



2014 IEEE Magnetics Society Distinguished Lecture Series

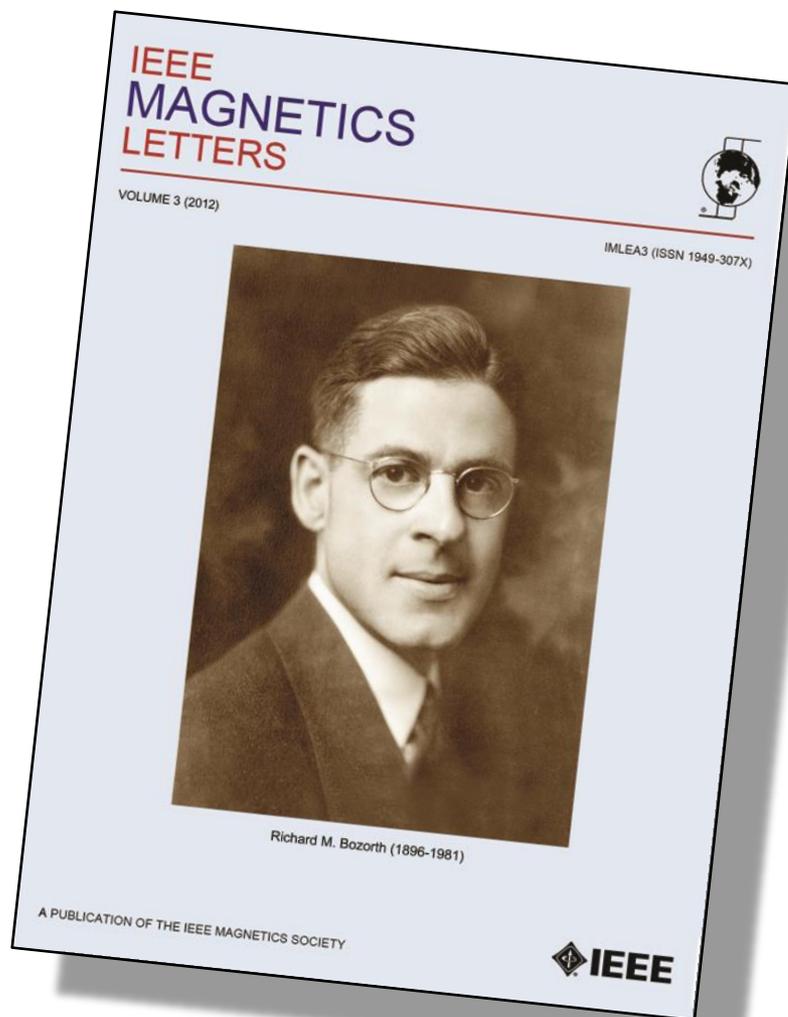
Opportunities and Challenges in Two-Dimensional Magnetic Recording

Jon Coker
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- **Introduction**
- **Magnetic Recording Models in Two Dimensions**
- **Testing in Two Dimensions**
- **Magnetic Signal Processing in Two Dimensions**
- **Questions**

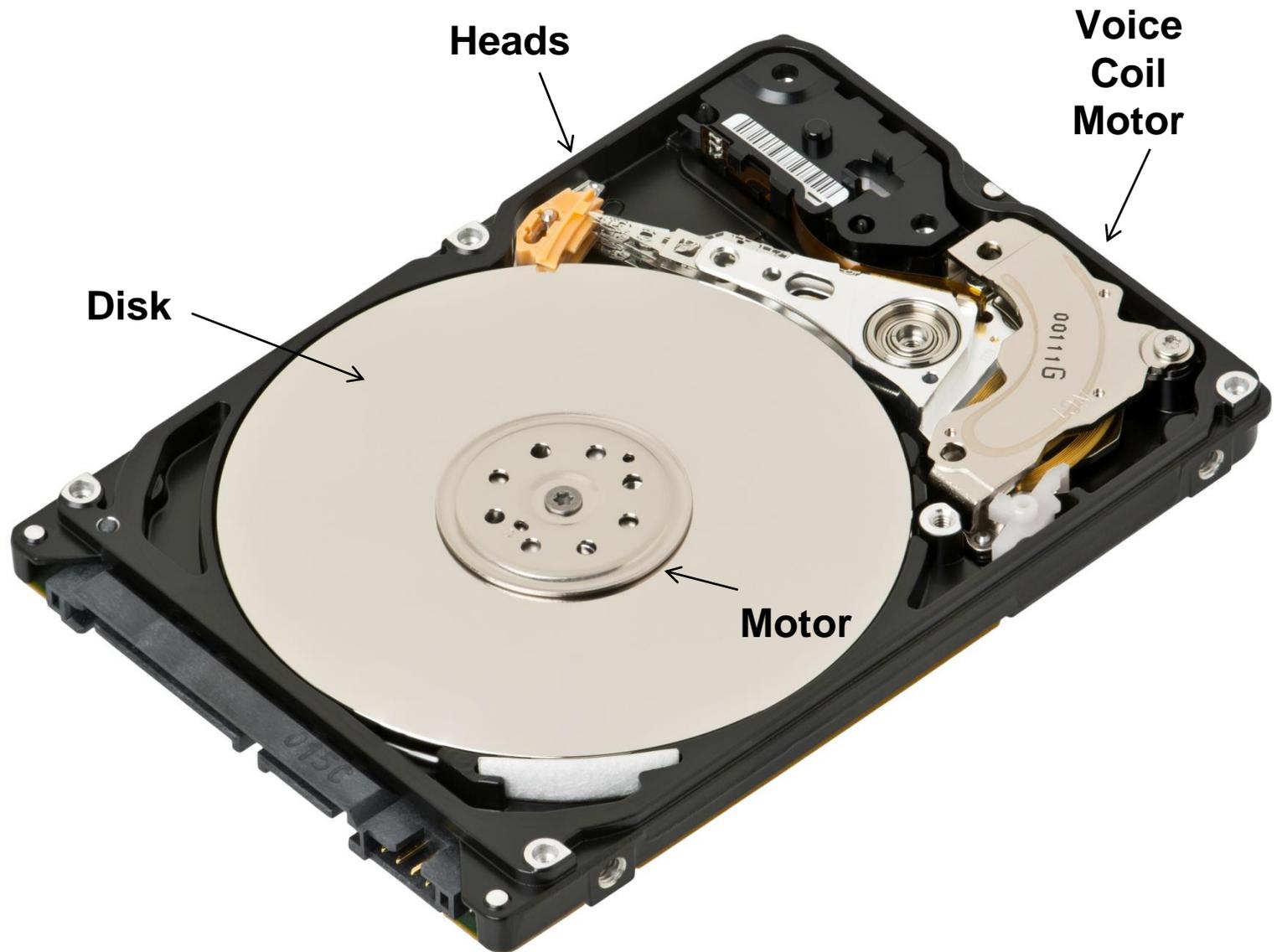


Image from wikipedia.com

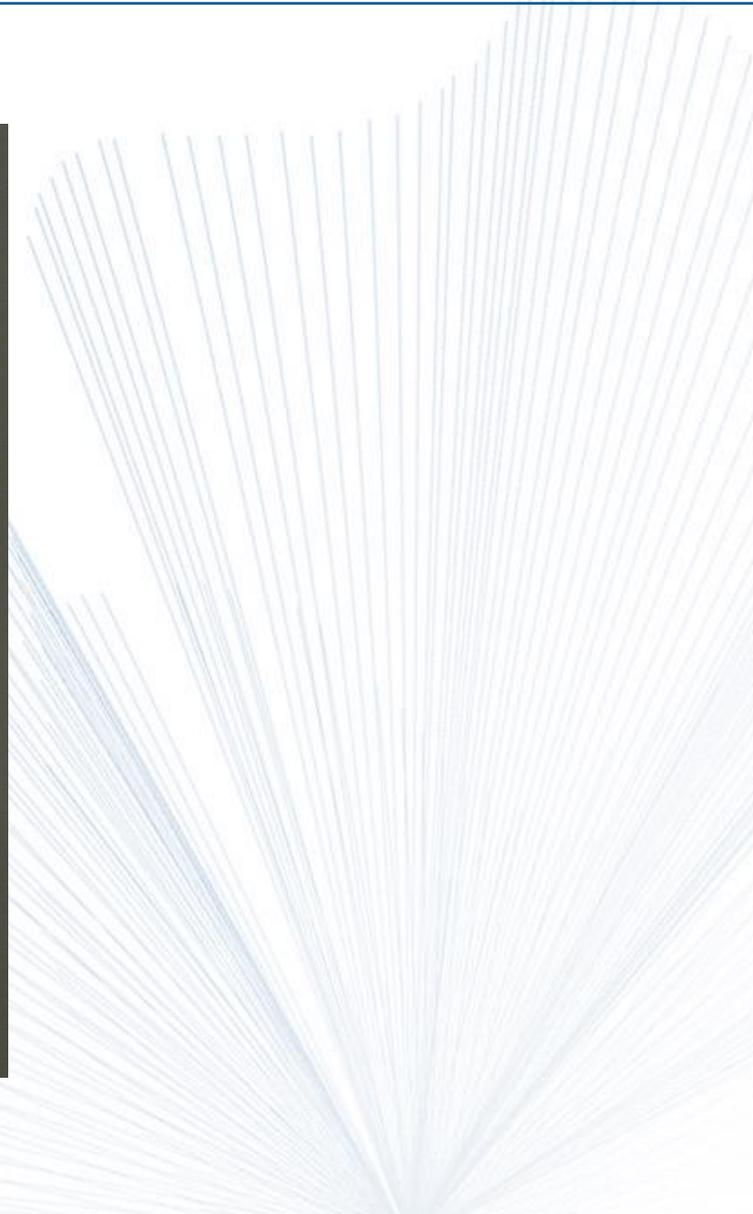
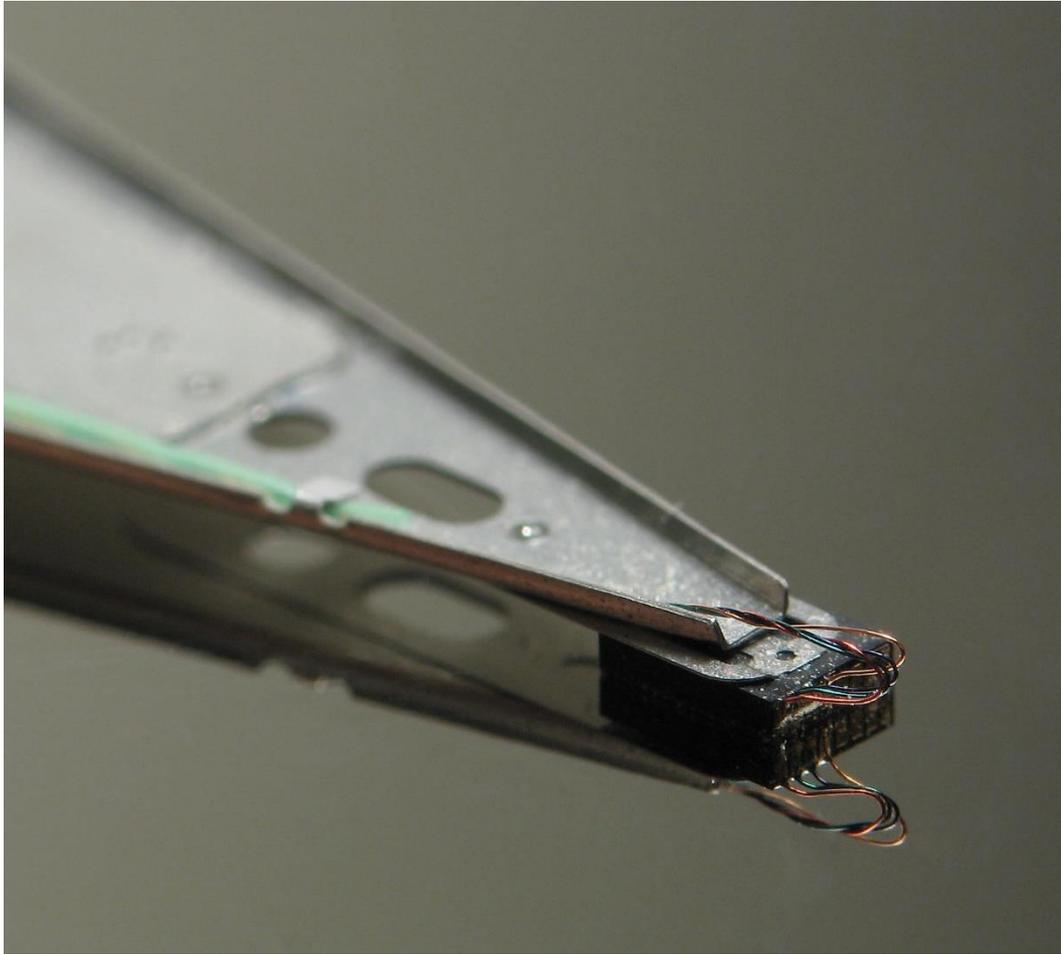
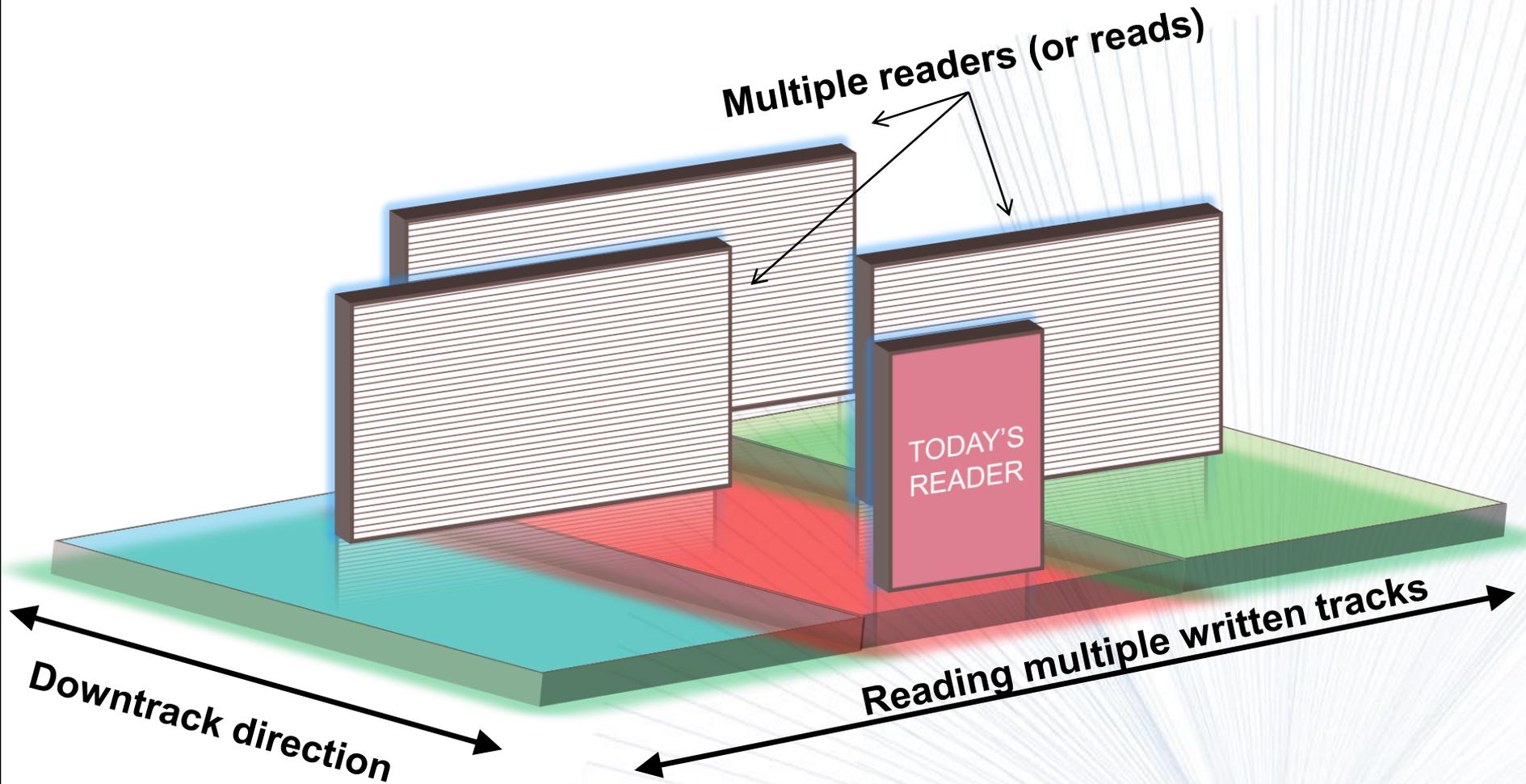
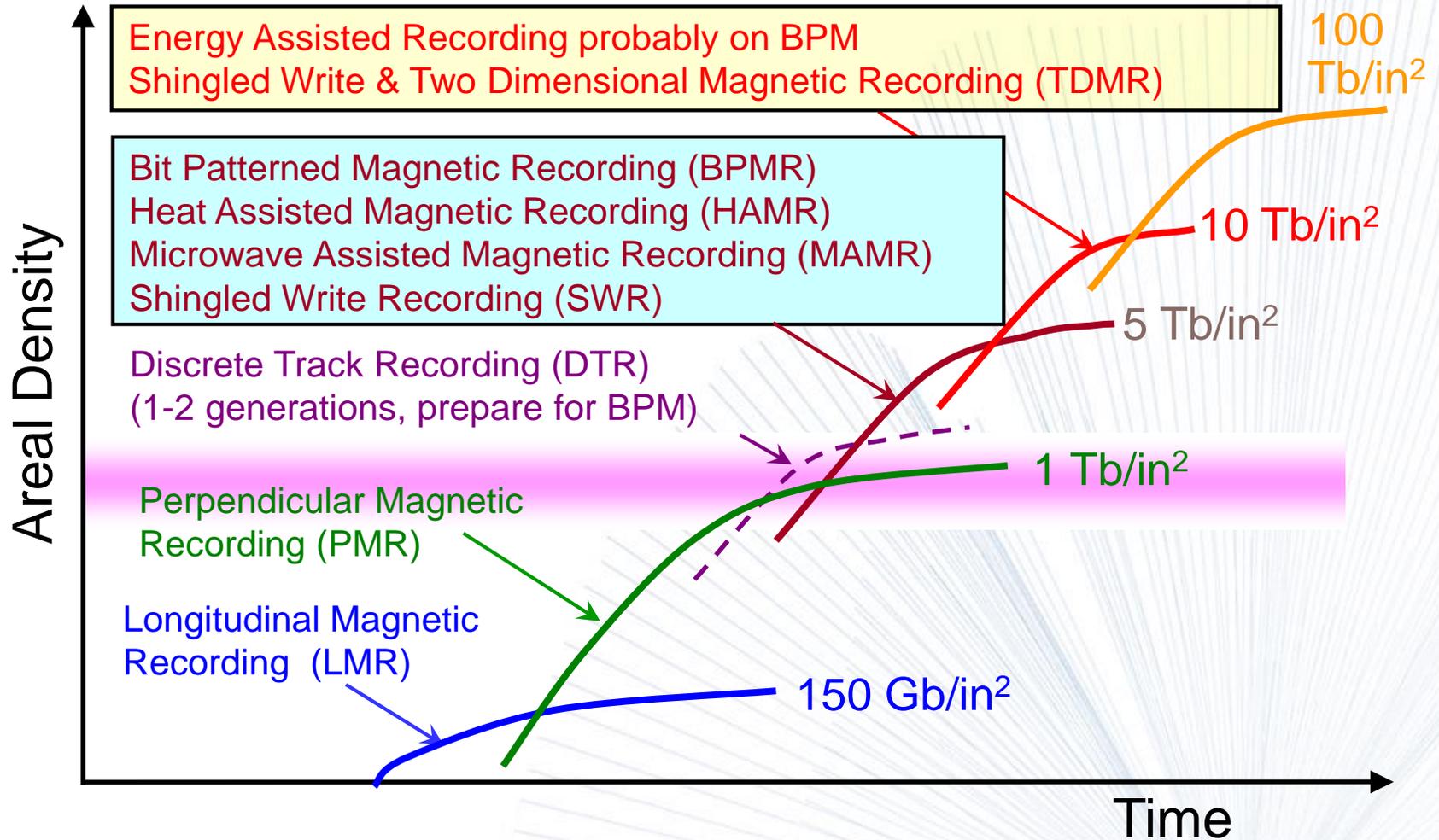


