

ELEC 405 - Homework 1

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Contents

| | | |
|----------|-------------------------------------------------------------|----------|
| 1 | Sampling and Matrix Notation | 1 |
| 2 | Vector Space of Polynomials | 2 |
| 3 | Affine functions | 2 |
| 4 | Yet Another Proof of Cauchy-Schwarz... | 3 |
| 5 | Inequality Proof | 3 |
| 6 | First steps in multivariable quadratic functions | 4 |
| 7 | On Gradients.... | 4 |
| 8 | What is a homework without a computational part?.... | 4 |

1. Sampling and Matrix Notation

Let $x \in \mathcal{R}^n$, ie. $x = [x_1 \ x_2 \ \dots \ x_n]$

1. Let $y \in \mathcal{R}^{2n-1}$ and

$$y_k = \begin{cases} x_{\frac{k+1}{2}} & \text{if } k \text{ is odd} \\ 0 & \text{otherwise} \end{cases} \quad (1)$$

for $k = 1, \dots, 2n-1$ and $y = [y_1 \ y_2 \ \dots \ y_{2n-1}]^T$.

Answer: y^T is of the form $[x_1 \ 0 \ x_2 \ 0 \ \dots \ x_n]$. So, $A \in \mathcal{R}^{2n-1 \times n}$ and it should be of the form:

$$A_{ij} = \begin{cases} 1 & \text{if } i \text{ is odd and } j = \frac{i+1}{2} \\ 0 & \text{otherwise} \end{cases} \quad (2)$$

2. Assume n is even and $y \in \mathcal{R}^{\frac{n}{2}}$. Define $y_k = x_{2k}$ for $k = 1, \dots, \frac{n}{2}$.

Answer: $A \in \mathcal{R}^{\frac{n}{2} \times n}$ and

$$A_{ij} = \begin{cases} 1 & \text{if } j = 2 \times i \\ 0 & \text{otherwise} \end{cases} \quad (3)$$

3. Assume n is even and $y \in \mathcal{R}^{\frac{n}{2}}$. Define $y_k = \frac{x_{2k} + x_{2k-1}}{2}$ for $k = 1, \dots, \frac{n}{2}$.

Answer: $A \in \mathcal{R}^{\frac{n}{2} \times n}$ and

$$A_{ij} = \begin{cases} \frac{1}{2} & \text{if } j = 2 \times i \text{ or } j = 2 \times i - 1 \\ 0 & \text{otherwise} \end{cases} \quad (4)$$

2. Vector Space of Polynomials

Let \mathcal{P}_{n-1} represent the vectors space of polynomials with degree less than or equal to $n - 1$. Therefore, each element in this vector space can be written as

$$p(x) = c_0 + c_1x + c_2x^2 + \cdots + c_{n-1}x^{n-1} \quad (5)$$

Answer: $b = \frac{dp(x)}{dx}, b \in \mathcal{R}^{n-1}$ and then $b = [1 \times c_1 \quad 2 \times c_2 \quad \cdots \quad (n-1) \times c_{n-1}]^T$. Then

$$b = Ha$$

$H \in \mathcal{R}^{(n-1) \times n}$ and

$$H_{ij} = \begin{cases} i & \text{if } i = j + 1 \\ 0 & \text{otherwise} \end{cases}$$

$$H = \begin{bmatrix} 0 & 1 & & & \\ 0 & 0 & 1 & & \\ & & \cdots & \ddots & \\ & & & n-1 & \end{bmatrix}$$

3. Affine functions

1. Let $x, y \in \mathcal{R}^n$ and $\alpha + \beta = 1$.

$$\begin{aligned} f(\alpha x + \beta y) &= A(\alpha x + \beta y) + b \\ &= \alpha Ax + \beta Ay + \alpha b + (1 - \alpha)b \\ &= \alpha Ax + \beta Ay + \alpha b + \beta b \\ &= \alpha(Ax + b) + \beta(Ax + b) \\ &= \alpha f(x) + \beta f(y) \end{aligned}$$

2. Let $f : \mathcal{R}^n \rightarrow \mathcal{R}^m$ and be an affine function and $g : \mathcal{R}^n \rightarrow \mathcal{R}^m$ such that $g(x) = f(x) - f(0)$. We will show that $g(x)$ is linear. For a function to be linear, it must satisfy two properties: additivity and homogeneity.

