

Earth like magnetic field vector magnetometry: Rb atoms, EIT, and machine learning

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Magnetic fields and magnetometers capabilities



Magnetic field

- Human brain: 0.1 – 1 pT
- Human heart: 100 pT

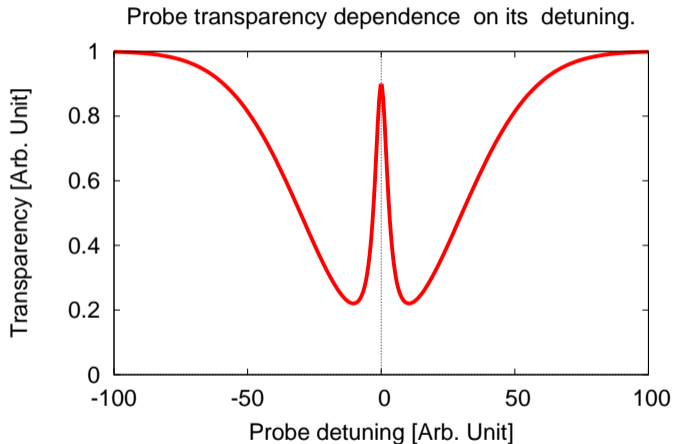
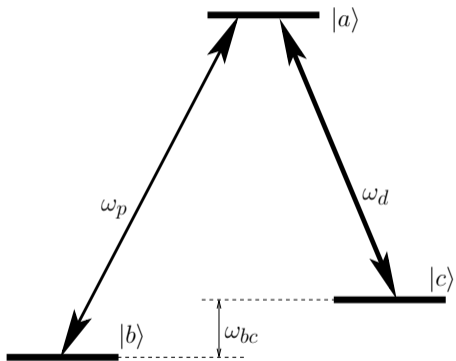
Magnetometers

- SQUID: 1 fT
- SERF: 1 fT

Goal

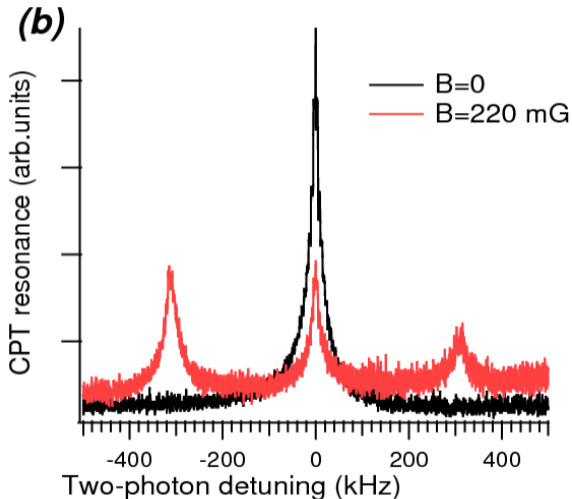
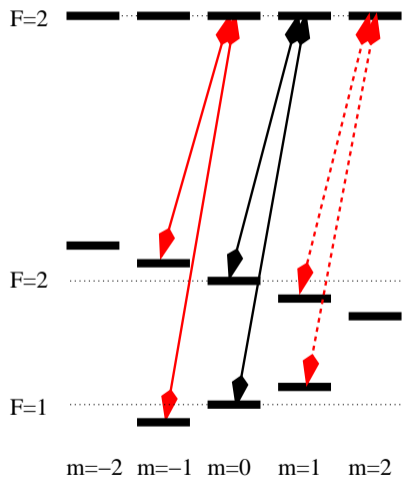
compact ($< 1 \text{ cm}^3$) **vector** magnetometer with pT precision
capable to work at Earth magnetic field ($50 \mu\text{T}$)

Electromagnetically Induced Transparency (EIT)

