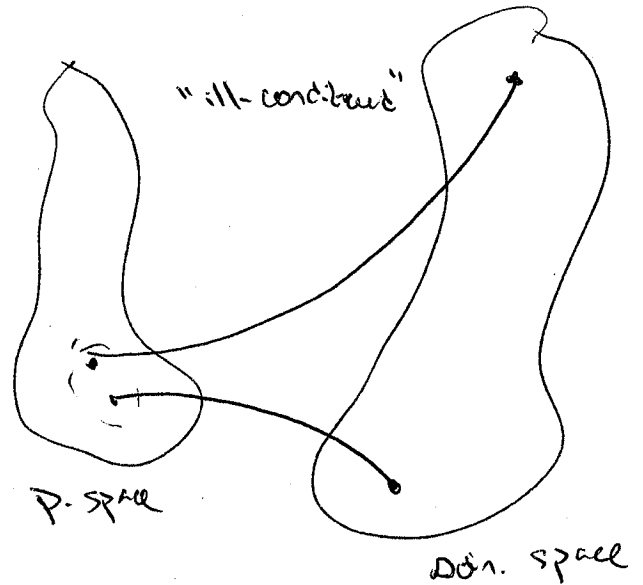
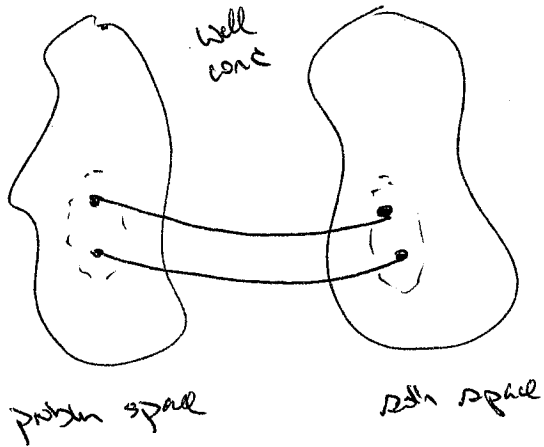


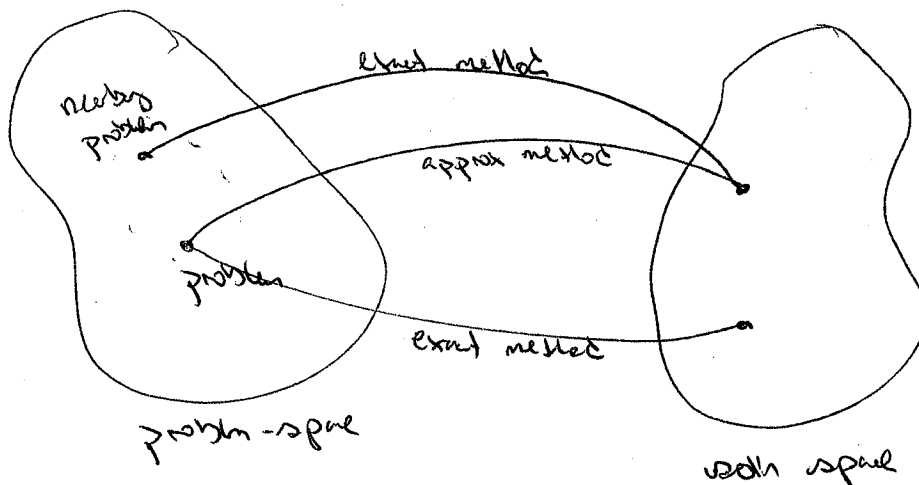
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- Conditioning + stability
 - Sensitivity to perturbed data
 - Ex: scalar addition.

Part III : Conditioning + Stability :

A problem is well-conditioned if small changes in the data lead to small changes in the soln



An algorithm or method is called stable if it always produces a soln to a nearby problem.



if a problem is ill-conditioned, stability of the algorithm is not ~~very~~ useful.
less

Problem conditioning is essential to numerical methods b/c we work with rounded data.

If an algorithm is applied to a problem with data d to get a sol'n s , then $s = S(d)$.

We are interested in the relative error $\frac{\|S(d) - S(\hat{d})\|}{\|s\|}$ where \hat{d} is rounded (or otherwise polluted) data.

Ex 1: scalar addition.

- data: $a, b \in \mathbb{R}$
- solution: $s = a + b$
- perturbed data: $\tilde{a} = a + \Delta a$, $\tilde{b} = b + \Delta b$
- perturbed sol'n: $\tilde{s} = \tilde{a} + \tilde{b}$

$$\begin{aligned} \text{relative error} : \frac{\|\tilde{s} - s\|}{\|s\|} &= \frac{\|\tilde{a} + \tilde{b} - (a + b)\|}{\|s\|} = \frac{\|\Delta a + \Delta b\|}{|a + b|} \leq \frac{|\Delta a| + |\Delta b|}{|a + b|} \\ &= \frac{1}{|a + b|} \left(|a| \frac{|\Delta a|}{|a|} + |b| \frac{|\Delta b|}{|b|} \right) \end{aligned}$$

$$\Rightarrow \underbrace{\frac{\|\tilde{s} - s\|}{\|s\|}}_{\text{rel. pert. in sol'n}} \leq \underbrace{\frac{|a| + |b|}{|a + b|}}_{\text{amplification}} \times \underbrace{\max \left\{ \frac{|\Delta a|}{|a|}, \frac{|\Delta b|}{|b|} \right\}}_{\text{relative pert. in data.}}$$

$c = \frac{|a| + |b|}{|a + b|}$ magnifies the perturbation to the data.

if $\text{sign}(a) = \text{sign}(b)$, $c = 1$

if $\text{sign}(a) = -\text{sign}(b)$, $c > 1$

if $|a + b| \ll |a| + |b|$ (cancellation error), $c \gg 1$: c "blows up" perturbation of data.

\Rightarrow subtraction with cancellation is ill-conditioned!

More precise: back rotation

let $\delta(x)$ be a small perturbation ~~to~~ of x

$$\delta(f) = f(x + \delta x) - f(x) \quad ("f" \text{ is data}, "x" \text{ is data})$$

The absolute condition number $\hat{\kappa} = \lim_{\delta \rightarrow 0} \sup_{\|\delta x\| \leq \delta} \frac{\|\delta f\|}{\|\delta x\|}$

It looks like a ~~condition number~~ derivative.

JF ~~is~~ $J(x)$ is the Jacobian of f at x : $J_{ij} = \frac{\partial f_i}{\partial x_j}$

then $\delta f \approx J(x) \delta x$ and $\hat{\kappa} = \|J(x)\|$.

Usually we are concerned with the relative condition number

$$\kappa = \kappa(x) = \lim_{\delta \rightarrow 0} \sup_{\|\delta x\| \leq \delta} \left(\frac{\frac{\|\delta f\|}{\|f\|}}{\frac{\|\delta x\|}{\|x\|}} \right)$$

$$\Rightarrow \frac{\|\delta f\|}{\|f\|} \leq \kappa \frac{\|\delta x\|}{\|x\|}$$

\nwarrow change in f
 \uparrow magnitude at next step
 \nearrow change in data

~~scribbles~~