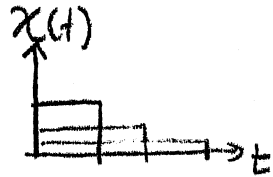


# Math 424 - Assignment 1 - Solutions

Chapter 0, problem 8:

$$\text{Let } X_n(t) = \begin{cases} \frac{1}{n} & \text{on } [0, n] \\ 0 & \text{on } (n, \infty) \end{cases}$$



then  $X_n(t)$  converges uniformly to 0 as given any  $\epsilon > 0$ , then for  $\frac{1}{n} < \epsilon$  i.e.  $\frac{1}{\epsilon} < n$

then  $|X_n(t) - 0| = \frac{1}{n} < \epsilon$  so that we have uniform convergence.

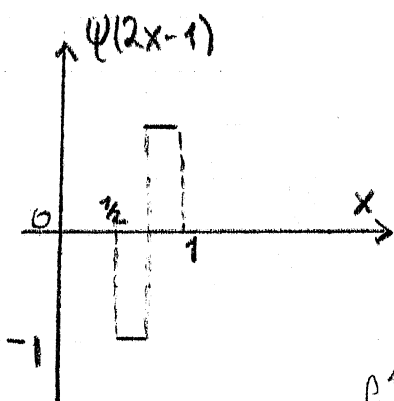
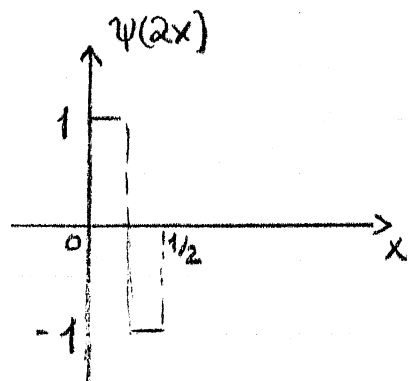
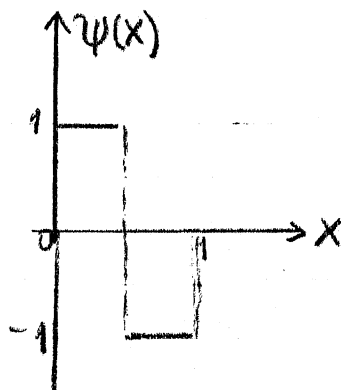
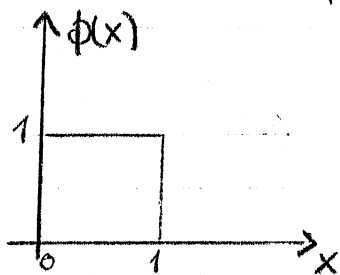
$$\text{But } \int_0^{\infty} |X_n(t)|^2 dt = \int_0^n X_n(t)^2 dt = \int_0^n \frac{1}{n} dt = 1 \quad \forall n$$

$$\Rightarrow X_n \not\rightarrow 0 \text{ in } L^2[0, \infty)$$

so the result 0.10 does not hold on  $[0, \infty)$ .

□

### Chapter 0, problem 15:



You can actually immediately see from the graphs that the functions are mutually orthogonal.

E.g. let us compute more explicitly

$$\int_0^1 \psi(x) \psi(2x) dx = \int_0^{1/4} dx + \int_{1/4}^{1/2} -1 dx = 0$$

or also immediately  $\int_0^1 \psi(2x) \psi(2x-1) dx = 0$  as supports are disjoint.

Formally you should verify all orthogonality relations

Scaling  $\psi(2x)$  by  $\sqrt{2}$  &  $\psi(2x-1)$  by  $\sqrt{2}$  to get

$\psi_1(x) = \sqrt{2} \psi(2x)$      $\psi_2(x) = \sqrt{2} \psi(2x-1)$  we get an

orthonormal basis e.g. to check  $\int_0^1 2 \psi^2(2x) dx = 2 \cdot 1/2 = 1$

Now simply apply Theorem 0.21 i.e. compute

$$\alpha_1 := \langle f, \phi \rangle = \int_0^1 x dx = 1/2 \quad \alpha_2 := \langle f, \psi \rangle = \int_0^{1/2} x dx - \int_{1/2}^1 x dx$$

$$= \frac{1}{2} x^2 \Big|_0^{1/2} - \frac{1}{2} x^2 \Big|_{1/2}^1$$

$$= \frac{1}{2} \cdot \frac{1}{4} - \frac{1}{2} + \frac{1}{2} \cdot \frac{1}{4} = -\frac{1}{4}$$

$$\alpha_3 := \langle f, \psi_1 \rangle = \sqrt{2} \int_0^{1/4} x \, dx - \sqrt{2} \int_{1/4}^{1/2} x \, dx = \sqrt{2} \left( \frac{1}{2} \cdot \frac{1}{16} - \frac{1}{2} \cdot \frac{1}{4} + \frac{1}{2} \cdot \frac{1}{16} \right)$$