Image from wikipedia.com

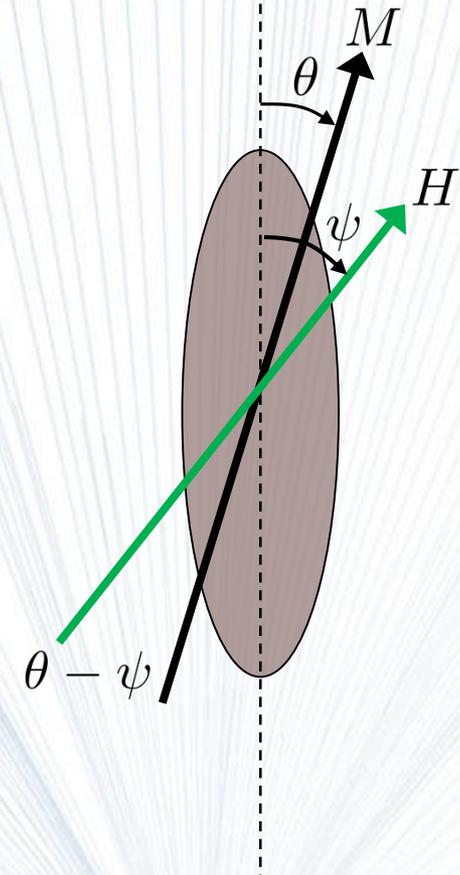
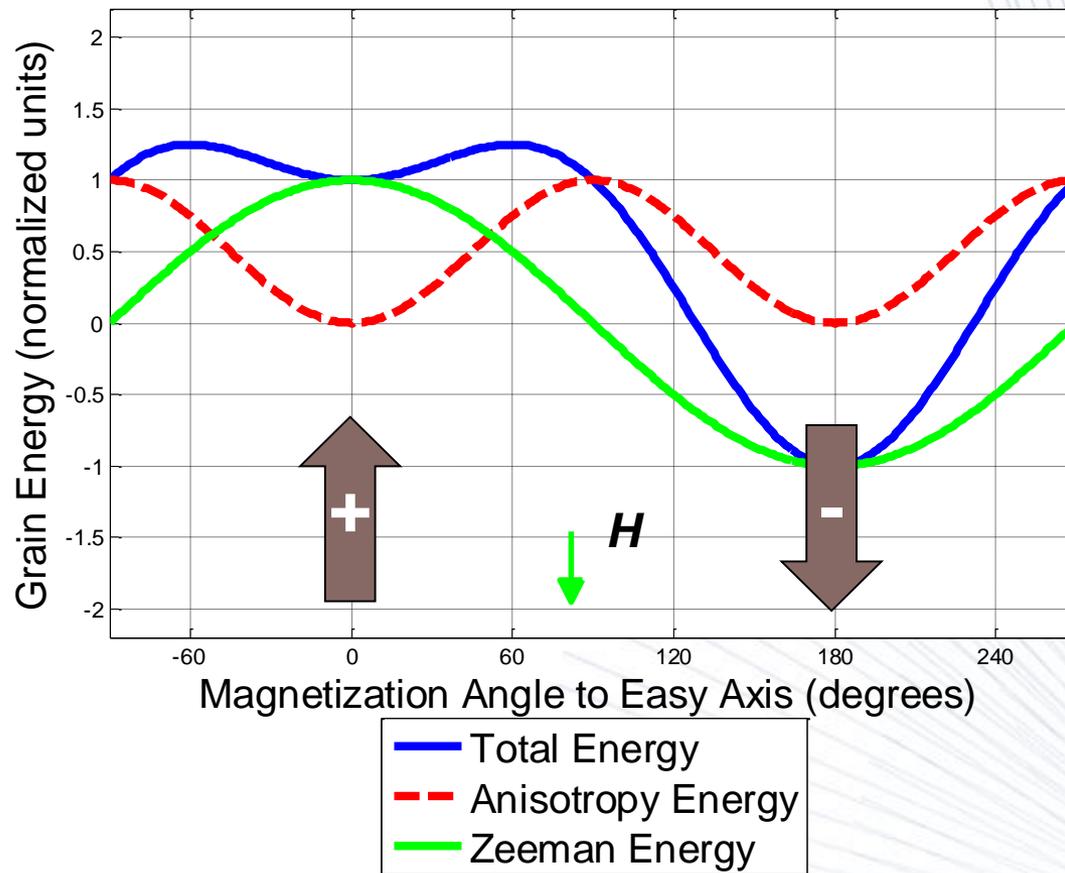


Detecting data in this configuration is a multidisciplinary 2D problem.

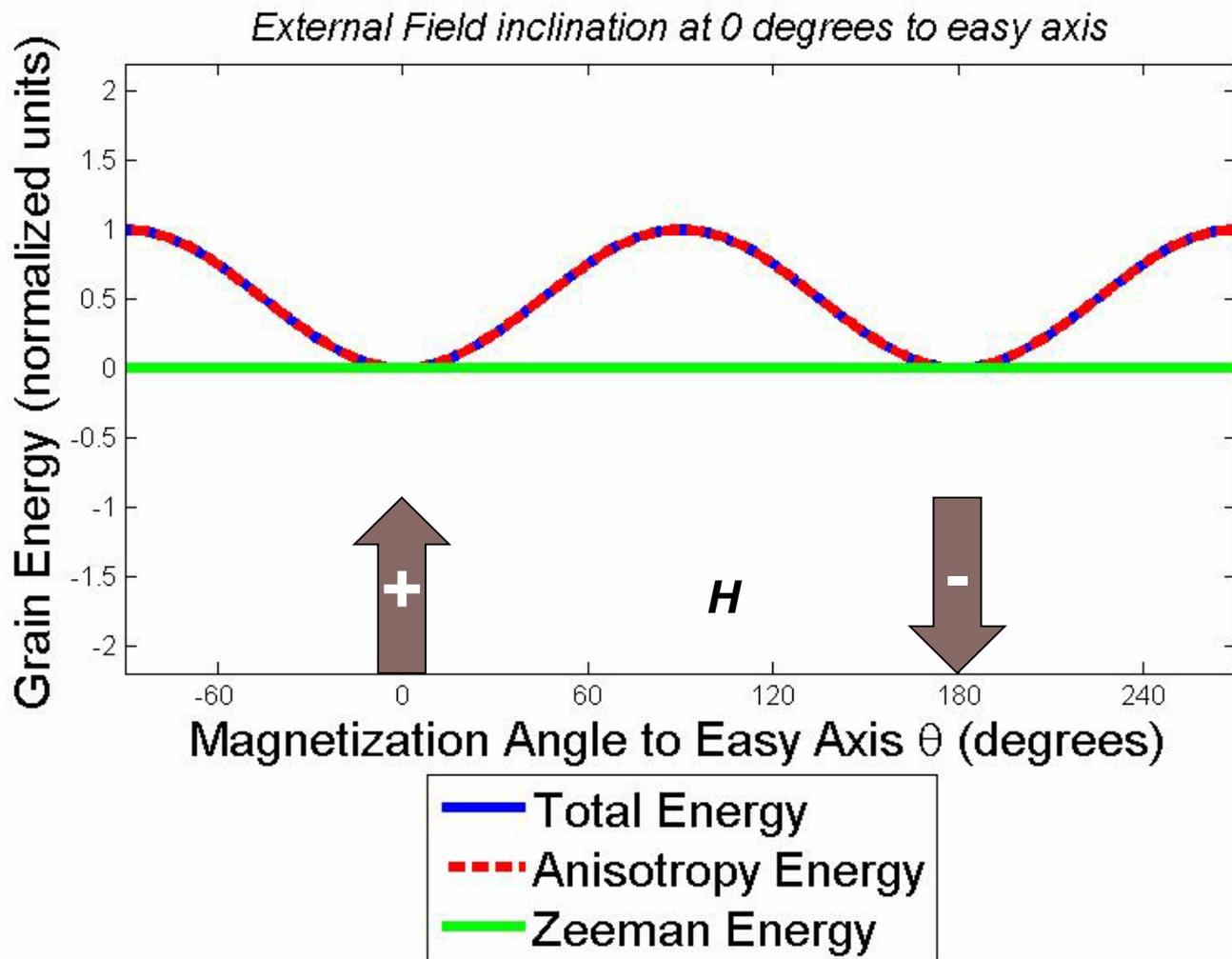


Further reading: Y. Shiroishi, Intermag 2009, FA-01

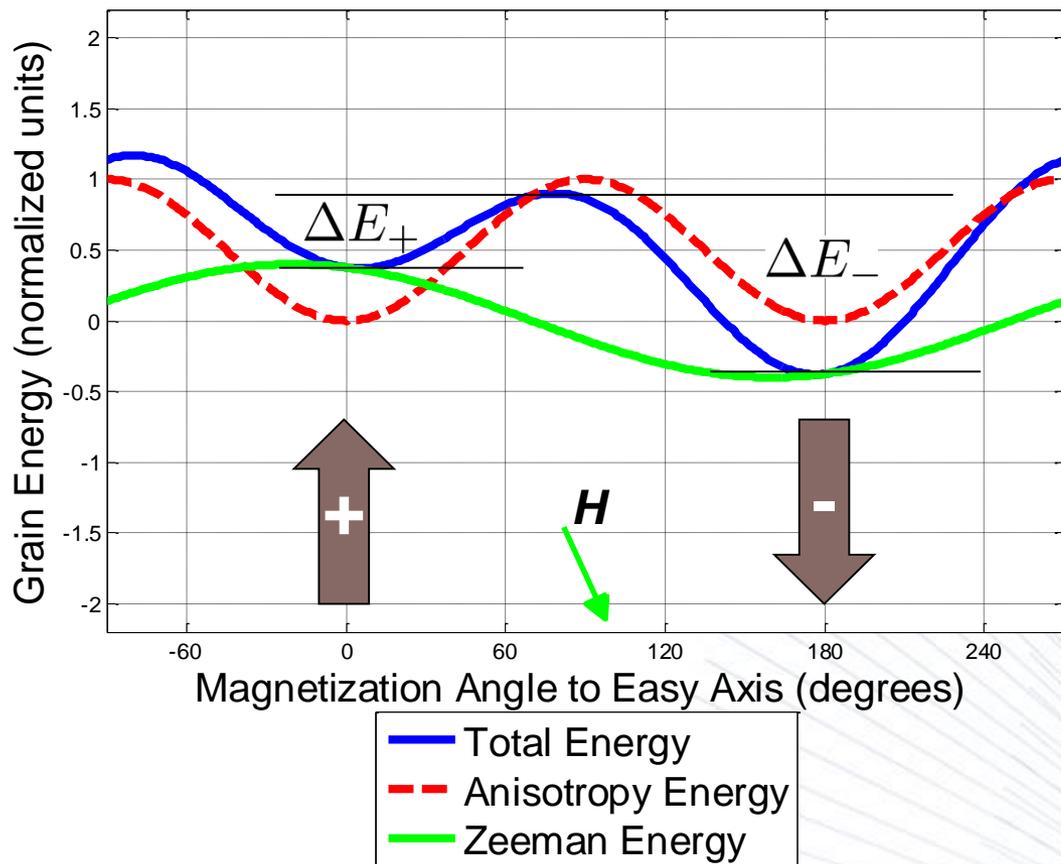
$$E = \overbrace{K_u V \sin^2(\theta)}^{\text{ANISOTROPY}} + \dots - \overbrace{H M V \cos(\theta - \psi)}^{\text{ZEEMAN}} + \overbrace{E_T}^{\text{THERMAL}}$$