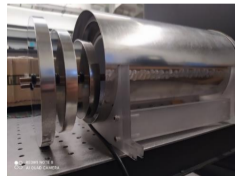
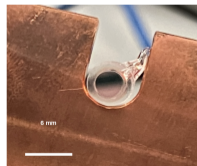
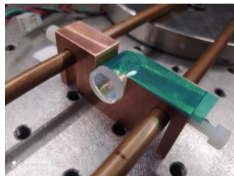
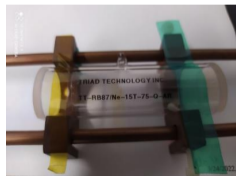
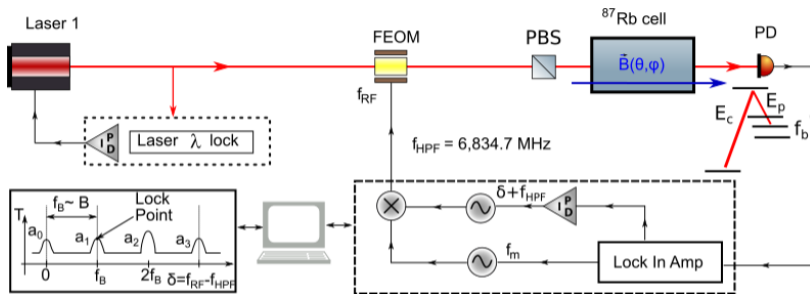


Simple EIT magnetometer

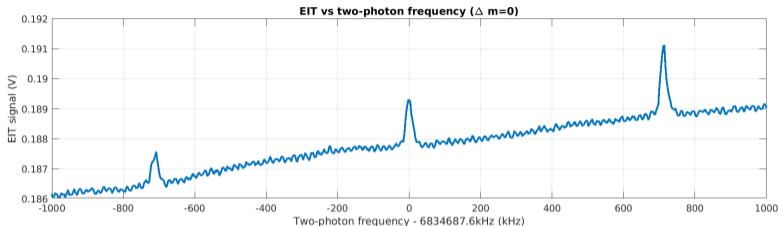
EIT with circularly polarized light



Conceptual design



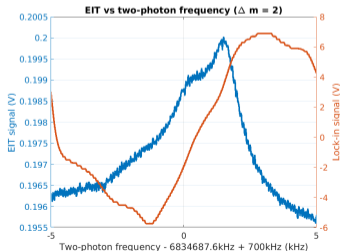
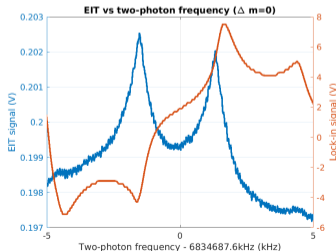
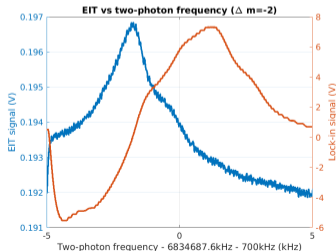
EIT signals vs two-photon detuning. $B=50\mu\text{T}$, $f_c = 6'834'687.6\text{kHz}$



