\text{Vectorized}}{\hat{\theta} := \hat{\theta} - \alpha \hat{\Delta}} : \frac{1}{m} \sum_{i=1}^{m} \left( h_{\theta}(x^{(i)}) - y^{(i)} \right) x^{(i)}$$

$$\Delta = \begin{bmatrix} \lambda_{\theta} \\ \Delta_{1} \\ \Delta_{2} \end{bmatrix}$$

$$\chi^{(i)} = \begin{bmatrix} \chi_{\theta}^{(i)} \\ \chi_{1}^{(i)} \end{bmatrix}$$