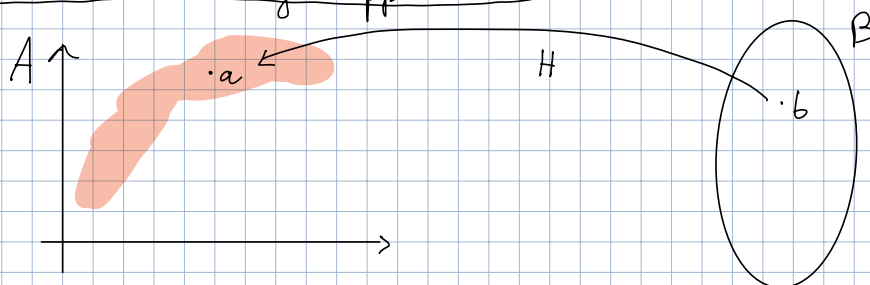


3. Machine learning approaches




H : depends on $\begin{cases} F \\ \text{The prior on } a \text{ (R)} \end{cases}$

H : can be (partially) learned

Choices to make:

i) Training set

supervised: $\{(b_i, a_i)\}_{i=1}^N$
 $b_i = F(a_i) + \epsilon_i$

unsupervised $\{a_i\}_{i=1}^N \in$ 
 $[\{b_i\}_{i=1}^N]$

ii) Architecture / model for $H \approx H_\theta$

End-to-end (supervised)

Ignore F , learn H directly by minimizing

$$\star \min_{\theta} \frac{1}{N} \sum_{i=1}^N \|H_\theta(b_i) - a_i\|^2 + \text{Reg}(\theta)$$

H_θ : NN, θ are parameters

\rightarrow nothing but a regression problem in sup. learning

Plug-and-play (unsupervised: $\{a_i\}$)

Recall:

$$a_0 = 0 \quad \leftarrow \text{prox op.}$$

$$a_{j+1} = S \left(a_j - \gamma F^T(F a_j - b) \right)$$

stop: $a_K = H(b)$

Key obs: S is a denoiser:

$$S(a) = \underset{x}{\operatorname{argmin}} \frac{1}{2} \|x - a\|^2 + f(x)$$

Learn S ! How? Minimize \star with a Training set:

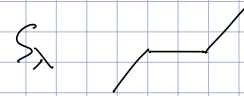
$$\{(a_j + \tilde{\epsilon}_j, a_j)\}_{j=1}^N$$

Unrolling / Unfolding (supervised)

$$\begin{aligned} a_0 &= 0 \\ a_{j+1} &= S(a_j - \gamma F^T(Fa_j - b)) \\ \text{stop: } a_K &= H(b) \end{aligned}$$

prox op

e.g. $S = \Phi^T S_\lambda \Phi$



$H: b \mapsto a_K$ is a NN with S_λ as nonlinearity

Key idea: replace S at iteration j by a NN N_{θ_j}

The full H_θ is:

$$\begin{aligned} a_0 &= 0 \\ a_{j+1} &= N_{\theta_j}(a_j - \gamma F^T(Fa_j - b)) \\ H_\theta(b) &= a_K \end{aligned}$$

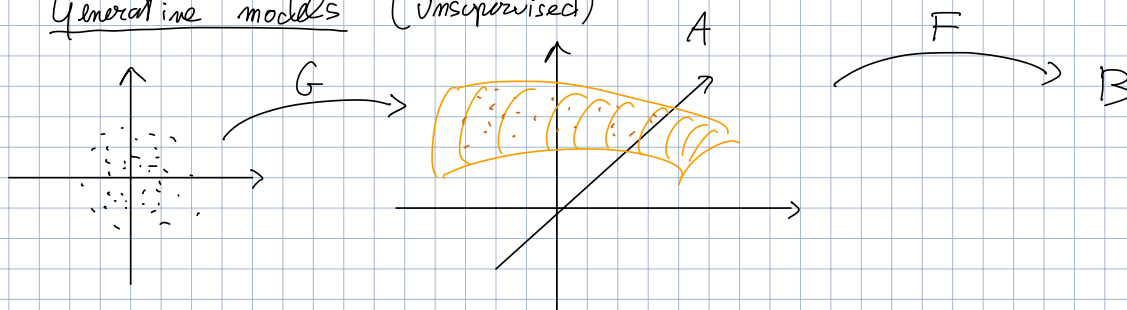
is a NN (shallow) itself

Train H_θ , $\theta = (\theta_0, \theta_1, \dots, \theta_{K-1})$, by minimizing \star

$$\min_{\theta} \frac{1}{N} \sum_{i=1}^N \|H_\theta(b_i) - a_i\|^2 + \operatorname{Reg}(\theta)$$

The Training depends on F !

Generative models (unsupervised)



Instead of solving the IP:

$$F(a) = b$$

$$a = G(z)$$

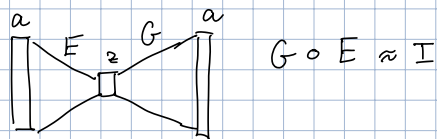
Solve

$$F(G(z)) = b$$

then $a_{\text{rec}} = G(z_{\text{rec}})$

Many options:

- variational autoencoders
- GANs
- diffusion models



Deep image priors (no learning, no training set)

Given a NN

$$G_{\theta} : \mathbb{R}^d \longrightarrow A \subset \mathbb{R}^n$$

(fix an architecture). Sample $z \sim N(0, I)$ in \mathbb{R}^d (fix it!)

Solve

$$\min_{\theta} \|F(G_{\theta}(z)) - b\|_B^2 + \text{Reg}(\theta)$$

$$a = G_{\theta}(z), z \text{ fixed.}$$

References

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