$\Delta m = -2$, $f_{center} - 700\text{kHz}$

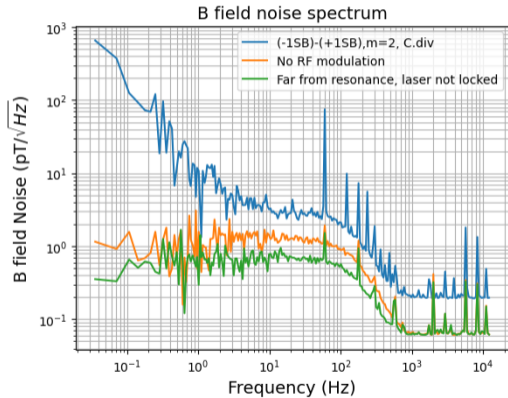
$\Delta m = 0$, $f_{center} = 0$

$\Delta m = 2$, $f_{center} + 700\text{kHz}$

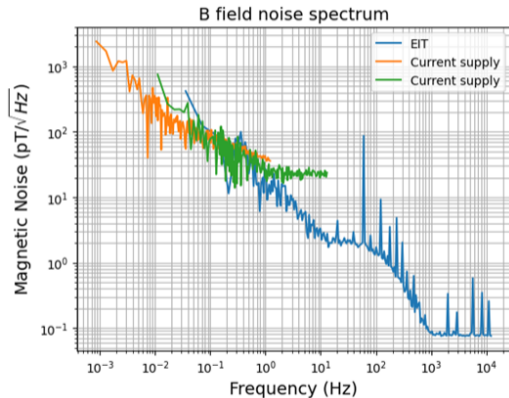


B field sensitivity

Limiting noise sources

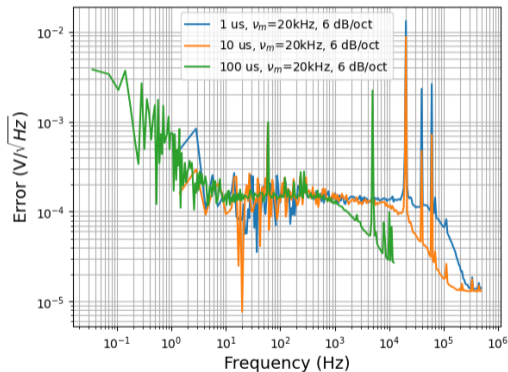


Equivalent current source noise

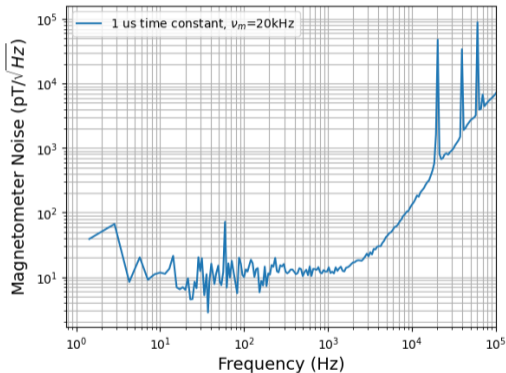


B field sensitivity normalized response

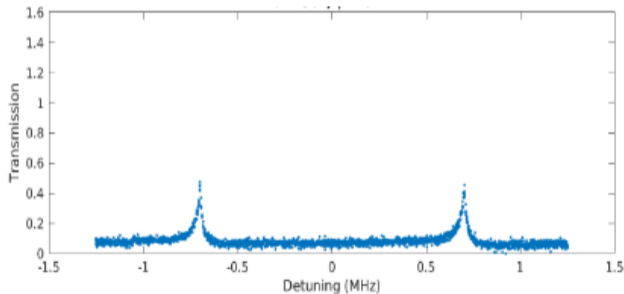
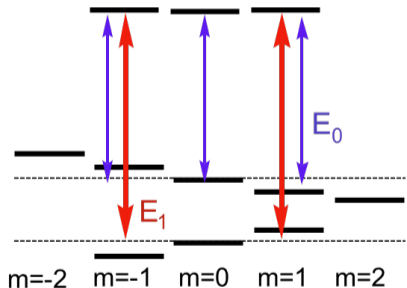
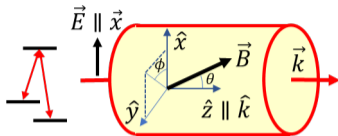
Lock-in time constant limiting response



True magnetometer sensitivity



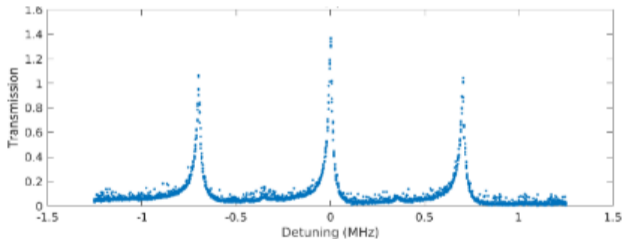
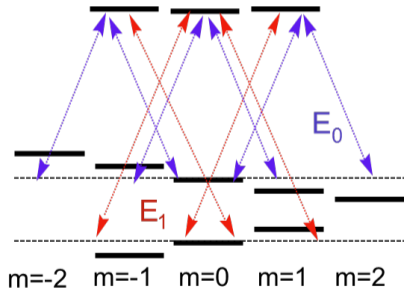
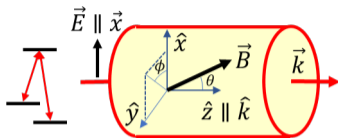
Compass idea



