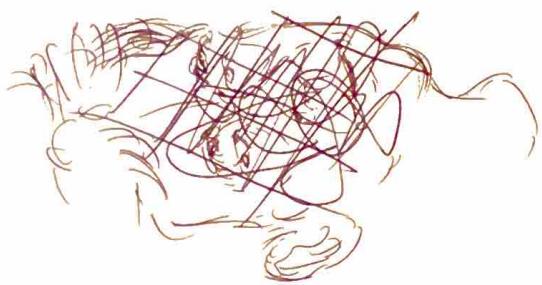




$$\text{Kullback-Leibler divergence} \\ \text{KL}[q||p] = \mathbb{E}_{q(m)} \ln q(m|\phi) - \mathbb{E}_{p(m)} \ln p(m|\phi)$$



Bayesian solution $P(\text{model}) = \frac{P(\text{data})}{P(\text{model})}$

4. The semi-analytic function and
a. which has a prior distribution
and fit of data no prior, no posterior



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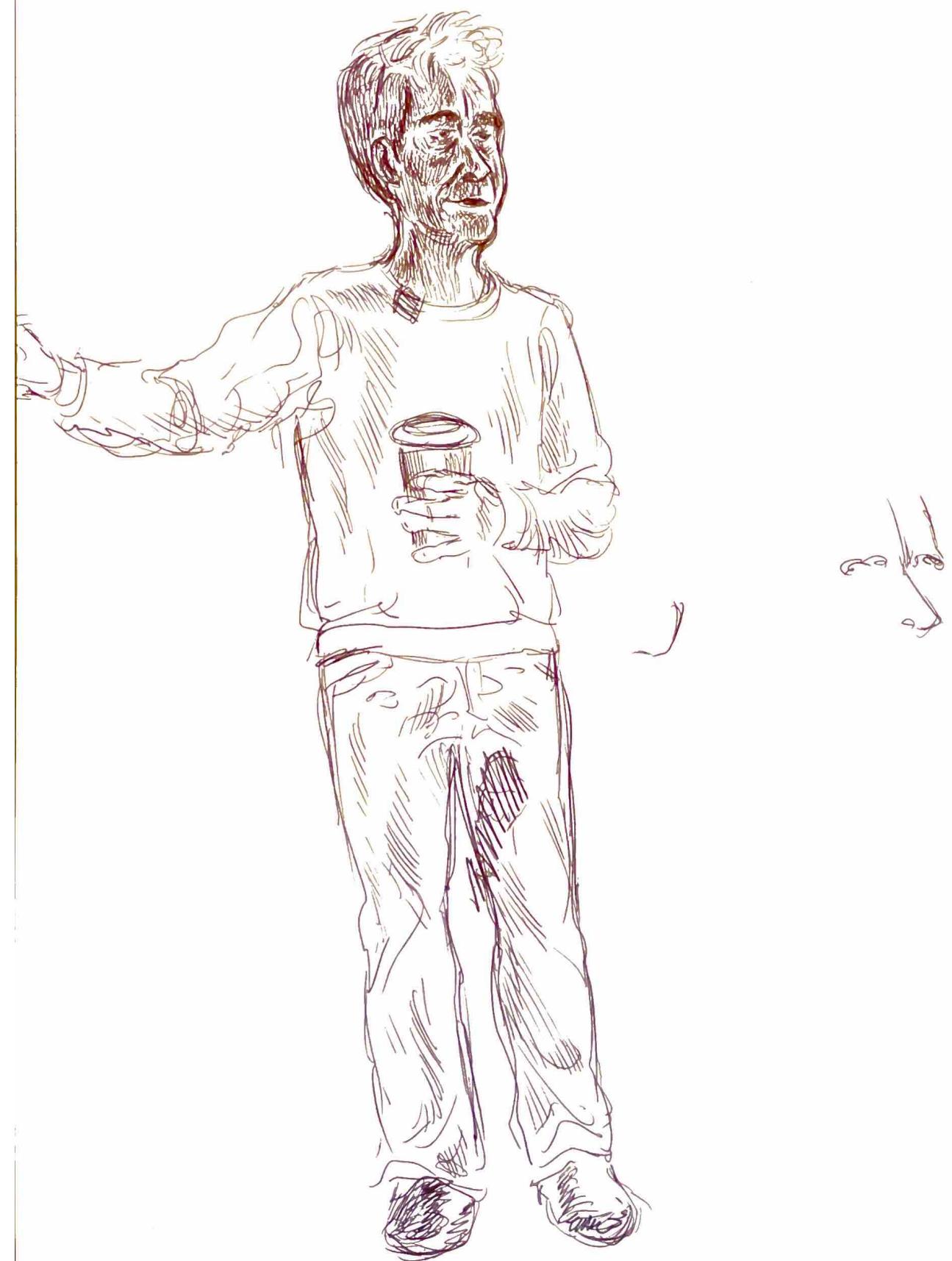
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~~Method~~ Derivation of discrete adjoints - Adjoint method