$$= \sqrt{2}/2 \left( -\frac{1}{8} \right) = -\sqrt{2}/16$$

$$\alpha_4 := \langle f, \psi_2 \rangle = \sqrt{2} \int_{1/2}^{3/4} x \, dx - \sqrt{2} \int_{3/4}^1 x \, dx$$

$$\frac{\sqrt{2}}{2} \left( x^2 \Big|_{1/2}^{3/4} - x^2 \Big|_{3/4}^1 \right) = \sqrt{2}/2 \left( \frac{9}{16} - \frac{1}{4} - 1 + \frac{9}{16} \right)$$

$$= -\frac{1}{8} \cdot \left( \sqrt{2}/2 \right) = -\sqrt{2}/16$$

So the orthogonal projection  $p(f)$  onto the space is

$$p(f) = \frac{1}{2} \phi(x) - \frac{1}{4} \psi(x) - \frac{1}{8} \psi(2x) - \frac{1}{8} \psi(2x-1) \quad \square$$

Chapter 0, problem 19:

The defining equation for the adjoint is

$$\langle T(f), g \rangle = \langle f, T^*(g) \rangle \quad f, g \in L^2(\mathbb{R})$$

Let us start with the LHS above

$$\langle T(f), g \rangle = \int_{\mathbb{R}} T(f)(x) \overline{g(x)} dx = \int_{\mathbb{R}} \int_{\mathbb{R}} f(y) K(x, y) dy \overline{g(x)} dx$$

$$= \iint_{\mathbb{R} \times \mathbb{R}} f(y) \overline{g(x)} K(x, y) dy dx$$

$$= \iint_{B \subset \mathbb{R}^2} f(y) \overline{g(x)} K(x, y) dy dx$$

where  $B$  is the bounded set where  $K$  is supported i.e.  $K(x, y) = 0$  on  $B^c$

conjugate symm.

$$= \iint_B g(x) \overline{K(x, y)} dx \overline{f(y)} dy$$

• on bdd. sets we can interchange

$$= \iint_{\mathbb{R}^2} g(x) \overline{K(x, y)} dx \overline{f(y)} dy$$

• all fcts are integrable on  $B$

eg.  $\int_a^b |g(x)| dx \leq (b-a)^{1/2} \int_a^b |g(x)|^2 dx < \infty$

cauchy schwarz

$$= \int_{\mathbb{R}} f(y) \int_{\mathbb{R}} \overline{g(x)} K(x, y) dx dy$$

$$= \int_{\mathbb{R}} f(y) \int_{\mathbb{R}} \overline{K(x, y)} g(x) dx dy = \langle f, T^*(g) \rangle$$

as defined in exercise

so since  $T^*$  as defined satisfies the adjoint equation, we are done.  $\square$

Chapter 0, problem 21:

Suppose  $\exists x \in V$  s.t.  $Ax = b$ . Assume for contradiction that there is also a vector  $w \in W$  s.t.  $A^*w = 0$  and  $\langle b, w \rangle \neq 0$

i.e.  $\langle Ax, w \rangle \neq 0$ . Using the definition of an adjoint we get  $\langle x, A^*w \rangle \neq 0$  but  $A^*w = 0 \Rightarrow$  contradiction.

Conversely, we need to show that if  $Ax \neq b$  for any  $x \in V$  then we can find  $w \in W$  s.t.  $A^*w = 0$  and  $\langle b, w \rangle \neq 0$

Define  $w = b - \underbrace{\text{projection}_{\text{range } A}}_{:= p(b)}(b) \neq 0$  as  $b \neq Ax$  for any  $x$

Also  $\langle w, z \rangle = 0 \quad \forall z \in \text{range } A$  by construction of the projection

But  $(\text{range } A)^\perp = \ker(A^*)$  by problem 20 so  $w \in \ker A^*$  i.e.

$$A^*w = 0$$

It remains to show  $\langle b, w \rangle \neq 0$

If  $\langle b, w \rangle = \langle b, b - p(b) \rangle = 0 \Leftrightarrow \langle b, b \rangle = \langle b, p(b) \rangle$   
 $\|b\|^2 = \langle b, p(b) \rangle$

$\downarrow$  c.s.  
But  $\langle b, p(b) \rangle \leq \|b\| \|p(b)\|$  so  $\|p(b)\| = \|b\|$

But  $b = b - p(b) + p(b)$  and  $b - p(b) \in \text{range}(A)^\perp$

so  $\|b\|^2 = \|b - p(b)\|^2 + \|p(b)\|^2$

$\|b - p(b)\|^2 > \|p(b)\|^2$  if the projection is non-trivial  $\Rightarrow \text{⚡}$

Notice that so far we have implicitly assumed  $b \neq 0$ ; but we need to look at this special case:

$b=0 \Rightarrow \langle w, b \rangle = 0 \forall w$  i.e. we need to show that  $\exists x$  s.t.  $Ax=0$  but  $x \neq 0$  does the job.  $\square$

Chapter 0, problem 23:

Suppose  $\{\phi_i\}_{i \in I}$  I some index set are orthonormal in a vector space  $V$  and also assume (for contradiction) that

$$\sum_{k=1}^n c_{i_k} \phi_{i_k} = 0 \text{ for some } c_{i_k} \neq 0;$$

i.e. that there exist a linear dependence relation.