V. I. Yudin *et al.* Phys. Rev. A 82, 033807

Kevin Cox *et al.* Phys. Rev. A 83, 015801

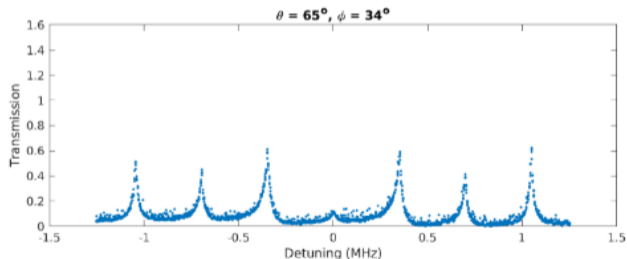
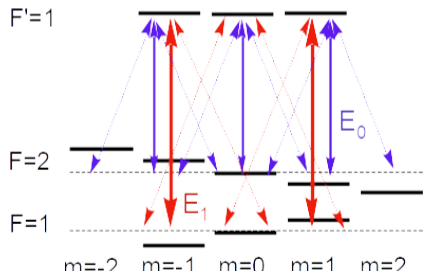
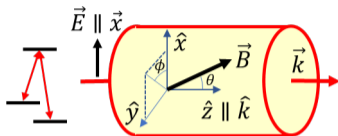
Compass idea



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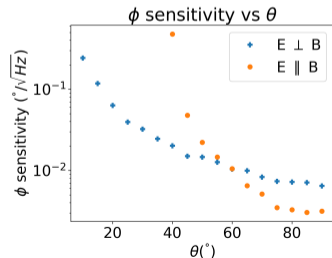
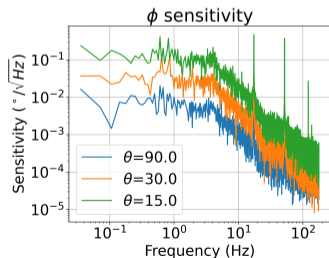
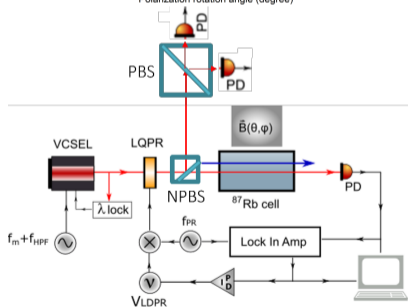
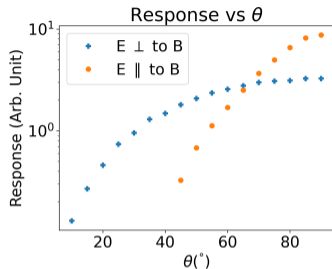
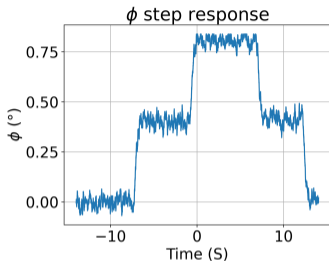
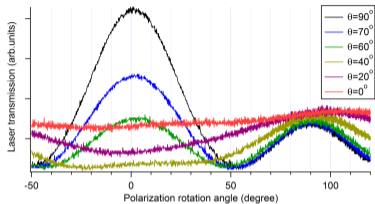
Kevin Cox *et al.* Phys. Rev. A 83, 015801