Suggested reading: Zhang and Bertram, *IEEE Trans. on Magnetics*, Vol 34, No. 5 1998
R. Wood, *IEEE Trans. on Magnetics*, Vol. 41, No. 1, 2009



$$\Delta E_{\pm} \approx K_u V (1 \mp H/H_n)^\alpha$$



$$P_{\pm}(H) \approx e^{-\Delta E_{\pm}/kT}$$

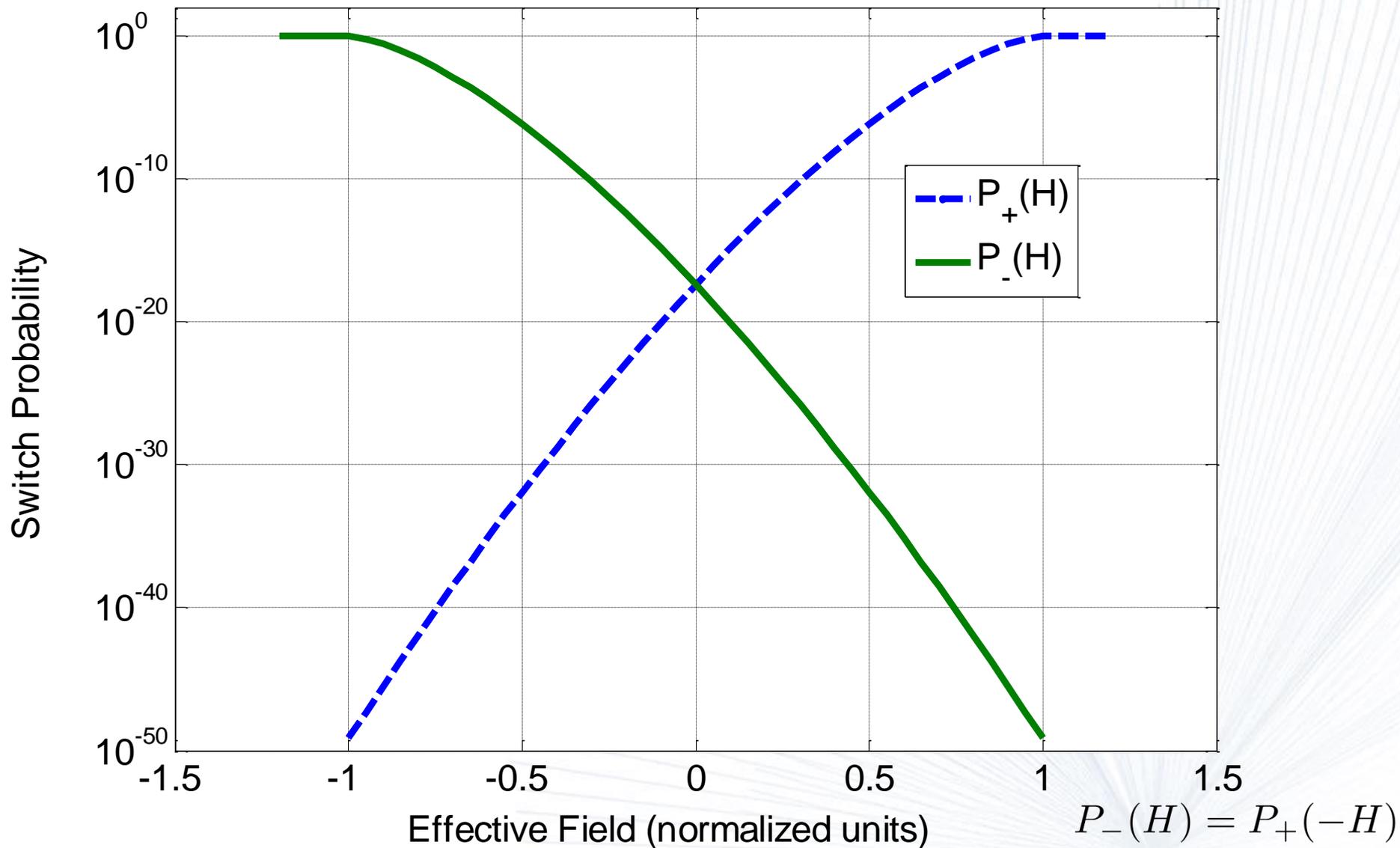
Probability that a grain switches in a given moment

$$\tau_0 = 1/f_0$$

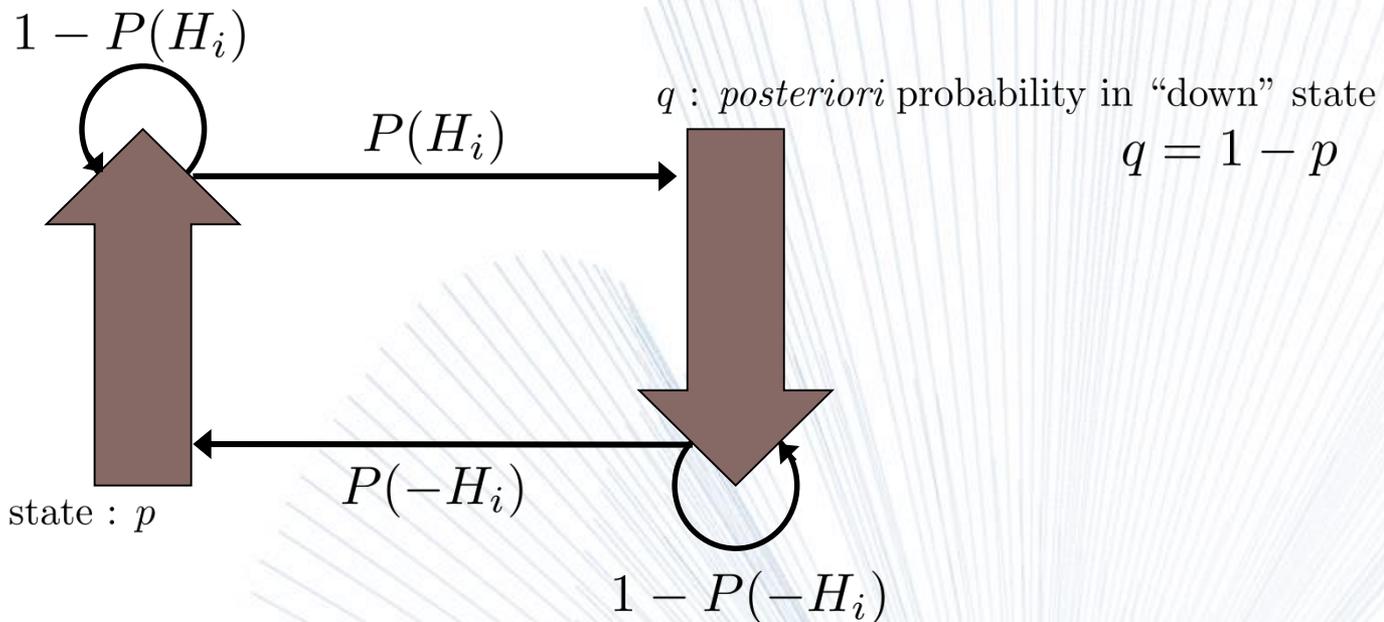
Suggested reading: Zhang and Bertram, *IEEE Trans. on Magnetics*, Vol 34, No. 5 1998
 Victora and Chen., *Proc. of the IEEE*, Vol. 96, No. 11, Nov 2008
 Victora, *Physical Review Letters*, Vol. 63, No. 4, July 1989

Exemplary Switch Probabilities vs. Field Strength

Plotted for $K_u V / kT = 40$

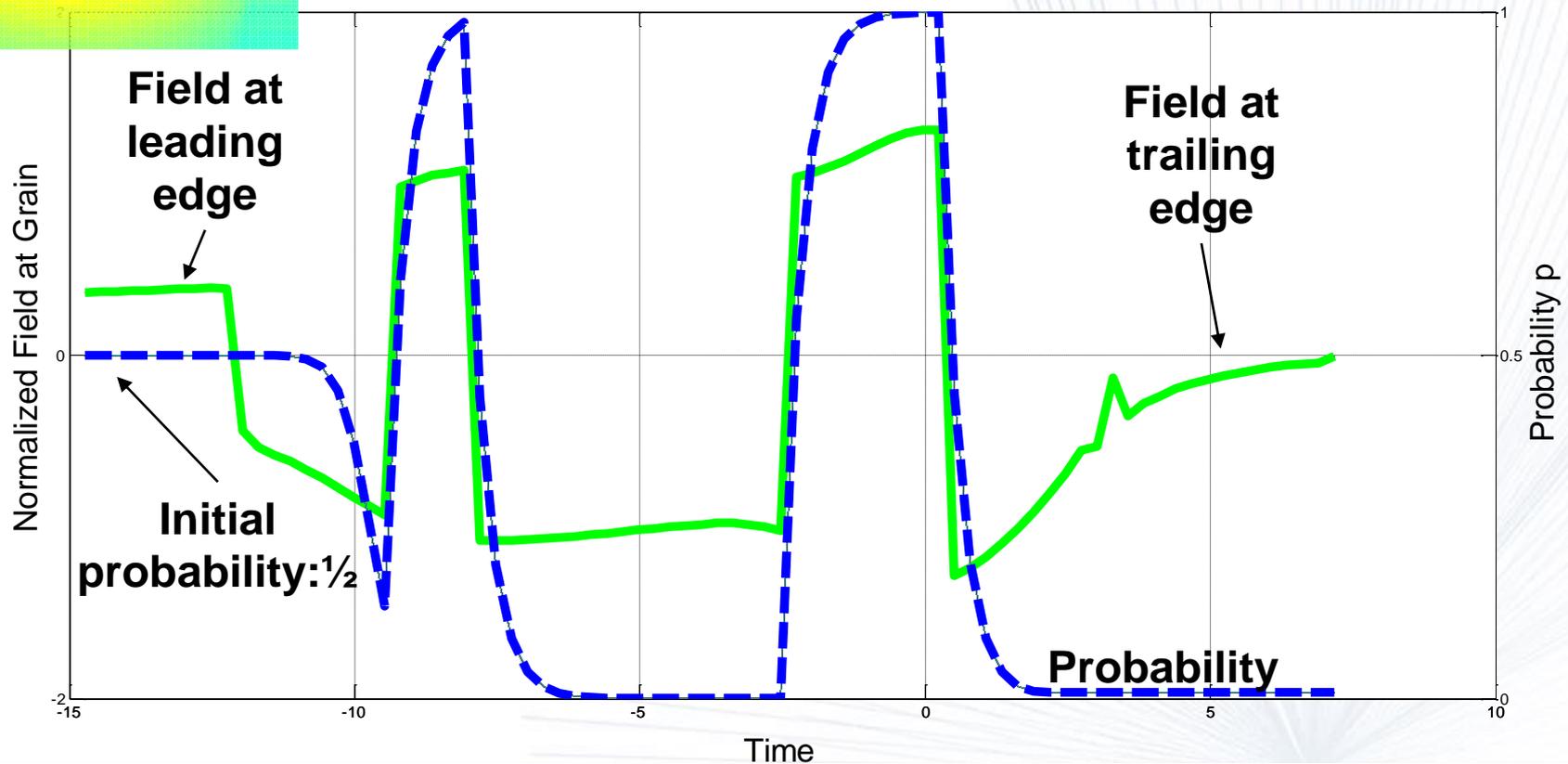


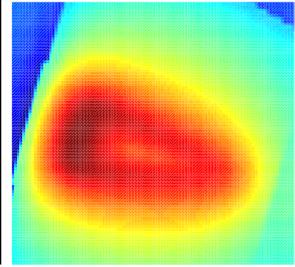
A Discrete Time Markov Model for Two-State Magnetization



$$\begin{pmatrix} p_{i+1} \\ q_{i+1} \end{pmatrix} = \underbrace{\begin{pmatrix} 1 - P(H_i) & P(-H_i) \\ P(H_i) & 1 - P(-H_i) \end{pmatrix}}_{\mathbf{T}(H_i)} \begin{pmatrix} p_i \\ q_i \end{pmatrix}$$

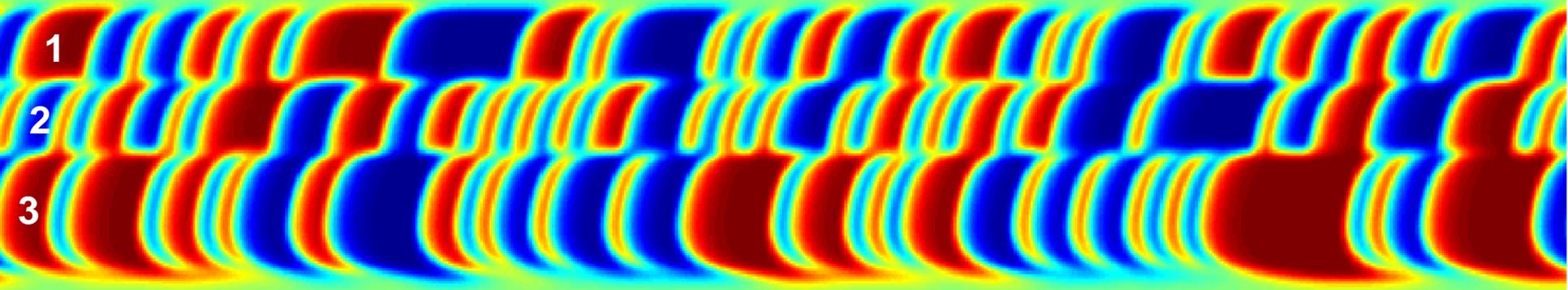
$$H_{eff}(x - vt, y) i_w(t)$$





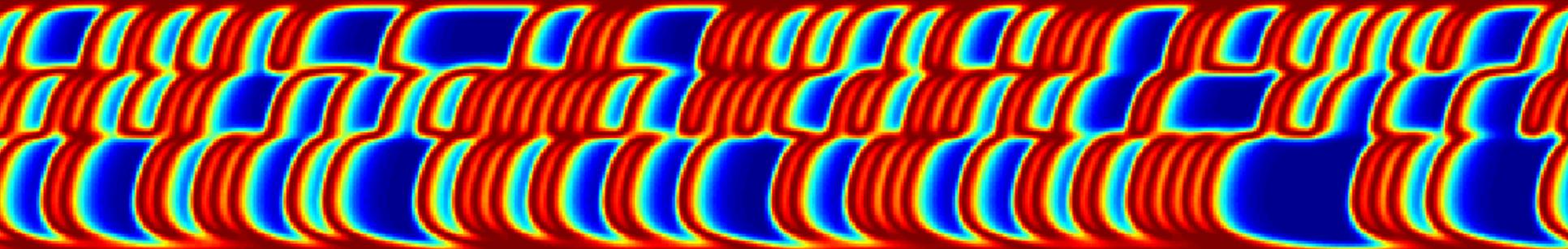
$$H_{eff}(x - vt, y) i_w(t)$$

SIGNAL

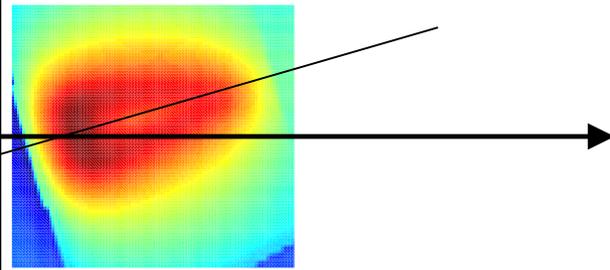


$$\sigma_M^2(x, y) = 4(1 - p(x, y))p(x, y) \quad \longleftrightarrow \quad \bar{M}(x, y) = 2p(x, y) - 1$$

