

# Reminder

- Invitation to complete course feedback by end of Saturday

# Mathematics for Robotics (ROB-GY 6013 Section A)

- Exam Format
- Mental Map
- Things to expect/not expect
- Extra Study Material Posted
- Recorded session link:  
<https://nyu.zoom.us/rec/share/ZgyIF5gTGop6TkLgRc6bdkHKN72Z7P3vW9gyaCXXgad6whzJv6TY0oRu3ftl0XI.th903pisLGRm5BLX>

# Exam Format

- **Not cumulative.**
- **No proofs.**
- The exam will be closed-book and closed notes. A formula sheet will be provided to you (shared on NYU Brightspace).
- There will be both short answer and long answer questions.
  - Short answers can be true/false, multiple choice, or fill-in-the-blank.
  - For long answer questions, you must show all your work to obtain full credit.

# Mental Map (How to think about the last 8 weeks)

- **Lengths, Distance, and Orthogonality**

- Norms and Inner Products (Inner Product Spaces)
- Gram-Schmidt Process

- **Road to Least Squares**

- (Pre-)Projection Theorem, Normal Equations
- Weighted Least Squares, Recursive Least Squares

- **Symmetric Matrices**

- Positive-(semi)definiteness, Quadratic Form, Schur's Complement, other properties

- **Road to Kalman Filter**

- Probability, BLUE, MVE, Kalman Filter

- **Computations**

- QR factorization, SVD, LU Factorization, Euler's Method, Newton's Method

# Things to Expect/Not Expect

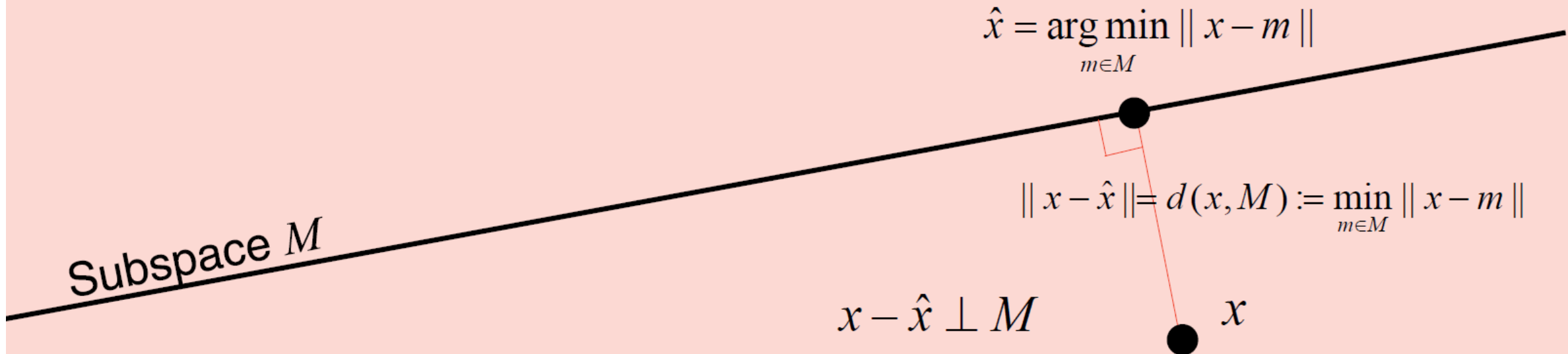
- **Week 7–8 content:**
  - There will be a problem that applies Gram-Schmidt process
  - Reminder that you can take the norms and inner products of functions
    - Reminder that many things that don't look like vectors are vectors and have norms/inner products. Be prepared.

# Things to Expect/Not Expect

- **Week 9 content:**
  - There may be a problem that requires you to creatively apply the normal equations (see next slide)
  - Use this big picture to guide you

Let  $(\mathcal{X}, \mathcal{F}, \langle \cdot, \cdot \rangle)$  be a **finite-dimensional** (real) **inner product space**,  $M$  be a **subspace** of  $\mathcal{X}$ , and  $x$  be an arbitrary **point** in  $\mathcal{X}$ .

**Inner product space**  $(\mathcal{X}, \mathcal{F}, \langle \cdot, \cdot \rangle)$



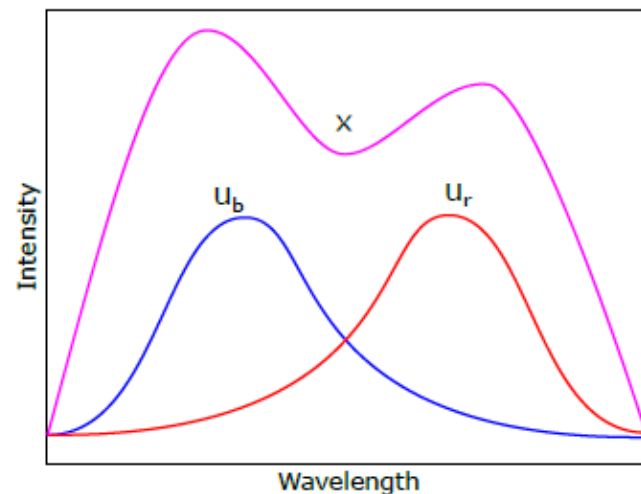


Figure 1: Diagram of our LED color profiles  $u_b$ , and  $u_r$ , and our target color  $x$ .