Compass idea

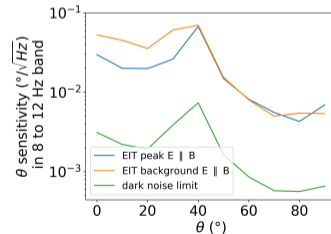
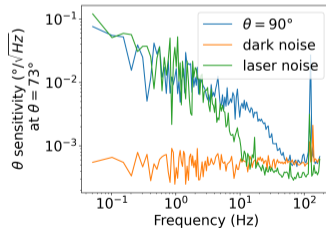
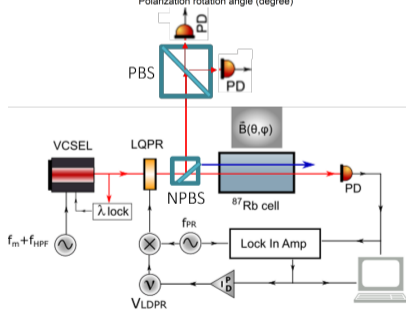
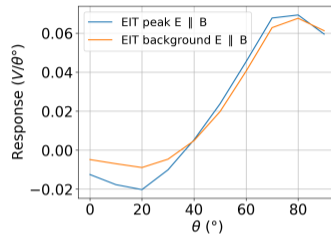
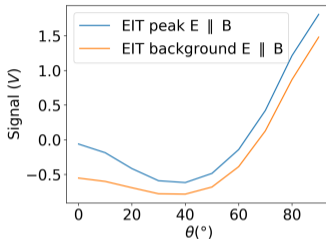
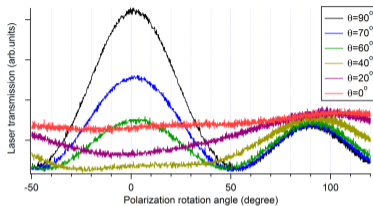


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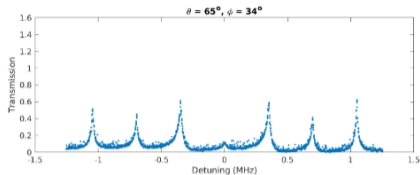
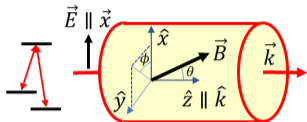
ϕ angle tracking sensitivity



θ angle tracking sensitivity, $E \parallel B$

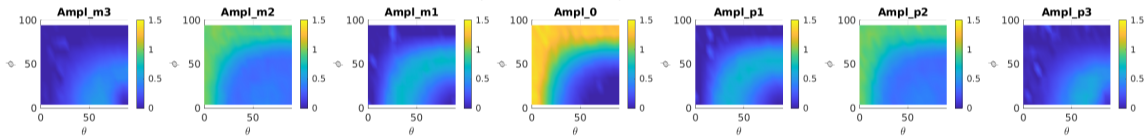


Linear polarization: angular dependence on θ and ϕ



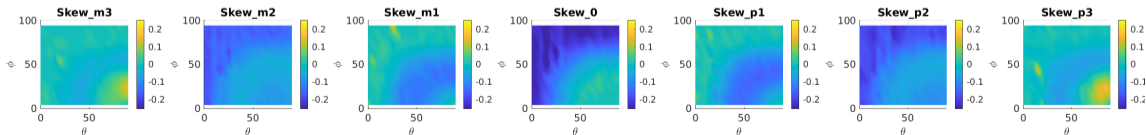
Amplitude

EIT parameter for linear polarization



Skew

EIT parameter for linear polarization



Principal component analysis: quick intro

We form measurement matrix M

$$\begin{array}{c} \text{\scriptsize } m \text{ measurements} \\ \updownarrow \\ \left(\begin{array}{cccc} m_{1,1} & m_{1,2} & m_{1,3} & \dots & m_{1,n} \\ m_{2,1} & m_{2,2} & m_{2,3} & \dots & m_{2,n} \\ \dots & \dots & \dots & \dots & \dots \\ m_{m,1} & m_{m,2} & m_{m,3} & \dots & m_{m,n} \end{array} \right) \end{array} \begin{array}{c} \text{\scriptsize } n \text{ features} \\ \leftarrow \text{-----} \rightarrow \\ \end{array} = M_{m \times n} = \left(\begin{array}{c} \text{Plot 1} \\ \text{Plot 2} \\ \dots \\ \text{Plot } m \end{array} \right)$$