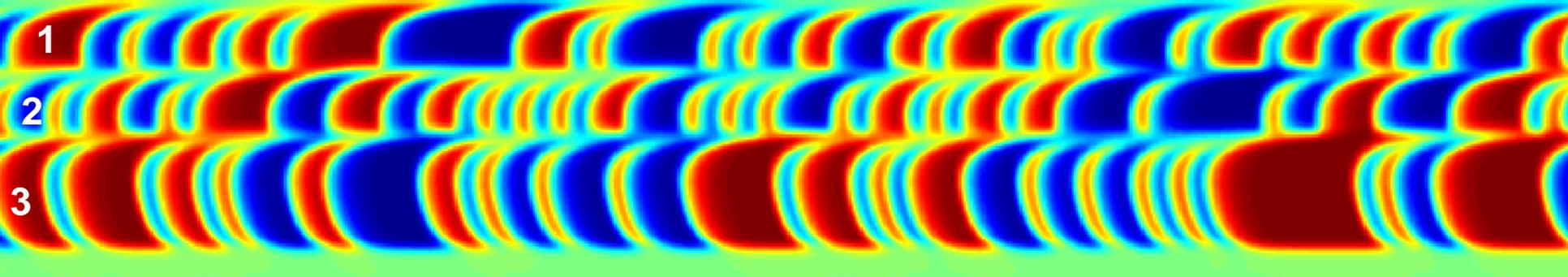
NOISE



Wrong-Way Shingling with Different Head Skew

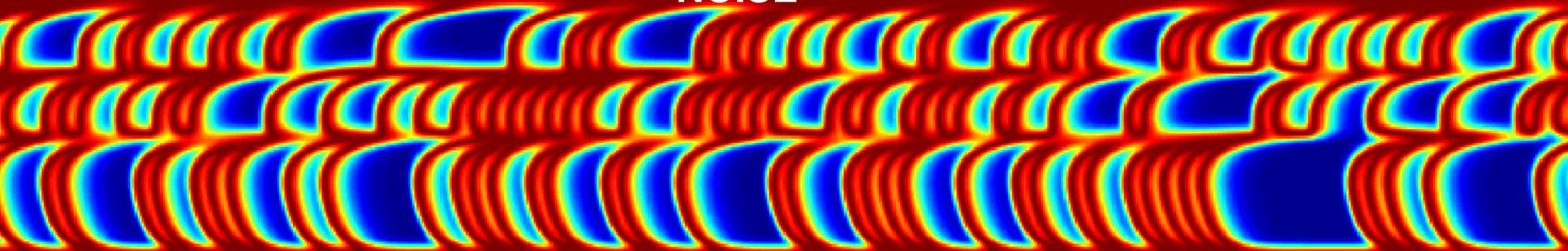


SIGNAL

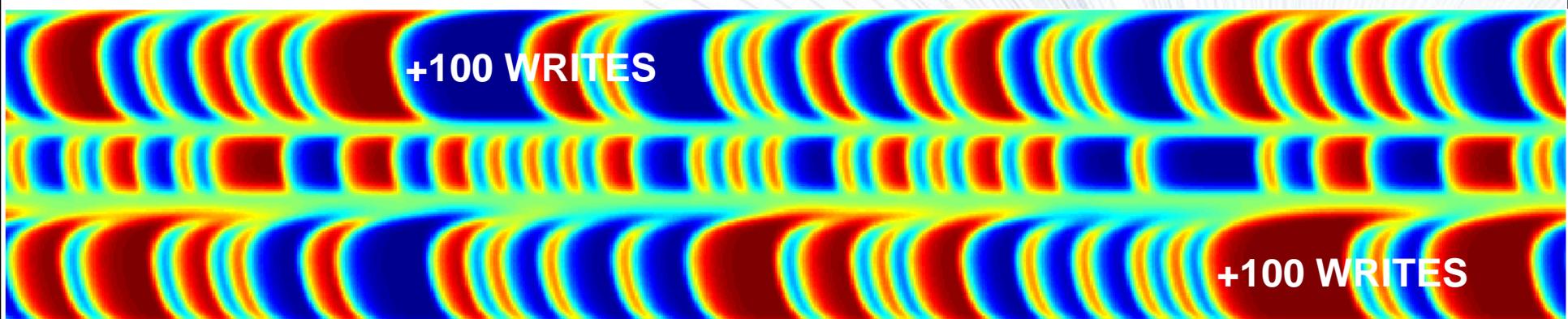
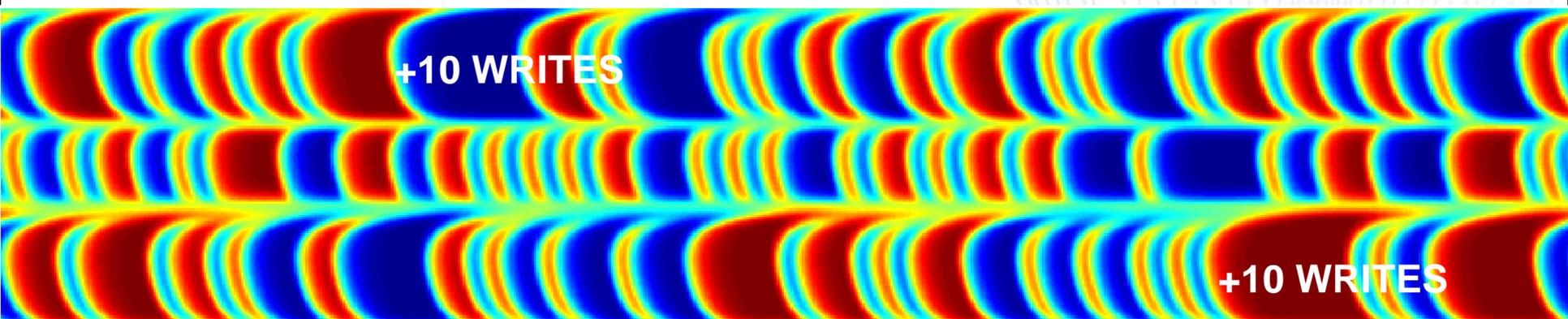
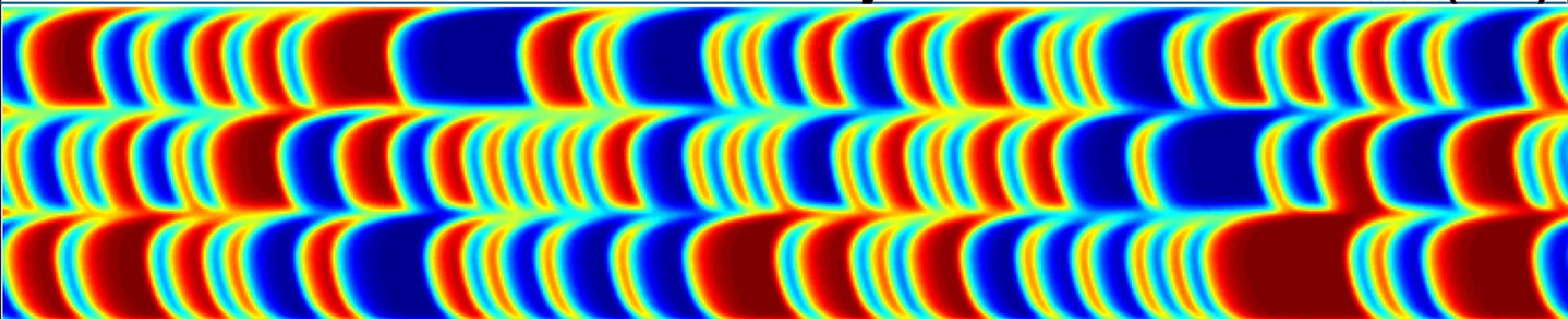


$$\sigma_M^2(x, y) = 4(1 - p(x, y))p(x, y) \quad \begin{array}{c} \uparrow \\ \downarrow \end{array} \quad \bar{M}(x, y) = 2p(x, y) - 1$$

NOISE

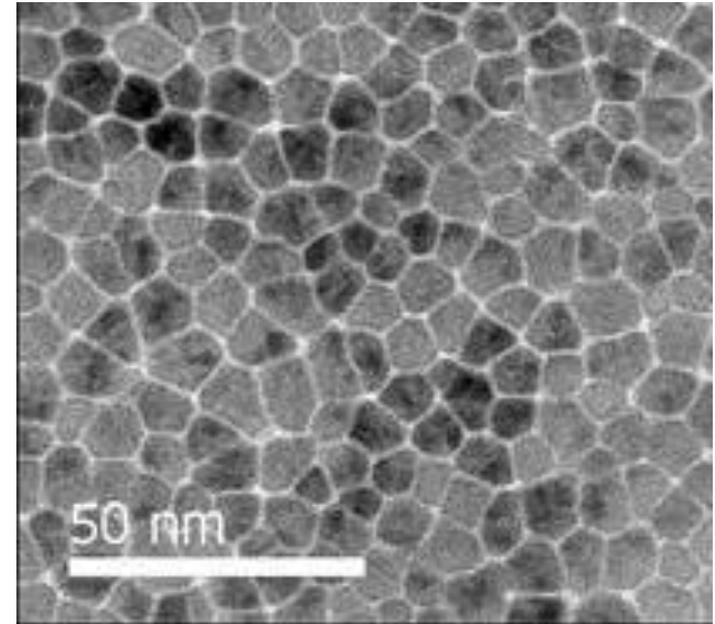
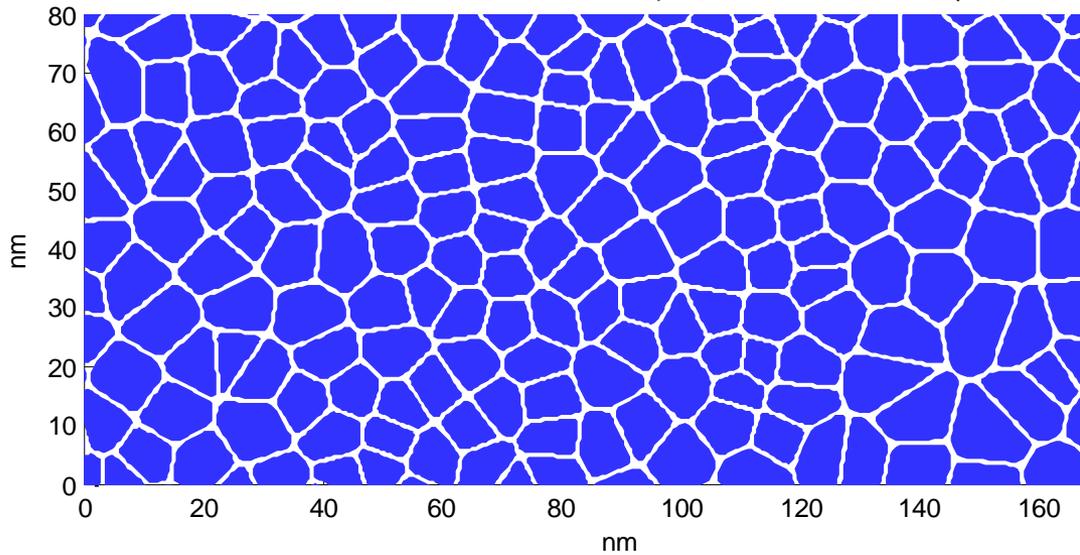


Standard PMR probability maps with Adjacent Track Interference (ATI)



(but: grains have finite and nonconstant volume)

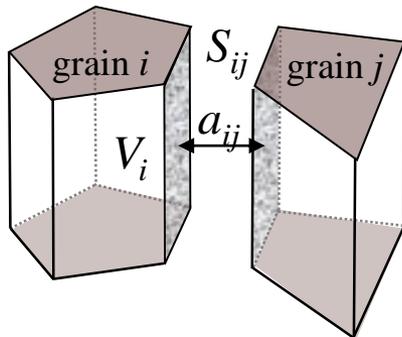
Voronoi Media



Suggested reading: Chan and Elidrissi, IEEE Trans. on Magnetics, Vol. 49, Issue 6, 2013

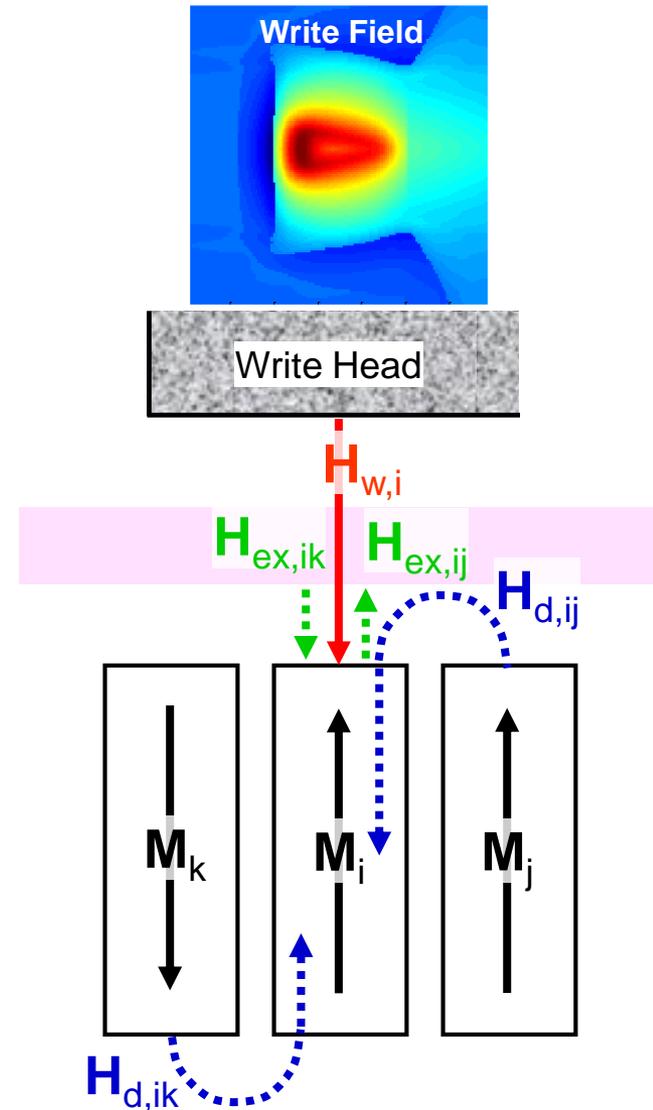
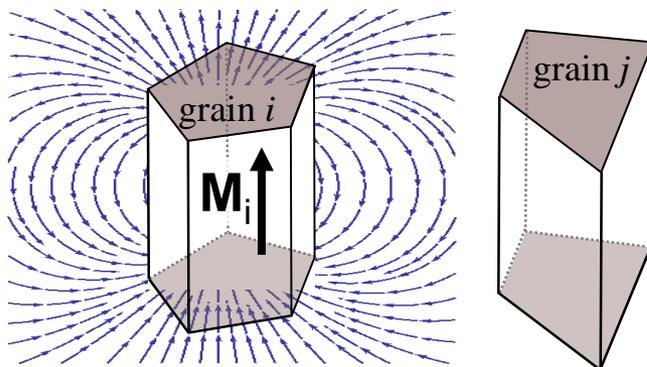
Intergranular Exchange

Neighbors tend to magnetize in *same* direction



Demag Fields

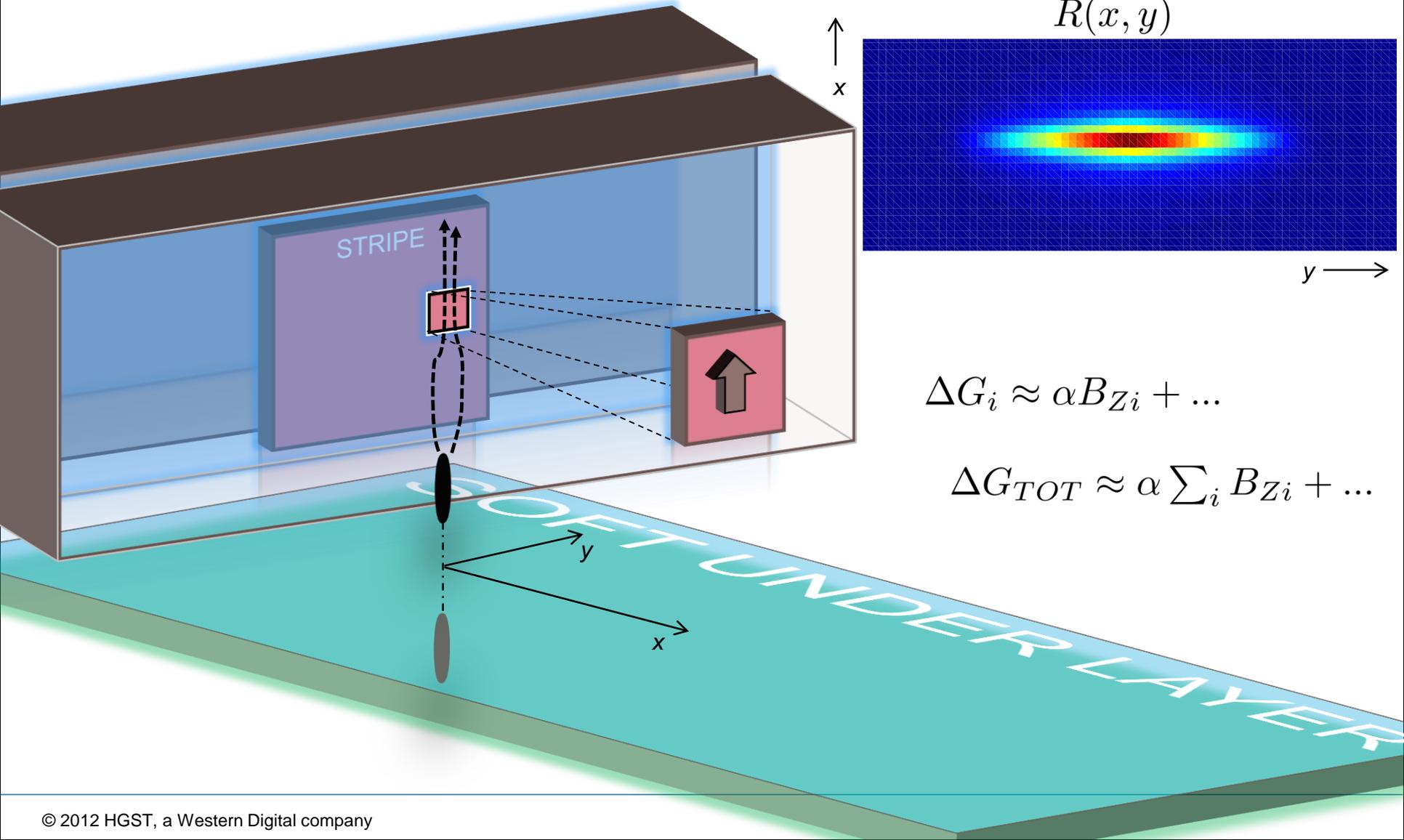
Neighbors tend to magnetize in *opposite* direction



Suggested reading: S. J. Greaves et al., *Journal of Magnetism and Magnetic Materials* 287, 2005.