- **Homogeneity:** $g(\alpha x) = \alpha g(x)$

$$\begin{aligned} g(\alpha x) &= f(\alpha x) - f(0) \\ &= f(\alpha x + (1 - \alpha)0) - f(0) \\ &= \alpha f(x) + (1 - \alpha)f(0) - f(0) \\ &= \alpha(f(x) - f(0)) \\ &= \alpha g(x) \end{aligned} \quad (6)$$

Eq (6) follows from the fact that f is affine. Therefore, g satisfies the homogeneity property.

- **Additivity:** $g(x + y) = g(x) + g(y)$

$$\begin{aligned} g(x + y) &= f(x + y) - f(0) \\ &= f\left(\frac{1}{2}(2x) + \frac{1}{2}(2y)\right) - f(0) \\ &= \frac{1}{2}(f(2x) - f(0)) + \frac{1}{2}(f(2y) - f(0)) \end{aligned} \quad (7)$$

$$\begin{aligned}
&= \frac{1}{2}g(2x) + \frac{1}{2}g(2y) \\
&= g(x) + g(y)
\end{aligned} \tag{8}$$

Eq (7) follows from the fact that f is affine. Eq(8) uses the homogeneity property of g proved earlier.

Therefore, g is linear, in other words $\exists A \in \mathcal{R}^{m \times n}$ such that $g(x) = Ax$ for any $x \in \mathcal{R}^n$. Then $f(x)$ can be rewritten as

$$f(x) = g(x) + f(0) = Ax + b$$

where $b = f(0)$ and $b \in \mathcal{R}$.

4. Yet Another Proof of Cauchy-Schwarz...

1. Let $a \geq 0, c \geq 0$ and $\forall \lambda \in \mathcal{R}, a + 2b\lambda + c\lambda^2 \geq 0$.

Let $f : \mathcal{R} \rightarrow \mathcal{R}$ and $f(\lambda) = a + 2b\lambda + c\lambda^2$. For f to be ≥ 0 for all $\lambda \in \mathcal{R}$, f should have one or zero real roots otherwise we can find $\lambda' \in \mathcal{R}$ such that $f(\lambda') < 0$ (Easy to visualize geometrically). So, the determinant Δ should be ≤ 0 .

$$\begin{aligned}
\Delta &= (2b)^2 - 4ac \leq 0 \\
b^2 &\leq ac \\
|b| &\leq \sqrt{ac}
\end{aligned} \tag{9}$$

(9) holds because $a, c \geq 0$.

2. Given $u, w \in \mathcal{R}^n$, explain why $(u + \lambda w)^T(u + \lambda w) \geq 0 \forall \lambda \in \mathcal{R}$.

$$\begin{aligned}
(u + \lambda w)^T(u + \lambda w) &= \langle u + \lambda w, u + \lambda w \rangle \\
&= \langle u, u + \lambda w \rangle + \langle \lambda w, u + \lambda w \rangle \\
&= \langle u, u \rangle + \langle u, \lambda w \rangle + \langle \lambda w, u \rangle + \langle \lambda w, \lambda w \rangle \\
&= \|u\|^2 + 2\lambda u^T w + \lambda^2 \|w\|^2 \\
&= (u + \lambda w)^2 \\
&\geq 0
\end{aligned} \tag{10}$$

(10) and previous ones follow from the properties of inner products.