and its approximation with product of 2 low ranked matrix W and P

$$\tilde{M}_{m \times n} = W_{m \times k} \times P_{k \times n}$$

Physical meaning of W and P

$$\begin{pmatrix} m_{1,1} & m_{1,2} & m_{1,3} \\ m_{2,1} & m_{2,2} & m_{2,3} \\ m_{3,1} & m_{3,2} & m_{3,3} \\ \dots & \dots & \dots \\ m_{m,1} & m_{m,2} & m_{m,3} \end{pmatrix} = \begin{pmatrix} w_{1,1} & w_{1,2} \\ w_{2,1} & w_{2,2} \\ w_{3,1} & w_{3,2} \\ \dots & \dots \\ w_{m,1} & w_{m,2} \end{pmatrix} \times \begin{pmatrix} p_{1,1} & p_{1,2} & p_{1,3} \\ p_{2,1} & p_{2,2} & p_{2,3} \end{pmatrix}$$

Every measurement can be approximated as a superposition of patterns

$$\begin{pmatrix} m_{2,1} & m_{2,2} & m_{2,3} \end{pmatrix} = \begin{pmatrix} w_{2,1} \end{pmatrix} \times \begin{pmatrix} p_{1,1} & p_{1,2} & p_{1,3} \end{pmatrix} + \begin{pmatrix} w_{2,2} \end{pmatrix} \times \begin{pmatrix} p_{2,1} & p_{2,2} & p_{2,3} \end{pmatrix}$$

- P - is the set of patterns or templates or directions
- W - is the set of weights or scores or components

Principal components analysis/decomposition (PCA)

$$\begin{pmatrix} m_{1,1} & m_{1,2} & m_{1,3} \\ m_{2,1} & m_{2,2} & m_{2,3} \\ m_{3,1} & m_{3,2} & m_{3,3} \\ \dots & \dots & \dots \\ m_{m,1} & m_{m,2} & m_{m,3} \end{pmatrix} = \begin{pmatrix} w_{1,1} & w_{1,2} \\ w_{2,1} & w_{2,2} \\ w_{3,1} & w_{3,2} \\ \dots & \dots \\ w_{m,1} & w_{m,2} \end{pmatrix} \times \begin{pmatrix} p_{1,1} & p_{1,2} & p_{1,3} \\ p_{2,1} & p_{2,2} & p_{2,3} \end{pmatrix}$$

There are many possible realizations of W and P matrices.

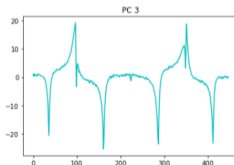
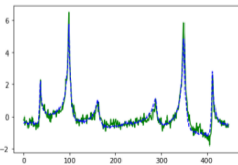
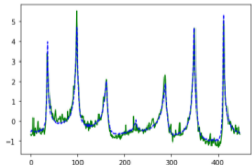
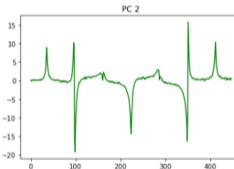
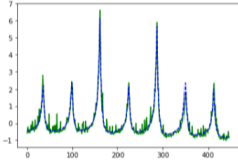
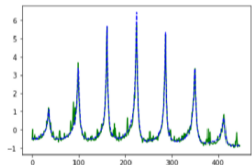
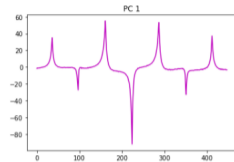
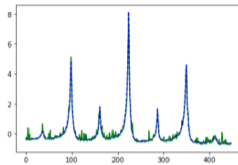
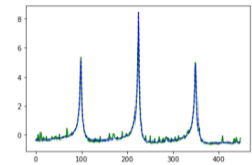
PCA requires

- minimizes $\|M_{m \times n} - W_{m \times k} \times P_{k \times n}\| = \epsilon_k$
- $\vec{P}_i \times \vec{P}_j = \delta_{ij}$, orthonormality of patterns
- $\epsilon_{k_1} < \epsilon_{k_2}$, if $k_1 < k_2$

Many practical data sets can be described with truncated $k < \min(m, n)$

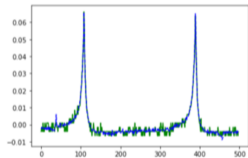
Use the single value decomposition (SVD) to get W and P .