The Read Sensitivity or “Impulse Response” Function

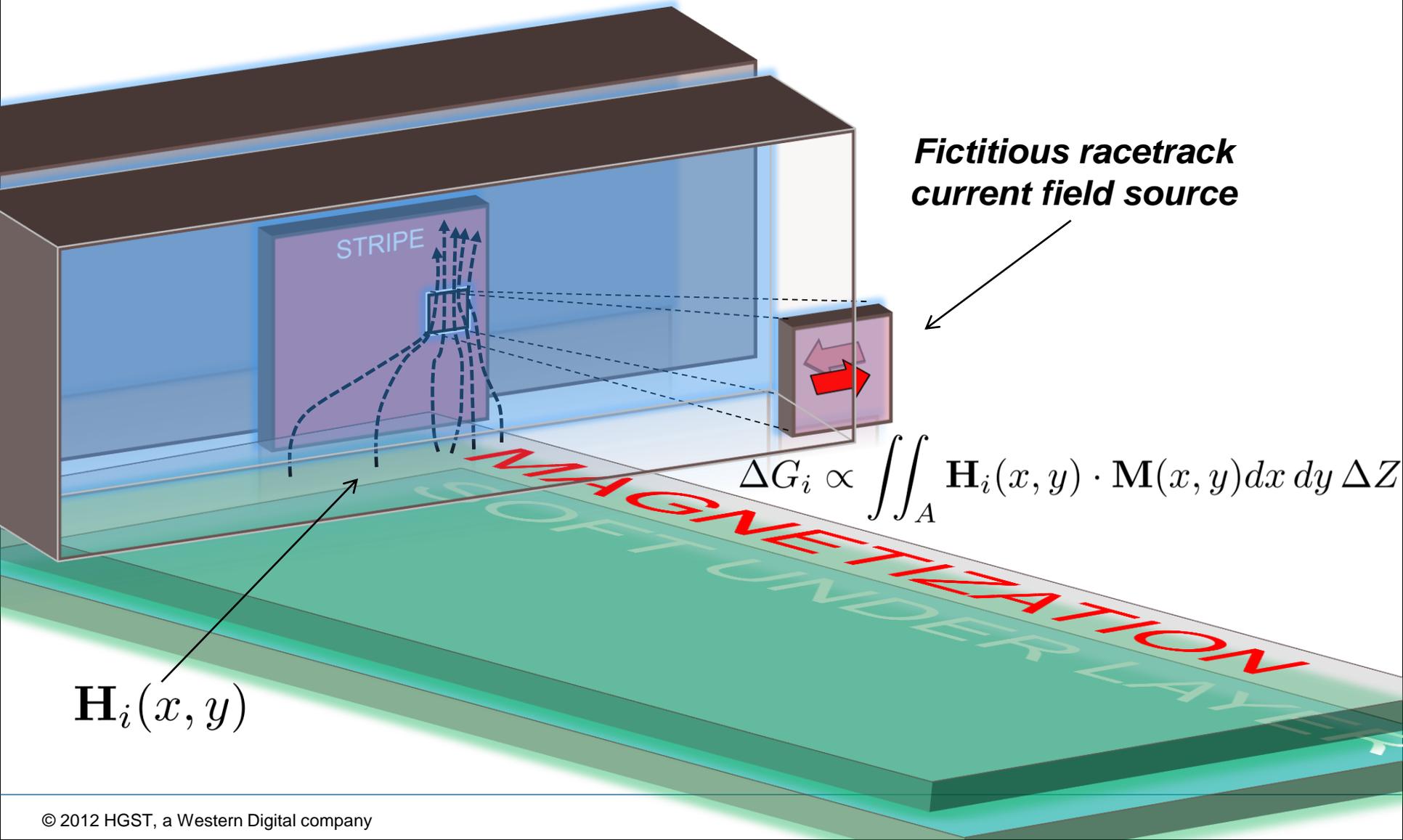
Suggested reading: Yuan and Bertram, *IEEE Trans. on Magnetics*, Vol. 30, Issue 3, 1994
Wood and Wilton, *IEEE Trans. on Magnetics*, Vol. 44, Issue 7, 2008



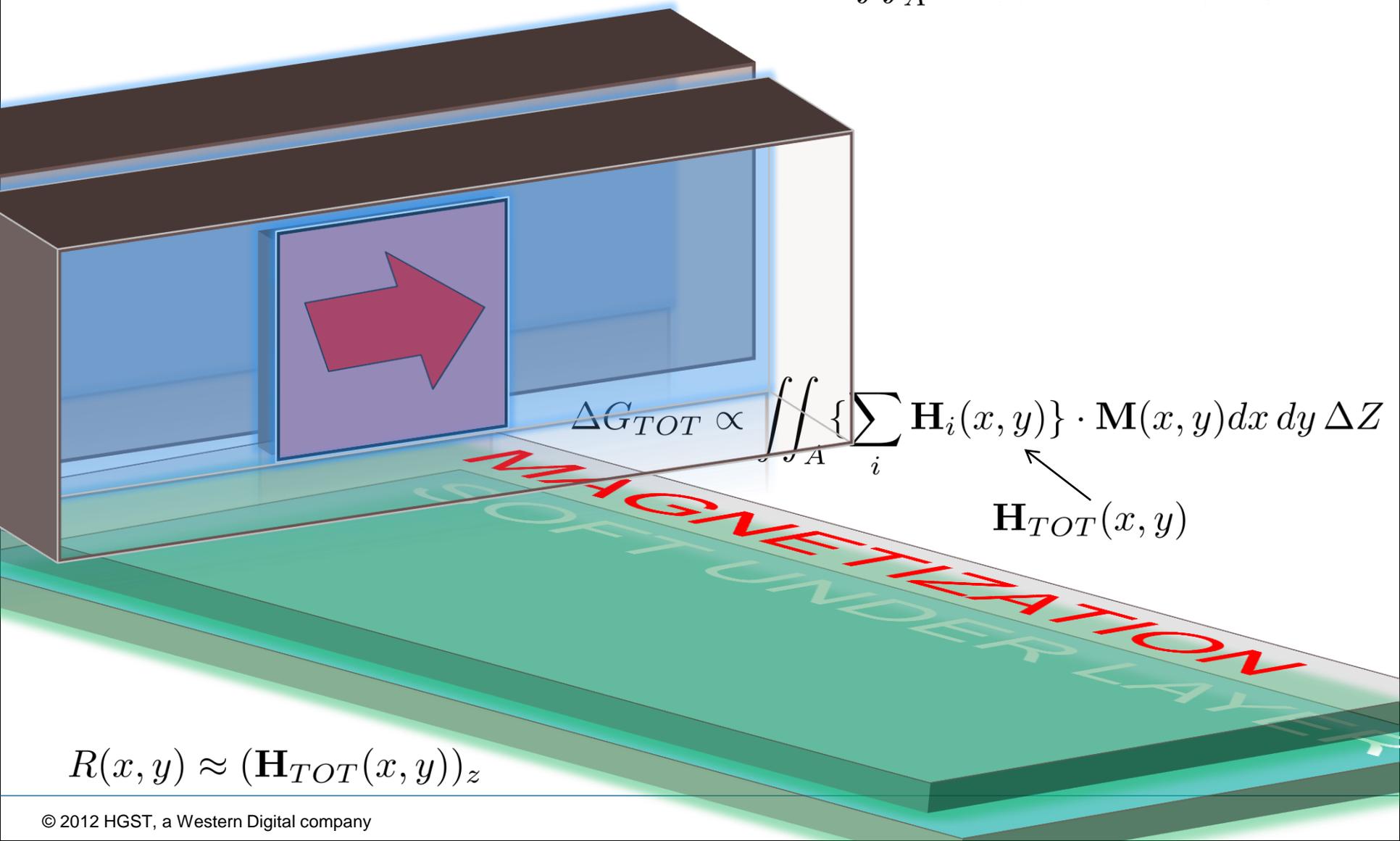
$$\Delta G_i \approx \alpha B_{Zi} + \dots$$

$$\Delta G_{TOT} \approx \alpha \sum_i B_{Zi} + \dots$$

Suggested reading: N. Smith, *IEEE Trans. on Magnetics*, Vol. 29, No. 5, 1993
Litvinov and Khizroev, *Journal of Applied Physics*, Vol. 97, 2005



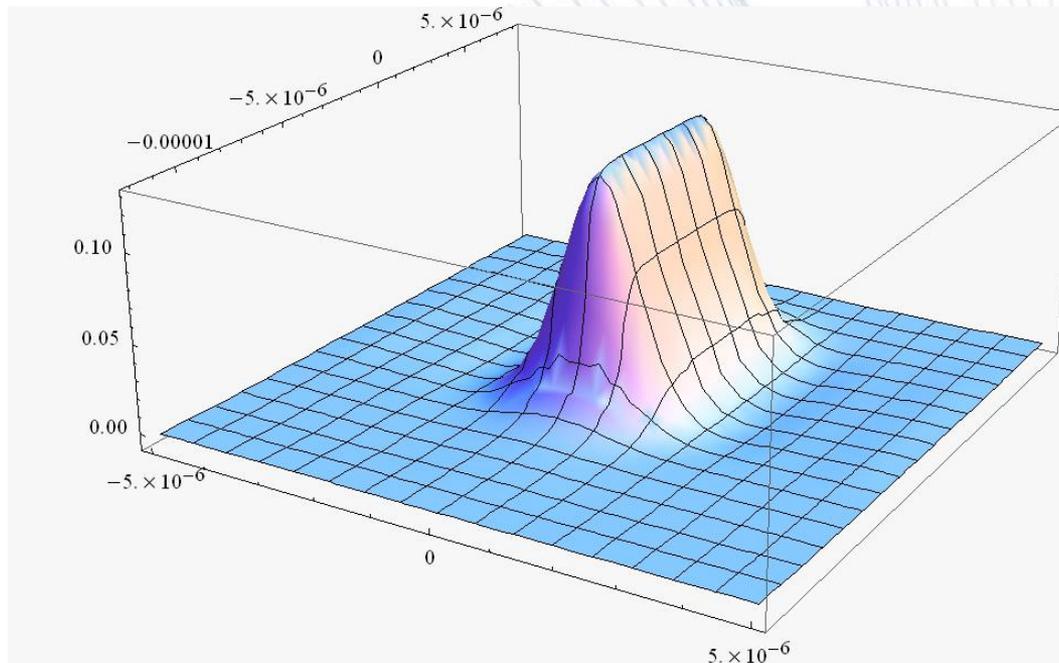
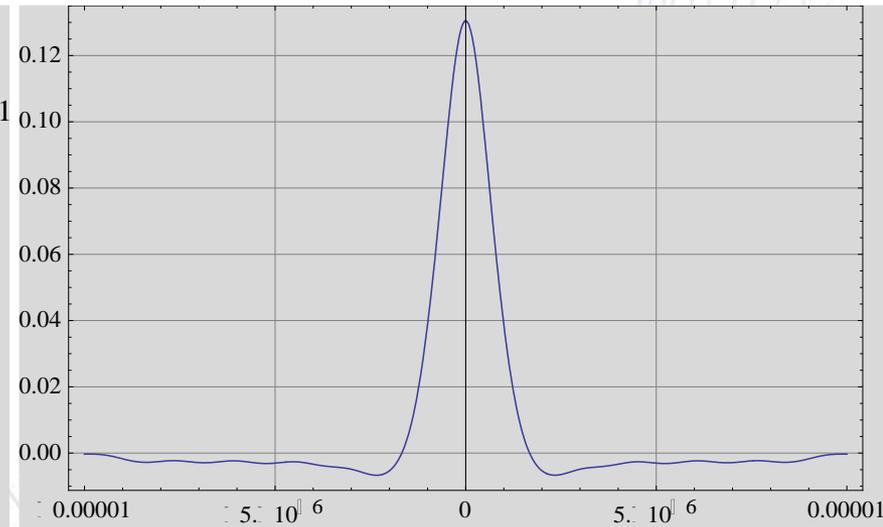
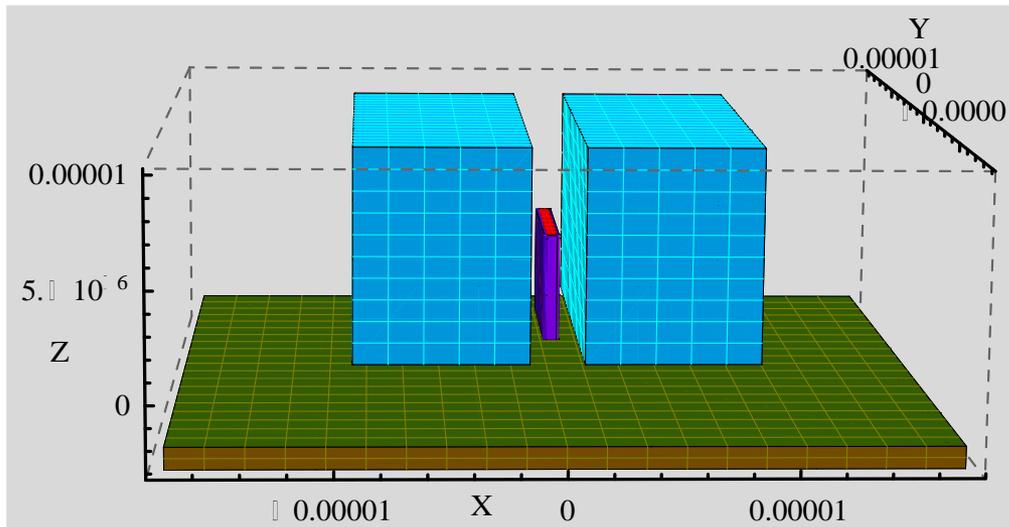
$$\Delta G_i \propto \iint_A \mathbf{H}(x, y) \cdot \mathbf{M}(x, y) dx dy \Delta Z$$



$$\Delta G_{TOT} \propto \iint_A \left\{ \sum_i \mathbf{H}_i(x, y) \right\} \cdot \mathbf{M}(x, y) dx dy \Delta Z$$

$\mathbf{H}_{TOT}(x, y)$

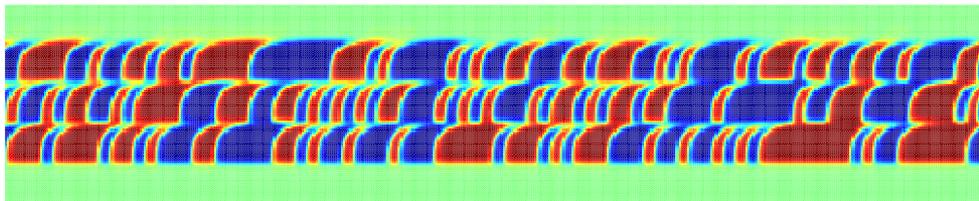
$$R(x, y) \approx (\mathbf{H}_{TOT}(x, y))_z$$



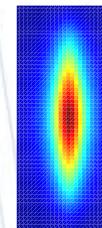
$R(x, y)$

Putting it all together for linear magnetic systems: generate readback signals by 2D convolution

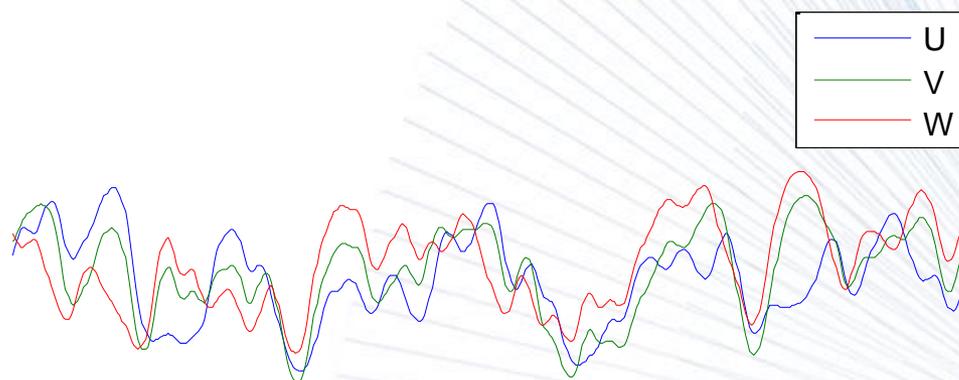
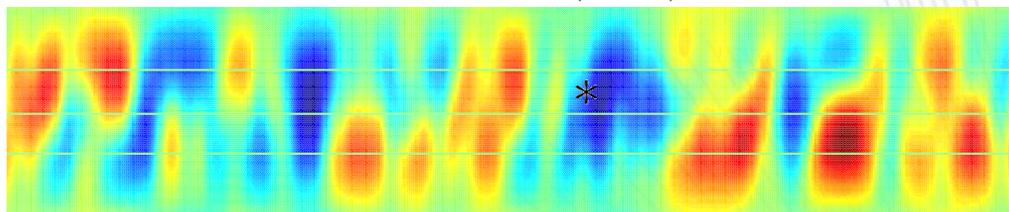
$$\bar{M}(x, y)$$