**7. (15 points)** The space of all color profiles can be specified by functions  $I : [\Lambda_{min}, \Lambda_{max}] \rightarrow \mathbb{R}$ , where  $I(\lambda)$  is the intensity of the color at each wavelength  $\lambda$ . We can represent this space of colors as an inner product space  $(\mathcal{X}, \mathbb{R}, \langle \cdot, \cdot \rangle)$  where  $\mathcal{X} := \{I : [\Lambda_{min}, \Lambda_{max}] \rightarrow \mathbb{R}\}$  is a finite-dimensional vector space of continuous functions and the inner product is defined as

$$\langle f, g \rangle := \int_{\Lambda_{min}}^{\Lambda_{max}} f(\lambda)g(\lambda)d\lambda.$$

Suppose we have two LEDs, one red, one blue, with corresponding intensities  $\{u_b, u_r\}$  that can be combined and scaled linearly to form different colors (see Figure. 1). Our goal is to best approximate a desired color with intensity function  $x$  using these two LEDs.

We are given the following properties:

$$\langle u_b, u_b \rangle = 1 \quad \langle u_b, u_r \rangle = 0.5 \quad \langle u_r, u_r \rangle = 1 \quad \langle x, u_b \rangle = 2.5 \quad \langle x, u_r \rangle = 1.5$$

**Find** a linear combination of the colors blue and red that best approximates our desired color profile  $x$ . That is, find:

$$\alpha^* = \begin{bmatrix} \alpha_b^* \\ \alpha_r^* \end{bmatrix} = \arg \min_{\alpha_b, \alpha_r \in \mathbb{R}} \|x - (\alpha_b u_b + \alpha_r u_r)\|^2$$

using the norm induced by the inner product. Record your answer in the box.

**Problem 7:**

The set of linear combinations of  $\{u_b, u_r\}$  is a subspace of  $\mathcal{X}$ , so we use the normal equations to find  $\alpha^*$ :

$$\begin{bmatrix} \langle u_b, u_b \rangle & \langle u_b, u_r \rangle \\ \langle u_b, u_r \rangle & \langle u_r, u_r \rangle \end{bmatrix} \begin{bmatrix} \alpha_b^* \\ \alpha_r^* \end{bmatrix} = \begin{bmatrix} \langle x, u_b \rangle \\ \langle x, u_r \rangle \end{bmatrix}$$

$$\begin{bmatrix} 1 & 1/2 \\ 1/2 & 1 \end{bmatrix} \begin{bmatrix} \alpha_b^* \\ \alpha_r^* \end{bmatrix} = \begin{bmatrix} 5/2 \\ 3/2 \end{bmatrix}$$

$$\begin{bmatrix} \alpha_b^* \\ \alpha_r^* \end{bmatrix} = \frac{4}{3} \begin{bmatrix} 1 & -1/2 \\ -1/2 & 1 \end{bmatrix} \begin{bmatrix} 5/2 \\ 3/2 \end{bmatrix}$$

$$\begin{bmatrix} \alpha_b^* \\ \alpha_r^* \end{bmatrix} = \begin{bmatrix} 7/3 \\ 1/3 \end{bmatrix}$$

(a) The answer is unique, since the Gram matrix:  $\begin{bmatrix} \langle u_b, u_b \rangle & \langle u_b, u_r \rangle \\ \langle u_b, u_r \rangle & \langle u_r, u_r \rangle \end{bmatrix}$  is full rank and invertible.



# Things to Expect/Not Expect

- **Weeks 10–13 content:**
  - Only **conceptual** and **short answer** questions on probability
    - If  $X_1$  and  $X_2$  are Gaussian random vectors, is their linear combination also a Gaussian random vector?
      - Answer: No
    - What does  $E\{\hat{x} - x\} = 0$  mean in the context of our derivations?
      - Answer: the estimation algorithm is **unbiased**.
  - You should review the derivations of BLUE, MVE, and Kalman filter from the perspective of understanding the major steps rather than memorizing anything.

# Things to Expect/Not Expect

- **Week 14 content:**
  - There will be a problem requiring computations for each of the following:
    - QR factorization (Largest will be  $3 \times 3$  matrix)
    - SVD (Largest will be  $3 \times 3$  matrix)
    - LU factorization (Largest will be  $3 \times 3$  matrix)
    - Euler's Method
  - **Formula sheet will not include anything on Week 14. You must memorize!**
  - **For the factorizations, you may use any method you like.**
  - **For Euler's Method, you must use Euler's method.**
- **Newton-Raphson method is not on the exam**

# Extra Study Material

- Formula sheet
- Handout for Euler's Method posted on NYU Brightspace
- Will post select solutions to homework along with code
- WolframAlpha and MATLAB are very convenient for creating matrices and practicing QR factorization, SVD, and LU Factorization.