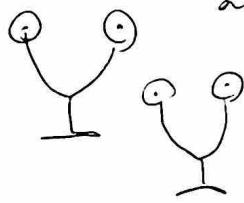
$$\chi[\underline{u}(\underline{m})], \quad \frac{\partial \chi}{\partial m_i} = \partial_i \chi = \frac{\partial \chi^T}{\partial \underline{u}} \partial_i \underline{u} = \nabla \chi^T \partial_i \underline{u}$$

$$\underline{L}(\underline{m}) \underline{u} = \underline{f} \Rightarrow \partial_i \underline{L} \underline{u} + \underline{\partial_i \underline{u}} = \underline{\phi} \quad // \text{re multiply the eq. per } \underline{V^T}$$

↓ ↓ different.

both depend
on model prms

product rule



f on the sources
and one independent
of the earth model prms

$$[V^T \underline{L} + \nabla \chi^T] \partial_i \underline{u}$$

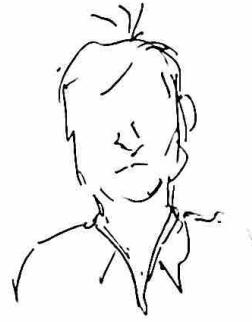
~~$$V^T \partial_i \underline{L} \underline{u} + V^T \underline{L} \partial_i \underline{u} + \nabla \chi^T \partial_i \underline{u} = \partial_i \chi$$~~

$$\underline{L}^T \underline{V} = -\nabla \chi \rightarrow \partial_i \chi = V^T \partial_i \underline{L} \underline{u}$$

adjoint eq.

V = adjoint field





Descent direction

$$\underline{x}(m.) - \underline{x}(m_0) < \phi$$

$$\lim_{\gamma \rightarrow 0} \underline{x}(m_0 - \gamma \underline{h}) - \underline{x}(m_0) < \phi$$

\underline{x}



$\underline{h}^\top \nabla \underline{x} < 0$ descent direction
any vector \underline{h} that satisfy
this eq. is the descent
direction

$$\underline{h} = -\nabla \underline{x} \Rightarrow -\underline{x}^\top \nabla \underline{x} = -|\underline{x}|^2 < 0$$

$$\underline{h} = -\underline{A} \nabla \underline{x} \Rightarrow -(\nabla \underline{x}^\top \underline{A} \nabla \underline{x})$$

only
positive
eigenvalues

always $> \phi$

you have infinite
descent directions

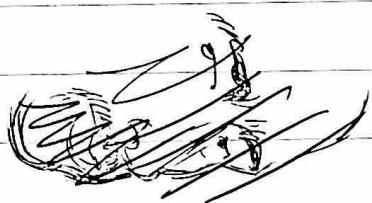
$< \phi$

$$\underline{h} = -\underline{A} \nabla \underline{x}$$



if $\underline{h} < \phi$





Dear
all,



**AI HU LL
PALACE HOTEL**

A vertical sketch of a woman's face, looking slightly to the right. She has dark hair, a high forehead, and is wearing a small hoop earring. To the right of the sketch, there is handwritten text in black ink. The text reads:

of Multilayer perception
of
of perceptron
theory
for the place
in your memory