3. Eq (10) is of the form of quadratic equation in the first part. Namely, here $a = \|u\|^2, c = \|w\|^2$ and $b = \langle u, w \rangle, a, c \geq 0$ for any $\lambda \in \mathcal{R}$. Then by the first part, we know that $|b| \leq \sqrt{ac}$ so

$$\begin{aligned}
|u^T w| &\leq \sqrt{\|u\|^2 \|w\|^2} \\
&\leq \sqrt{u^T u} \times \sqrt{w^T w}
\end{aligned} \tag{11}$$

(11) follows from the fact that $\|u\|^2 = \langle u, u \rangle = u^T u$.

5. Inequality Proof

Let $a, b \in \mathcal{R}^n$, and $x = a + b, y = -b$ by triangle inequality

$$\begin{aligned}
\|x + y\| &\leq \|x\| + \|y\| \\
\|a\| &\leq \|a + b\| + \|-b\| \\
\|a\| - \|b\| &\leq \|a + b\|
\end{aligned}$$

6. First steps in multivariable quadratic functions

$$f(x_1, x_2, x_3) = 2x_1^2 + 3x_1x_3 - 2x_3x_2 - x_2^2 + 4x_3^2$$

can be written of the form $x^T Ax$ and A is not unique because of the non-zero x_{ij} terms. As long as $A_{ij} + A_{ji}$ equals to the coefficient of $x_i \times x_j$. Two possible choices of A is $\begin{bmatrix} 2 & 0 & 3 \\ 0 & -1 & -2 \\ 0 & 0 & 4 \end{bmatrix}$ and its

transpose $\begin{bmatrix} 2 & 0 & 0 \\ 0 & -1 & 0 \\ 3 & -2 & 4 \end{bmatrix}$.

7. On Gradients....

Let $f : \mathcal{R}^n \rightarrow \mathcal{R}$ be a differentiable multivariate function.

$$\nabla f(x) = \begin{bmatrix} \frac{\partial f}{\partial x_1} \\ \vdots \\ \frac{\partial f}{\partial x_n} \end{bmatrix} \quad (12)$$

Find the explicit gradients of the following functions

1. $f(x) = a^T x + b$ for some $a \in \mathcal{R}^n, b \in \mathcal{R}$.
 $\nabla f(x) = a$ because $f(x) = a_1x_1 + x_2x_2 + \dots a_nx_n + b$ so $\frac{\partial f}{\partial x_i} = a_i$.
2. $f(x) = x^T Ax$ for some $A \in \mathcal{R}^{n \times n}$.
 $\nabla f(x) = (A + A^T)x$ because $f(x) = x^T Ax$ equivalently

$$\begin{aligned} f(x) &= x^T Ax \\ &= \sum_{i=1}^n \sum_{j=1}^n x_i A_{ij} x_j \\ &= \sum_{i=1}^n x_i A_{i1} x_1 + \sum_{j=1}^n x_1 A_{1j} x_j + \sum_{i=2}^n \sum_{j=2}^n x_i A_{ij} x_j \end{aligned} \quad (13)$$

$$\frac{\partial f}{\partial x_1} = \sum_{i=1}^n x_i A_{i1} + \sum_{j=1}^n A_{1j} x_j \quad (14)$$

Let $A = \begin{bmatrix} a_1^T \\ a_2^T \\ \vdots \\ a_n^T \end{bmatrix} = \begin{bmatrix} b_1 & b_2 & \dots & b_n \end{bmatrix}$ where $a_i, b_i \in \mathcal{R}^n$. Then (14) can be rewritten of the form $x^T b_1 + x^T a_1$. Note that $x^T b_1$ corresponds to the first row of $A^T x$ so $\frac{\partial f}{\partial x_1} = (A_i + A_i^T) x_i^T$

8. What is a homework without a computational part?....

$f(x) = x^T Ax \Rightarrow \nabla f(x) = 2Ax$, because A is symmetric.

11/8/2020

A1

Elec 405 - Homework 1

```
In [5]: using LinearAlgebra
        using Plots
```

8. Affine Approximation

$$a(\mathbf{x}) = f(\mathbf{x}_0) + \nabla f(\mathbf{x}_0)^T (\mathbf{x} - \mathbf{x}_0) \quad (7)$$

is an affine approximation of $f(\mathbf{x})$ around \mathbf{x}_0 .