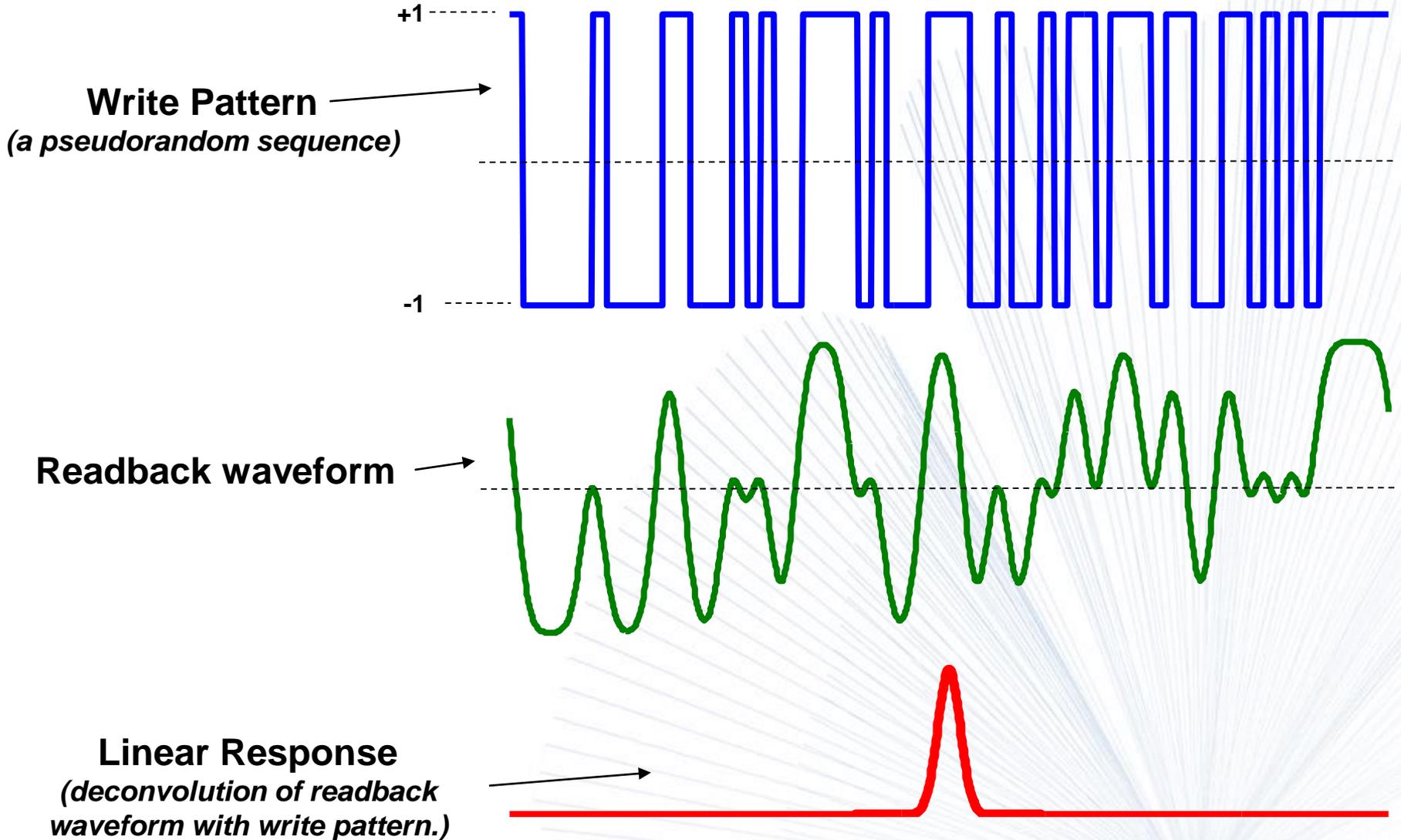
$$R(x, y)$$



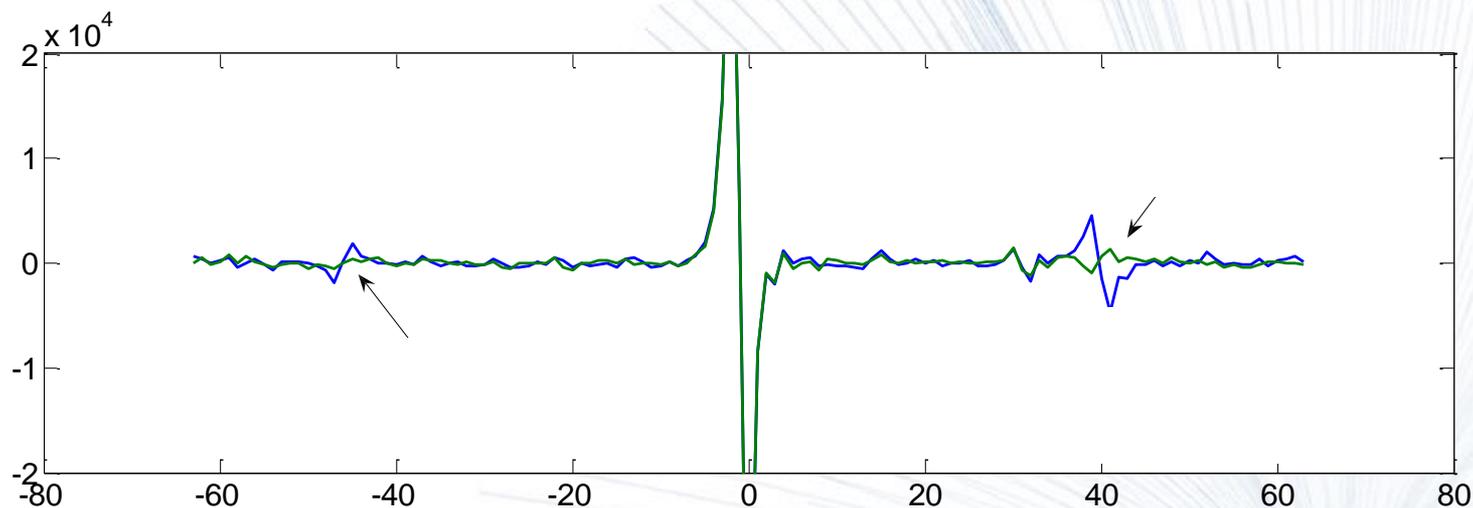
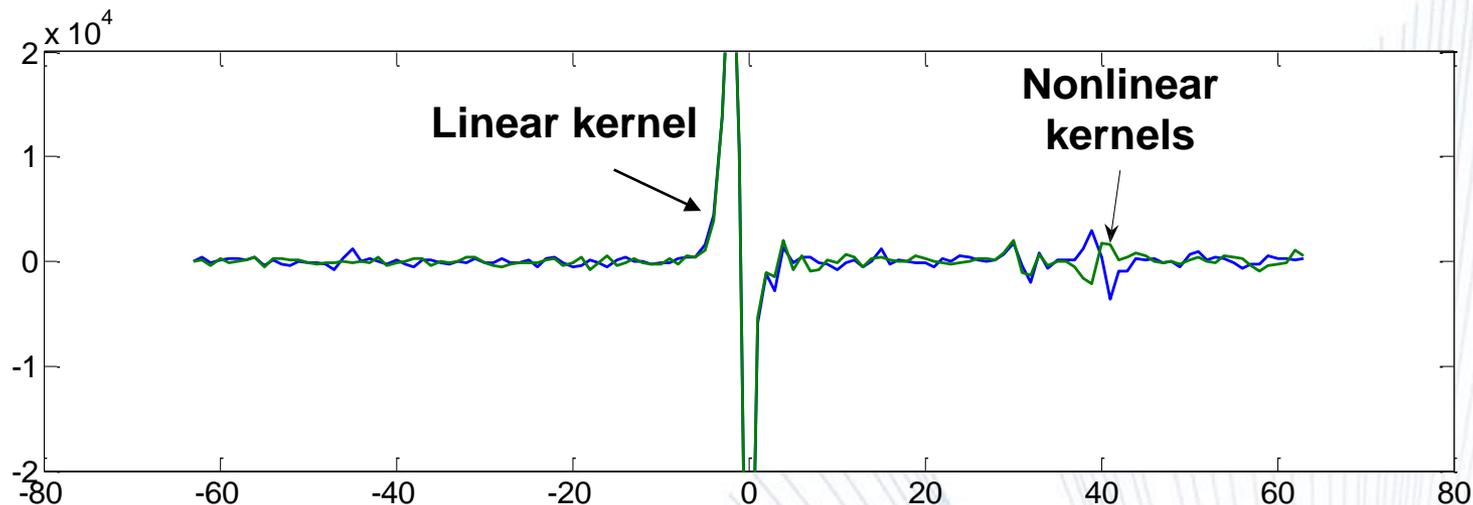
$$= X(x, y)$$



$$M(x, y) * R(x, y) = X(x, y)$$

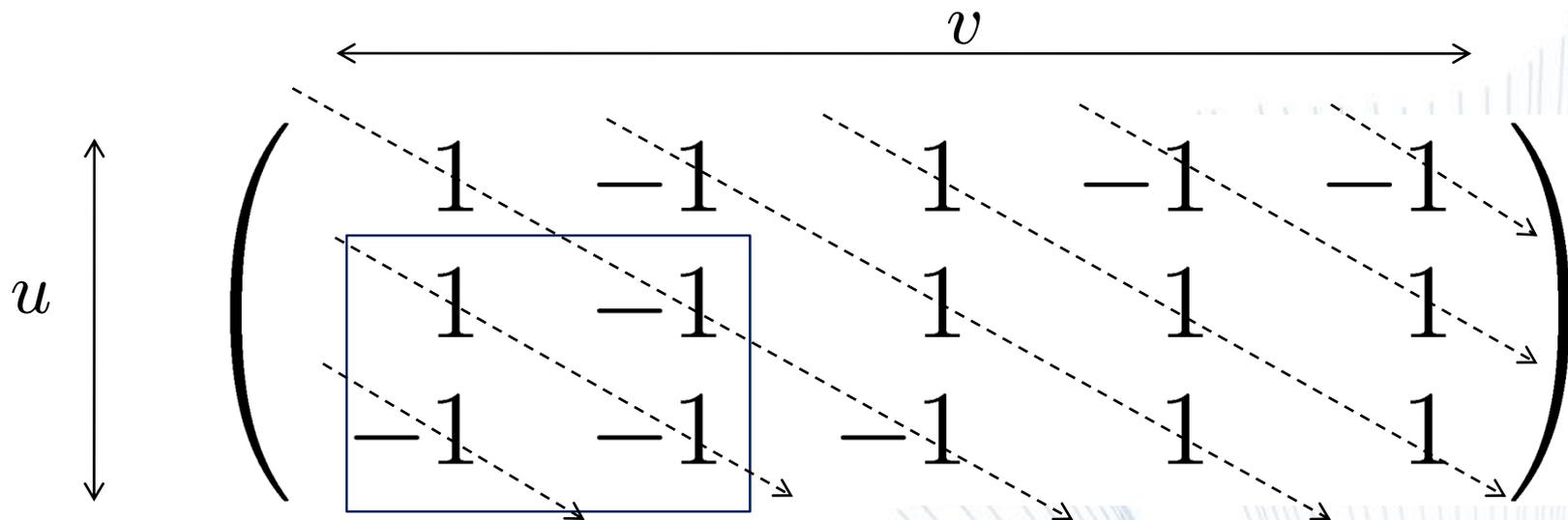


Typical Pseudorandom Sequence Test Results in One Dimension



These results are LMR recording technology, not PMR

Suggested reading: Palmer et al, *IEEE Trans. on Magnetics*, Vol. 24, Issue 6, 1988
Hermann, *IEEE Trans. on Magnetics*, Vol. 26, Issue 5, 1990



Can construct for any $uv = 2^N - 1$ and $\gcd(u, v) = 1$.

Some Properties

Almost DC balanced

2D window property

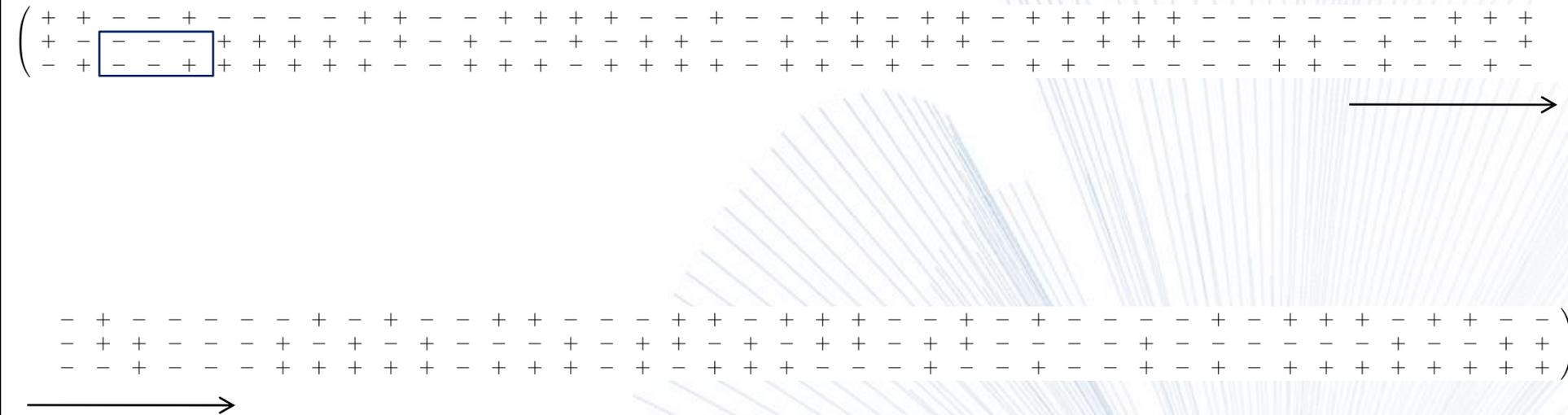
Two-valued 2D autocorrelation

Shift-multiply property

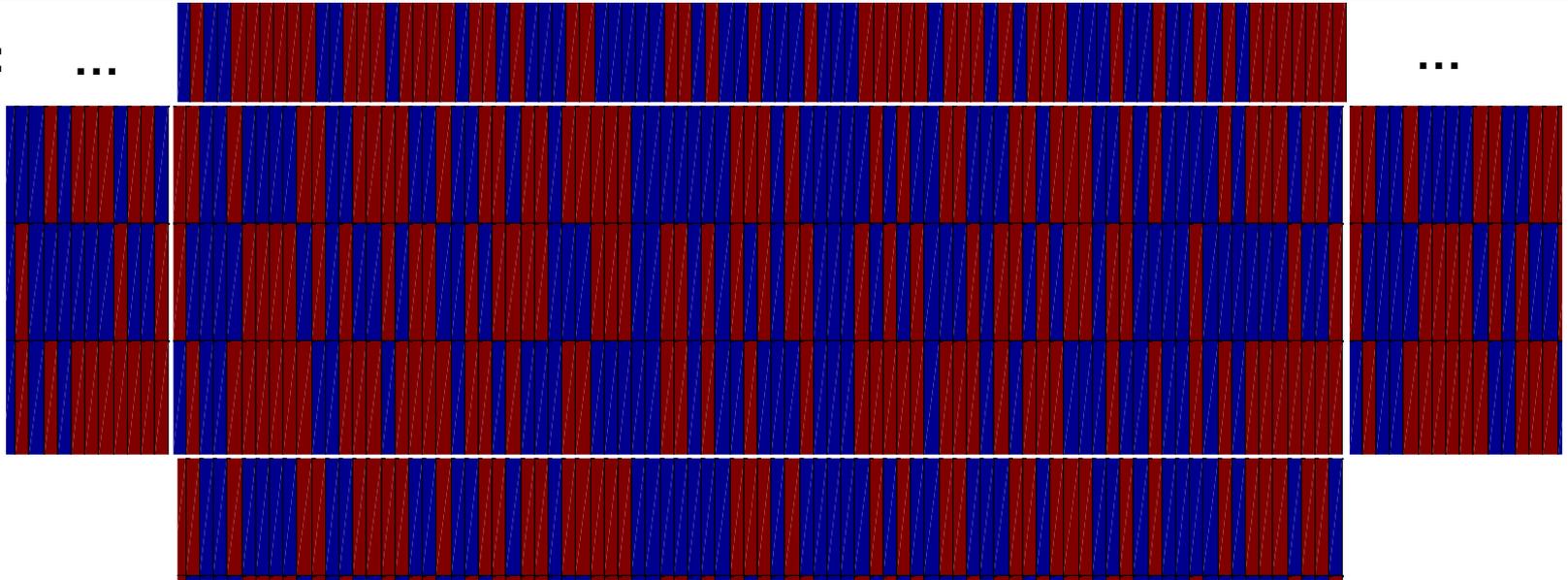
One dimensional PRBS

*Suggested reading: MacWilliams and Sloane, Proc. of the IEEE, Vol. 64, Num 12, December 1976
T. Etzion, Trans. on Information Theory, Vol. 34, No. 5, September 1988*