Taking inner products on both sides and using bilinearity

$$\langle 0, \phi_{i_1} \rangle = \langle \sum_{k=1}^n c_{i_k} \phi_{i_k}, \phi_{i_1} \rangle = \sum_{k=1}^n c_{i_k} \langle \phi_{i_k}, \phi_{i_1} \rangle =$$

$\underset{0}{\parallel}$

$$= c_{i_1} \langle \phi_{i_1}, \phi_{i_1} \rangle = c_{i_1} \neq 0 \rightarrow \text{orthonormality} \quad \rightarrow \text{⚡}$$

□

Chapter 0, problem 26:

- (a) Unfortunately, this question is not as precise as it should. It is not defined what is meant by least squares problem.

Essentially the discussion of pages 23-25 gives that we search for a minimum of

$$\tilde{E} = \sum_{i=1}^N |y_i - (mx_i + b)|^2$$

Now this is equivalent to minimizing

$$E(m, b) = \sum_{i=1}^N |mx_i + b - y_i|^2 \quad \text{since } |a| = |-a| = |-1||a|$$

(notice that we can assume  $x_i, y_i \in \mathbb{R}$ )

- (b) Suppose  $\frac{\partial E}{\partial b} = \frac{\partial E}{\partial m} = 0$ , then  $E$  has a critical point

Since we have to compute the derivatives anyway we can try to use the calculus criterion for minima of multivariate functions, i.e.

$$E_{bb} E_{mm} - E_{mb}^2 > 0 \quad \text{and} \quad E_{mm} > 0$$

$\Rightarrow$  local minimum

$$E_b = \sum 2(mx_i + b - y_i) \quad E_m = \sum 2(mx_i + b - y_i) x_i$$

$$E_{bb} = 2 \sum_{i=1}^N 1 = 2N \quad E_{mm} = 2 \sum_{i=1}^N x_i^2 \Rightarrow E_{mm} > 0 \quad \text{if } x_i \neq x_j \text{ and } N \geq 2$$

$$E_{mb} = 2 \sum_{i=1}^N x_i$$



$$E_{bb} E_{mm} - E_{mb}^2 = 4N \cdot \sum_{i=1}^N x_i^2 - \left( 2 \sum_{i=1}^N x_i \right)^2$$

$$= 4N \sum_{i=1}^N x_i^2 - 4 \left( \sum_{i=1}^N x_i \right)^2$$

$$= 4 \left( N \sum_{i=1}^N x_i^2 - \left( \sum_{i=1}^N x_i \right)^2 \right)$$

$$= 4 \left( (N-1) \sum_{i=1}^N x_i^2 - \sum_{\substack{i,j \\ i \neq j}} x_i x_j \right) > 0$$

complete the square for  $N \geq 2$   
 $x_i \neq x_j$   
 for some  $i, j$

So we have a local minimum if we have a critical point  $\Rightarrow$  global minimum.

Alternatively, you can also use the convexity of  $E(m, b)$  to get to the result...  $\lrcorner$

$$(c) \quad E_b = 0 \Leftrightarrow \sum 2(mx_i + b - y_i) = 0$$

$$E_m = 0 \Leftrightarrow \sum 2x_i (mx_i + b - y_i) = 0$$

$$\text{so that } \sum 2(mx_i - y_i) + Nb = 0 \text{ i.e. } b = -\frac{\sum 2(mx_i - y_i)}{N}$$

$$\Rightarrow 0 = \sum 2x_i \left( mx_i - \frac{\sum 2(mx_k - y_k)}{N} - y_i \right)$$

$$0 = \frac{1}{N} \sum 2x_i (Nmx_i - \sum 2(mx_k - y_k) - Ny_i)$$

$$0 = \sum_i 2Nmx_i^2 - \sum_i 2x_i \sum_k 2mx_k + \sum_i 2x_i \sum_k y_k - \sum_i Ny_i$$

$$0 = m \left( \sum_i 2Nx_i^2 - \sum_{i,k} 4x_i x_k \right)$$

$i=B$

$$\Rightarrow m = -A/B \quad \square$$