Note that this affine function defines an hyperplane in \mathbb{R}^{n+1} in the standard form

$$\begin{bmatrix} \nabla f(\mathbf{x}_0) \\ -1 \end{bmatrix}^T \left(\begin{bmatrix} \mathbf{x} \\ a(\mathbf{x}) \end{bmatrix} - \begin{bmatrix} \mathbf{x}_0 \\ f(\mathbf{x}_0) \end{bmatrix} \right) = 0. \quad (8)$$

Now consider the function

$$f(\mathbf{x}) = \mathbf{x}^T \mathbf{A} \mathbf{x} \quad (9)$$

defined over \mathbb{R}^2 , where

$$\mathbf{A} = \begin{bmatrix} 0.01 & 0.001 \\ 0.001 & 0.01 \end{bmatrix}. \quad (10)$$

```
In [6]: A = [0.01 0.001; 0.001 0.01]
        A
```

```
Out[6]: 2x2 Array{Float64,2}:
 0.01  0.001
 0.001 0.01
```

11/8/2020

A1

```
In [7]: function draw_hyperplane_3D_modified(a, b, camera=(60, 45))

#     pyplot()
x=range(-50, stop=50, length=100)
y=range(-50, stop=50, length=100)
g(x, y) = (dot(a, b) - dot(a[1: 2], [x; y]))/a[3]
plot!(x, y, g,
      label="$a'*(x-$b)=0",
      xlabel="x_1", ylabel="x_2", zlabel="x_3",
      st=:surface, camera=camera)

# # plot the normal vector, since b is on the plane plot the vector b to b+a
a_x1 = range(b[1], stop=(b[1]+15*a[1]), length=100)
a_x2 = range(b[2], stop=(b[2]+15*a[2]), length=100)
a_x3 = range(b[3], stop=(b[3]+15*a[3]), length=100)
plot!(a_x1, a_x2, a_x3, label="a")
end

function draw_hyperplane_3D(a, b, p, camera=(60, 45))
    pyplot()
    x=range(-2, stop=2, length=100)
    y=range(-2, stop=2, length=100)
    g(x, y) = (dot(a, b) - dot(a[1: 2], [x; y]))/a[3]
    p= plot(x, y, g,
           title="Hyperplane in 3D",
           label="$a'*(x-$b)=0",
           xlabel="x_1", ylabel="x_2", zlabel="x_3",
           xlims= (-2, 2), ylims=(-2, 2), zlims=(-2, 2),
           st=:surface, camera=camera)

    # plot the normal vector, since b is on the plane plot the vector b to b+a
    a_x1 = range(b[1], stop=(b[1]+a[1]), length=100)
    a_x2 = range(b[2], stop=(b[2]+a[2]), length=100)
    a_x3 = range(b[3], stop=(b[3]+a[3]), length=100)
    plot!(a_x1, a_x2, a_x3, label="a")
end
```

Out[7]: draw_hyperplane_3D (generic function with 2 methods)

```
In [8]: function affine_approx(x, f, f_der, x_0)
        return f(x_0) + dot(f_der(x_0), (x-x_0))
end

function f(x)
    return x' * A * x
end

function nabla_f(x)
    return (A+A')x
end
```