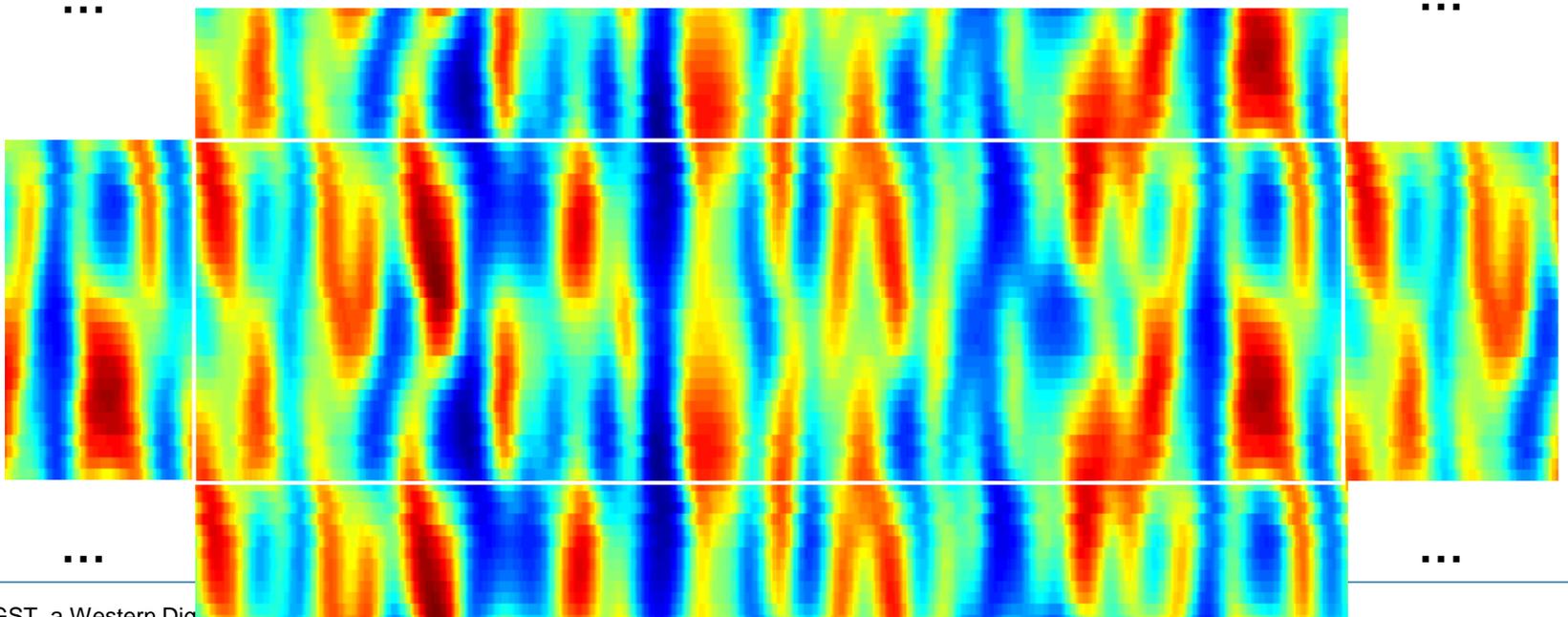
$$uv = 3 \times 85 = 255$$



Write: ...

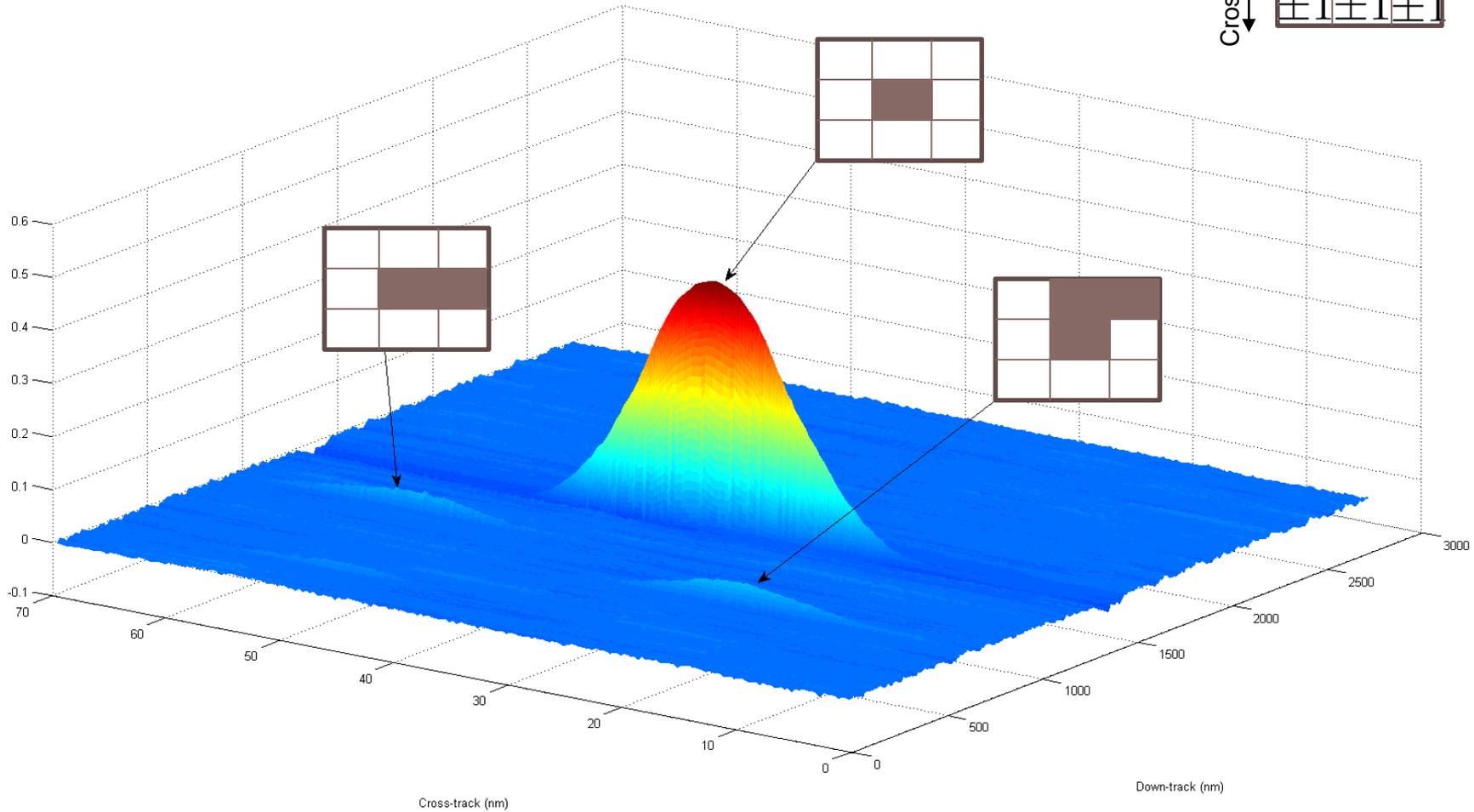
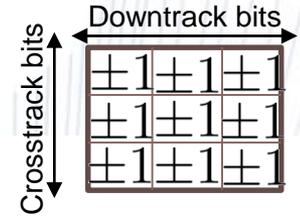


Read: ...

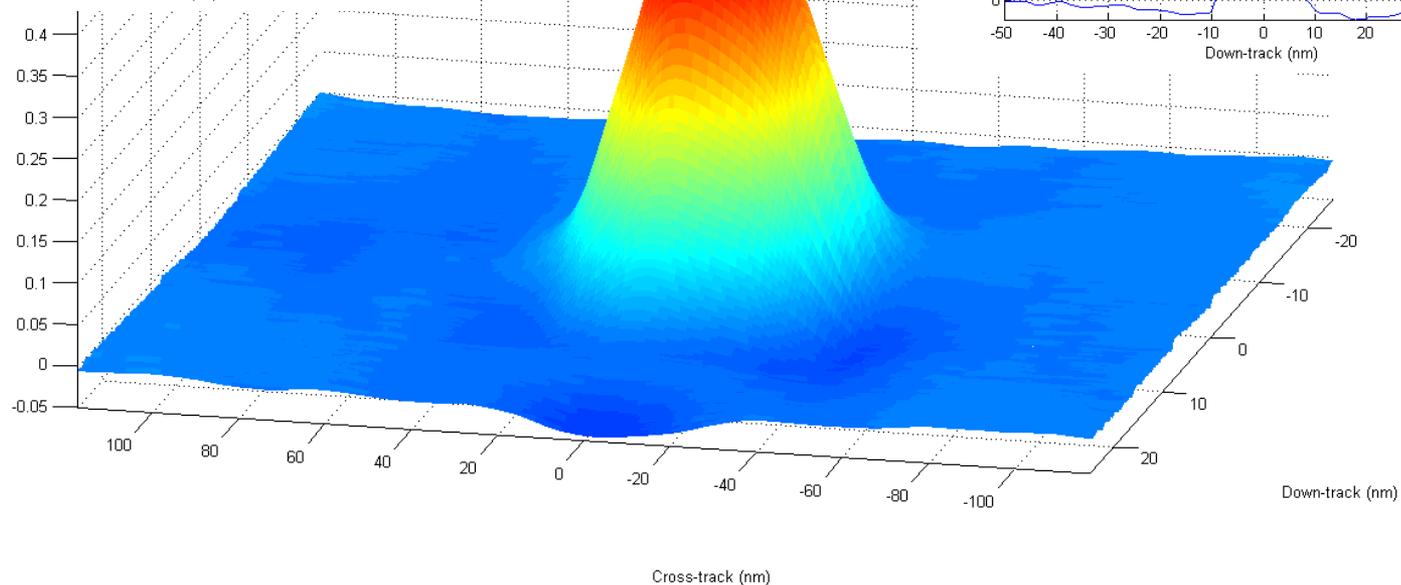
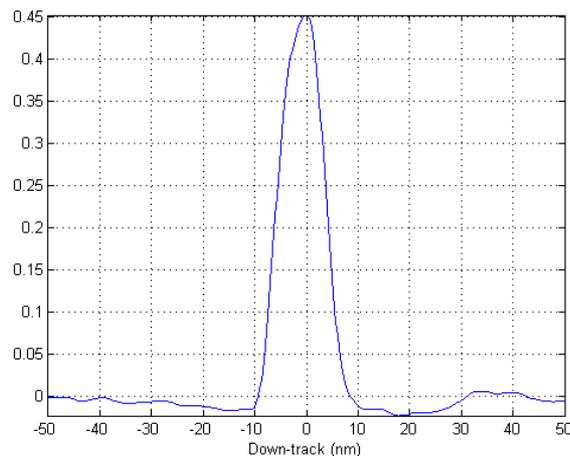
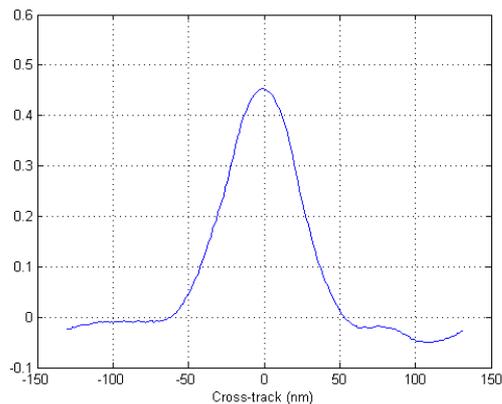


Measured 2D Deconvolution Result with Pseudorandom Array Pattern

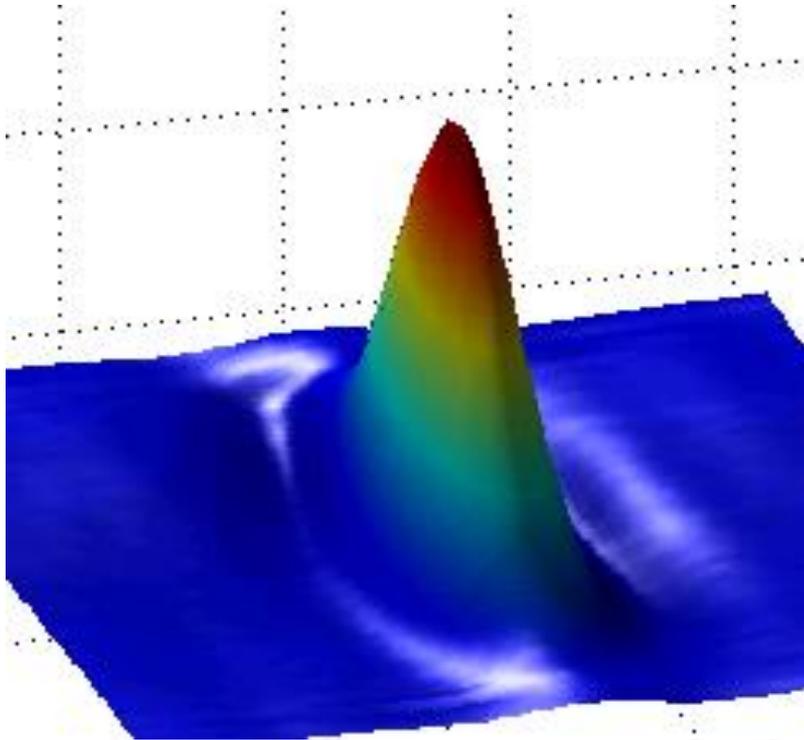
Volterra kernels: local interaction map key:



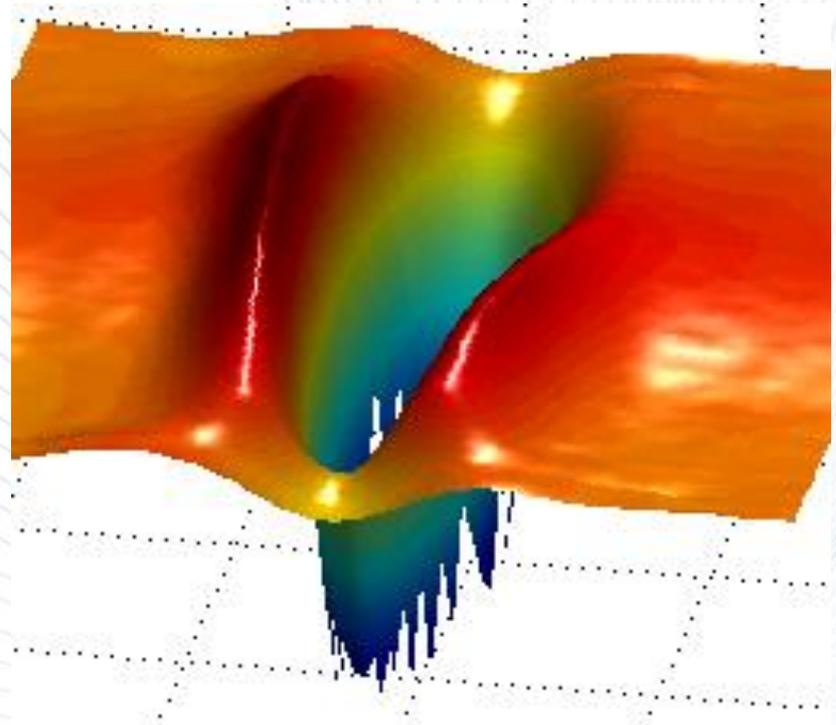
A Closer Look at the Measured "Patch Response" (the linear kernel)