Principal component analysis: get components

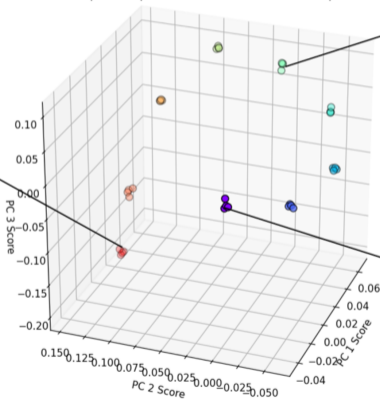


Principal component analysis: use components space

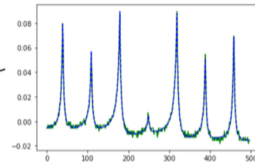
Phi: 0.0348104778286055
Theta: 89.9540402279556



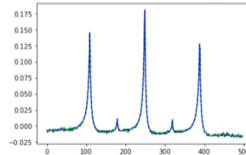
Principal Component Scores for all Values of $\phi = 0$



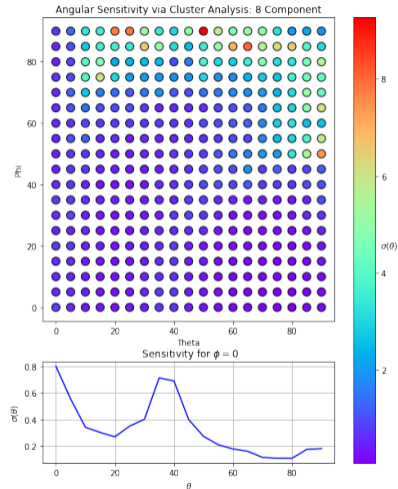
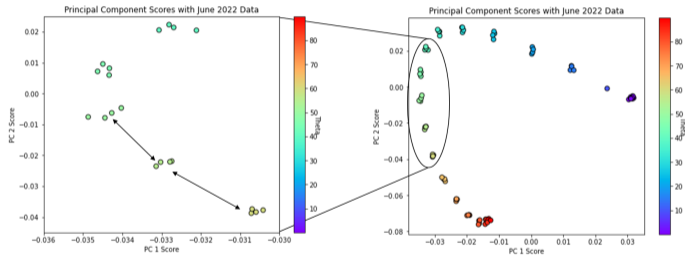
Phi: 0.0454438259069784
Theta: 50.1330676548235



Phi: 0.189462531619154
Theta: 10.0930794347724



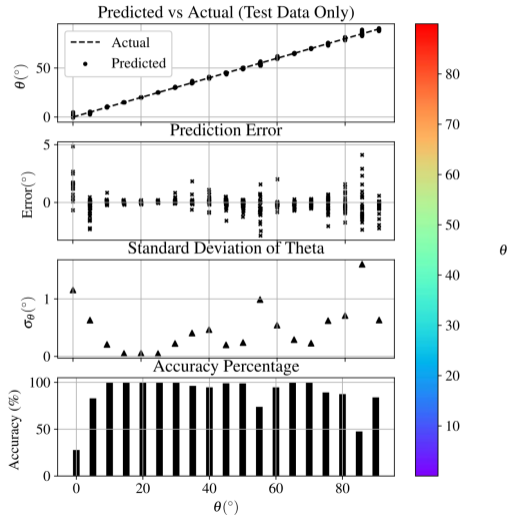
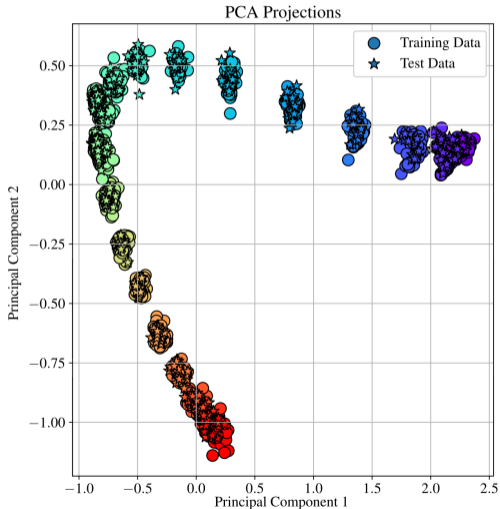
PCA precision



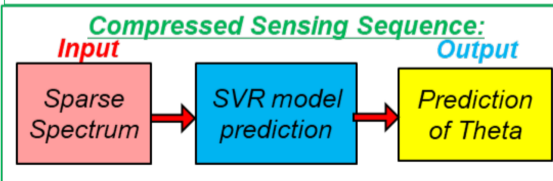
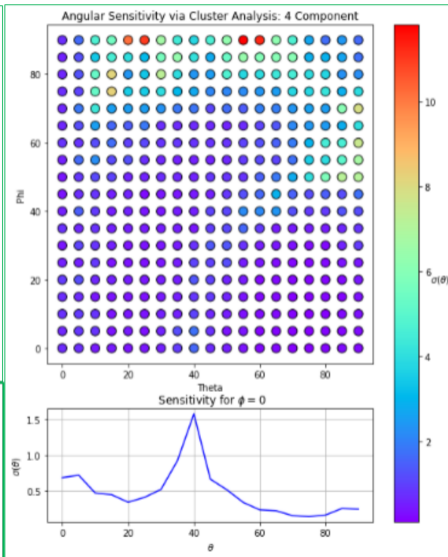
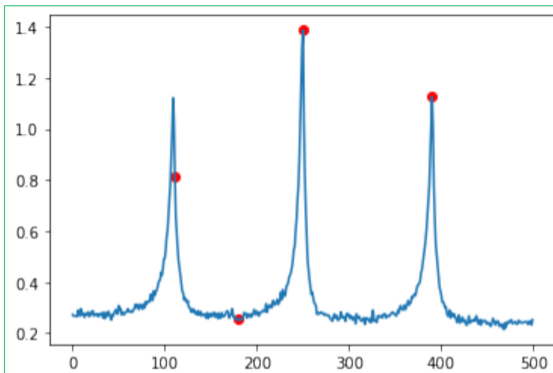
McKelvy *et al* Mach. Learn.: Sci. Technol. 4 (2023) 045048

PCA precision

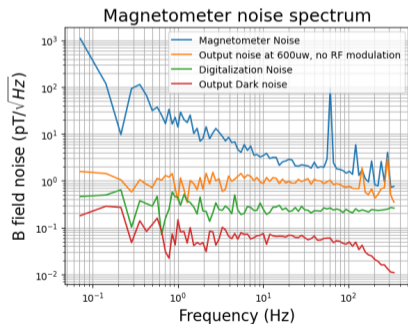
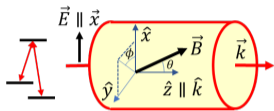
Summary of Model Performance in Determining θ with Classical PCA



Smart sensing



Compass summary



- 2-3 pT/ $\sqrt{\text{Hz}}$ sensitivity
- measures B-field vector $(10^{-2})^\circ / \sqrt{\text{Hz}}$ at optimum
- operates at the Earth magnetic field

McKelvy *et al* “Technical limits of sensitivity for EIT magnetometry” *Applied Optics* Vol. 62, No. 24, 6518

McKelvy *et al* “Application of kernel principal component analysis for optical vector atomic magnetometry” *Mach. Learn.: Sci. Technol.* 4 (2023) 045048

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