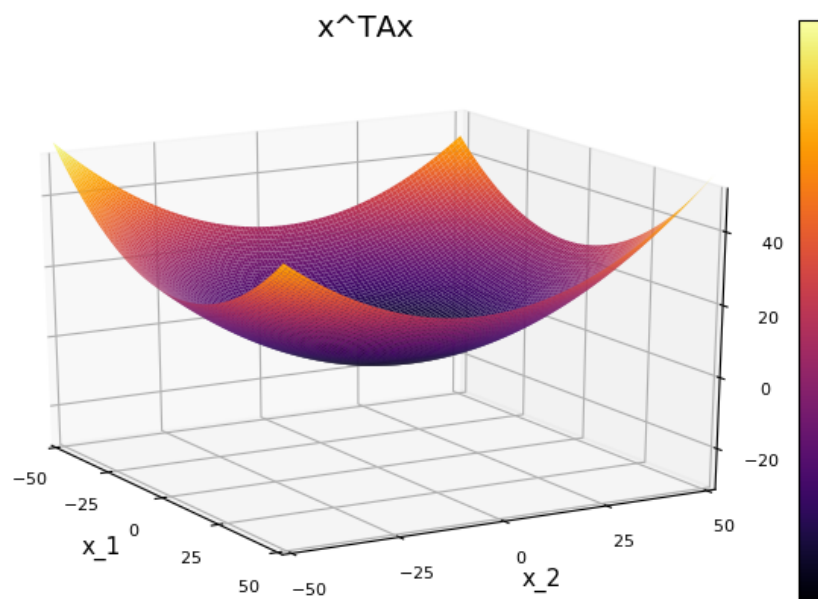
Out[8]: nabla_f (generic function with 1 method)

11/8/2020

A1

```
In [11]: camera=(60, 20)
# draw_hyperplane_3D()
pyplot()
x=range(-50, stop=50, length=100)
y=range(-50, stop=50, length=100)
g(x, y) = ([x y] * A * [x; y])[1]
p= plot(x, y, g,
        title="x^TAx",
        label="x^TAx",
        xlabel="x_1", ylabel="x_2",
        zlims=(-30,50),
        st=:surface, camera=camera)
```

Out[11]:



11/8/2020

A1

```

In [12]: camera=(30, -10)

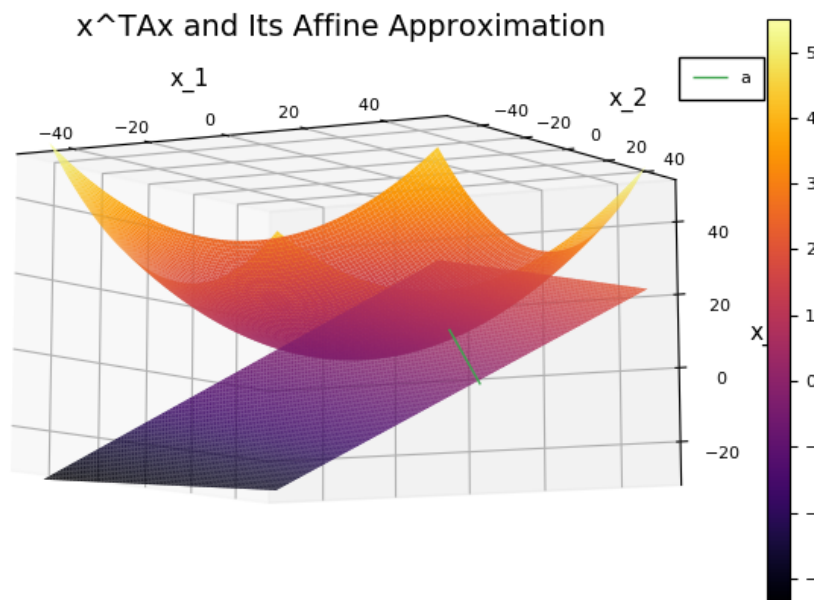
pyplot()
x=range(-50, stop=50, length=100)
y=range(-50, stop=50, length=100)
g(x, y) = ([x y] * A * [x; y])[1]
p= plot(x, y, g,
        title="x^TAx and Its Affine Approximation",
        label="x^TAx",
        xlabel="x_1", ylabel="x_2",
        zlims=(-30,50),
        st=:surface, camera=camera)

x_0 = [25; 0]
a = [nabla_f(x_0); -1]
b = [x_0; f(x_0)]

draw_hyperplane_3D_modified(a, b, camera)

```

Out[12]:



In []: