

Homework 8

Exercise 3.1.

1) We have

$$\begin{aligned}\mathbb{E}_{\sigma,u} [\|F(v) - u\|^2] &= \mathbb{E}_{\sigma,u} [\|\lambda F(v) + (1-\lambda)v - u\|^2] \\ &= \mathbb{E}_{\sigma,u} [\|\lambda(F(v) - u) + (1-\lambda)(v - u)\|^2] \\ &= \lambda^2 \mathbb{E}_{\sigma,u} [\|F(v) - u\|^2] \\ &\quad + (1-\lambda)^2 \mathbb{E}_{\sigma,u} [\|v - u\|^2] \\ &\quad + 2\lambda(1-\lambda) \mathbb{E} [\langle F(v) - u | v - u \rangle]\end{aligned}$$

2) We also have

$$\mathbb{E}_{\sigma,u} [\langle F(v) - u | v - u \rangle] = \mathbb{E}_u [\mathbb{E}_{\sigma} [\langle F(v) - u | v - u \rangle | u]]$$

and

$$\begin{aligned}\mathbb{E}_{\sigma} [\langle F(v) - u | v - u \rangle | u] \\ &= \sum_x \mathbb{E}_{\sigma_x, v_x} [\langle F(v)_x - u_x | u_x - v_x \rangle | u], \text{ 2 patches or pixels} \\ &\quad \text{(indep of } v_x \text{)}.\end{aligned}$$

$$\begin{aligned}\text{and } \mathbb{E}_{\sigma_x, v_x} [\langle F(v)_x - u_x | u_x - v_x \rangle | u] \\ &= \mathbb{E}_{v_x} [\mathbb{E}_{\sigma_x} [\langle F(v)_x - u_x | u_x - v_x \rangle | u, \sigma_x] | u] \\ &= \mathbb{E}_{v_x} [\langle F(v)_x - u_x | u_x - \mathbb{E}_{\sigma_x} [v_x | u, \sigma_x] \rangle | u] \\ &= \mathbb{E}_{\sigma_x} [\langle F(v)_x - u_x | \underbrace{u_x - \mathbb{E}_{\sigma_x} [v_x | u]}_0 \rangle | u] \\ &= 0\end{aligned}$$

$$\text{ie } \boxed{\mathbb{E}_{\sigma,u} [\langle F(v) - u | v - u \rangle] = 0}$$



- 3) In the end, minimising the MSE leads to minimising  $\lambda^2 E_{u,v} [\|F(v) - u\|^2] + (1-\lambda)^2 E_{u,v} [\|v - u\|^2]$

This is a polynome of order 2 of the form  $\lambda^2 a + (1-\lambda)^2 b = P(\lambda)$   
 ie  $P(\lambda) = 2\lambda a - 2(1-\lambda)b$   
 $= 2[\lambda(a+b) - b]$

The  $\lambda^*$  minimising the MSE is:

$$\lambda^* = \frac{E_{u,v} [\|v - u\|^2]}{E_{u,v} [\|F(v) - u\|^2] + E_{u,v} [\|v - u\|^2]}$$

$$\begin{aligned} \text{Finally, } E_{u,v} [\|v - u\|^2] &= E_u [E_v [\|v - u\|^2 | u]] \\ &= E_u [E_v [\|v - E[v|u]\|^2 | u]] \end{aligned}$$

be unbiased estimator.

This leads to

$$\lambda^* = \frac{E_u [V[v|u]]}{E_{u,v} [\|F(v) - u\|^2] + E_u [V[v|u]]}$$

- 4) The assumptions of prop 3.4 are verified. Thus  $E_v [\|F(v) - v\|^2] = E_{v,u} [\|F(v) - u\|^2] + E_u [V[v|u]]$

and

$$\lambda^* = \frac{E_u [V[v|u]]}{E_v [\|F(v) - v\|^2]}$$

By definition  $R_{NLS}(F) = E_v [\|F(v) - v\|^2]$ .

$$\lambda^* = \frac{d\sigma^2}{R_{NLS}(F)}$$



### Exercise 3.2:

let  $v$  be the noisy version of  $u$ , and  $\hat{u}(v)$  an estimator of  $u$ .

We have

$$\begin{aligned} & \mathbb{E}_v [\| \hat{u}(v) - \mathbb{E}_v [\hat{u}(v) | u] \|^2 | u] \\ &= \mathbb{E}_v [\| \hat{u}(v) - u - \mathbb{E}_v [\hat{u}(v) | u] + u \|^2 | u] \\ &= \mathbb{E}_v [\| \hat{u}(v) - u \|^2 | u] \\ &\quad + \mathbb{E}_v [\| \underbrace{\mathbb{E}_v [\hat{u}(v) | u] - u}_{\text{doesn't depend on } v} \|^2 | u] \\ &\quad - \mathbb{E}_v [2 \langle \hat{u}(v) - u | \mathbb{E}_v [\hat{u}(v) | u] - u \rangle | u] \\ &= \mathbb{E}_v [\| \hat{u}(v) - u \|^2 | u] \\ &\quad + \| \mathbb{E}_v [\hat{u}(v) | u] - u \|^2 \\ &\quad - 2 \underbrace{\langle \mathbb{E}_v [\hat{u}(v) | u] - u | \mathbb{E}_v [\hat{u}(v) | u] - u \rangle}_{2 \| \mathbb{E}_v [\hat{u}(v) | u] - u \|^2} \\ &= \mathbb{E}_v [\| \hat{u}(v) - u \|^2 | u] - \| \mathbb{E}_v [\hat{u}(v) | u] - u \|^2 \end{aligned}$$

In the end:

$$\boxed{\mathbb{E}_v [\| \hat{u}(v) - u \|^2 | u] = \| \mathbb{E}_v [\hat{u}(v) | u] - u \|^2 + \mathbb{E}_v [\| \hat{u}(v) - \mathbb{E}_v [\hat{u}(v) | u] \|^2 | u].}$$