Top View

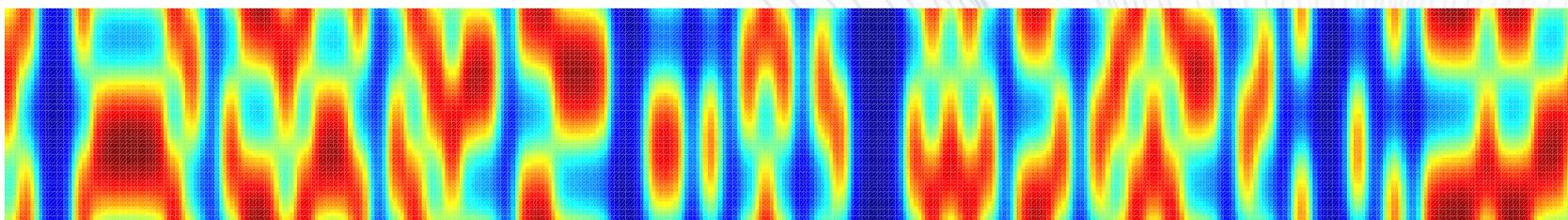
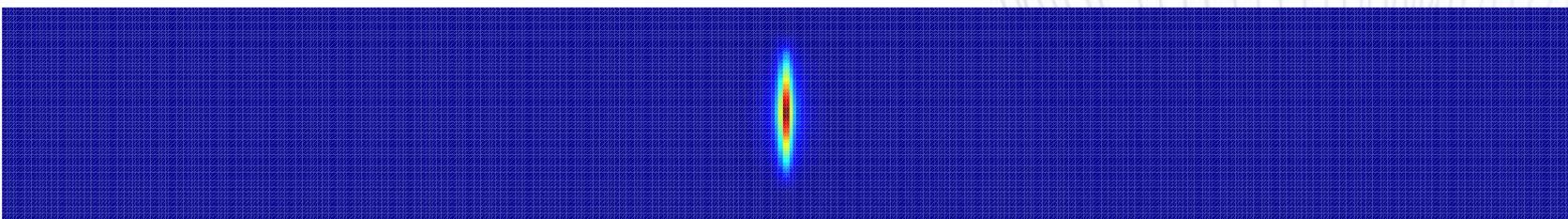
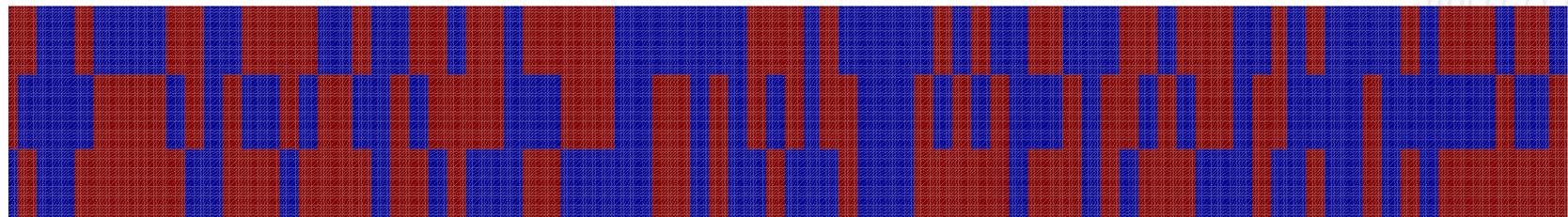


Bottom View



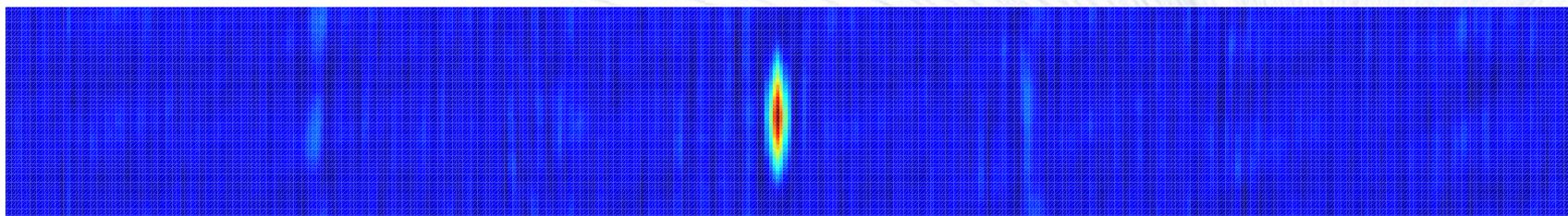
OK, what about the *Read Sensitivity Function*? (*a.k.a. 2D impulse response*)

$$M(x, y) * R(x, y) = X(x, y)$$

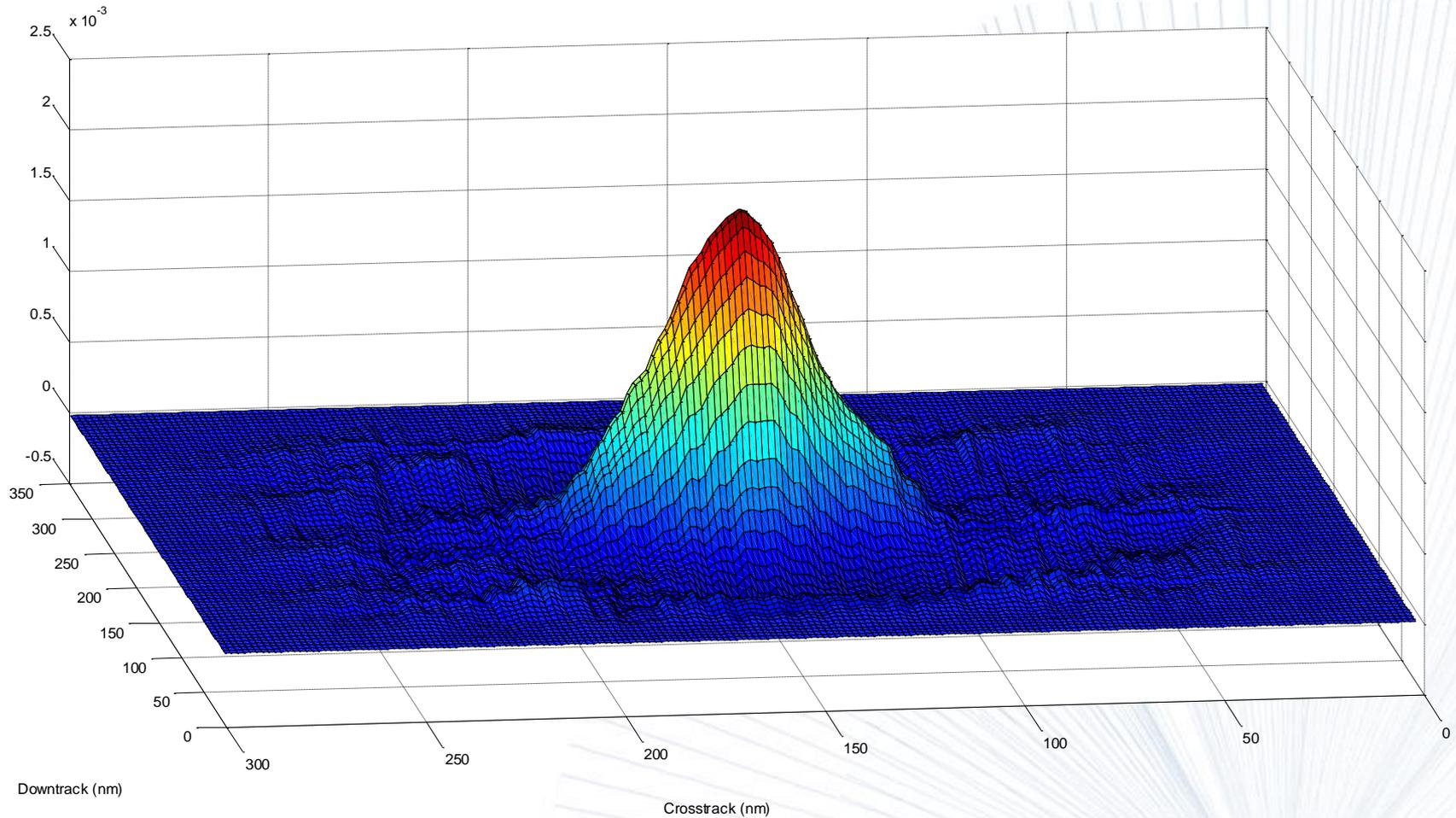


Simulated data

$$\hat{R} = X / \text{sgn}(X)$$

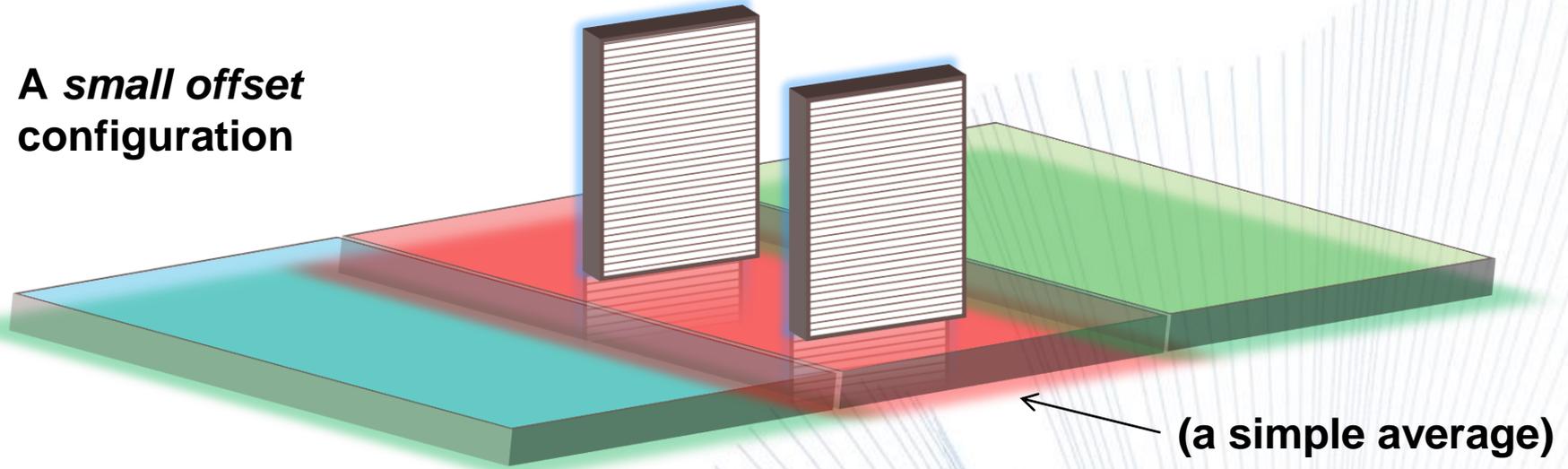


Measured 2D Read Sensitivity Function via the Projection Method

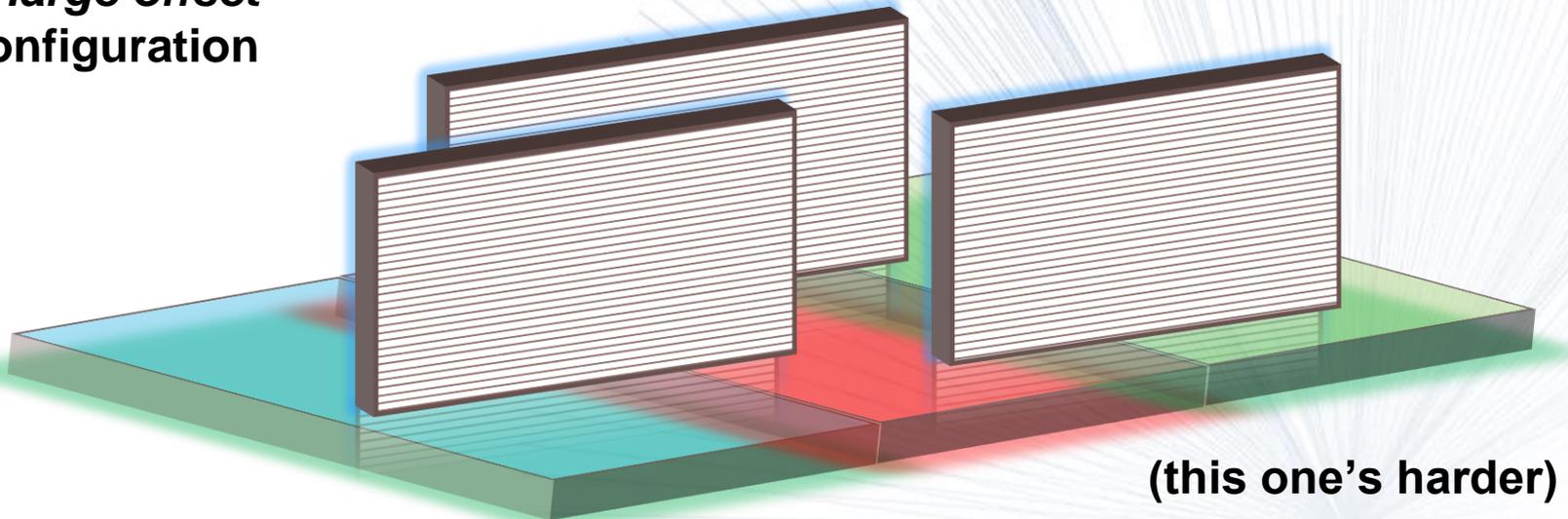


Two examples of TDMR Magnetic Systems and 2D Magnetic Signal Processing Configurations

A small offset configuration



A large offset configuration



Assumption: the two readers see identical magnetization but have independent “electronic” noises:

$$\begin{aligned} X_1 &= X_m + N_1 & N_1 &\sim \mathcal{N}(0, \sigma_1^2) \\ X_2 &= X_m + N_2 & N_2 &\sim \mathcal{N}(0, \sigma_2^2) \end{aligned}$$

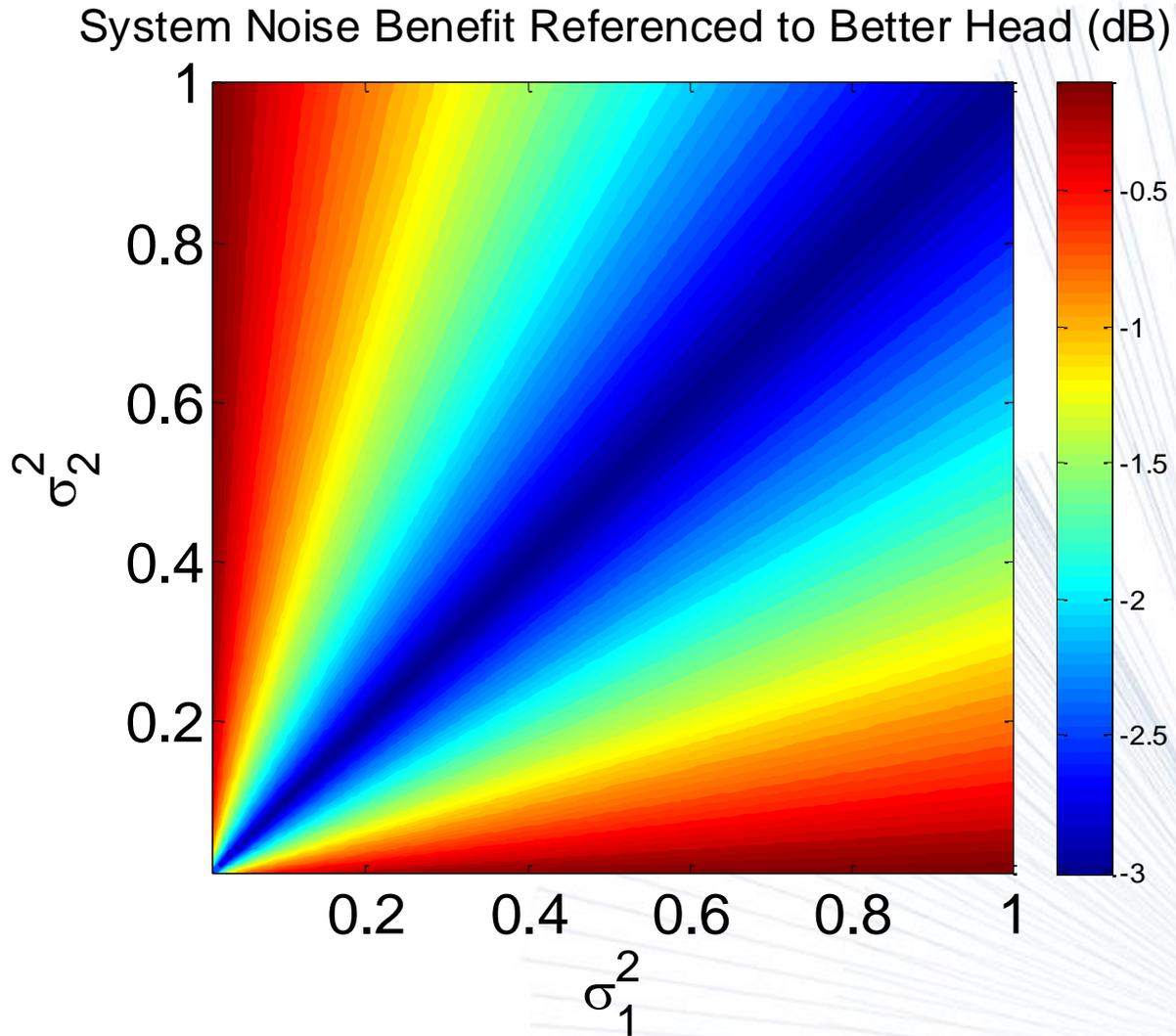
Then, an optimal technique is to use this estimate of X_m :

$$\hat{X}_m = \frac{\sigma_2^2}{\sigma_1^2 + \sigma_2^2} X_1 + \frac{\sigma_1^2}{\sigma_1^2 + \sigma_2^2} X_2$$

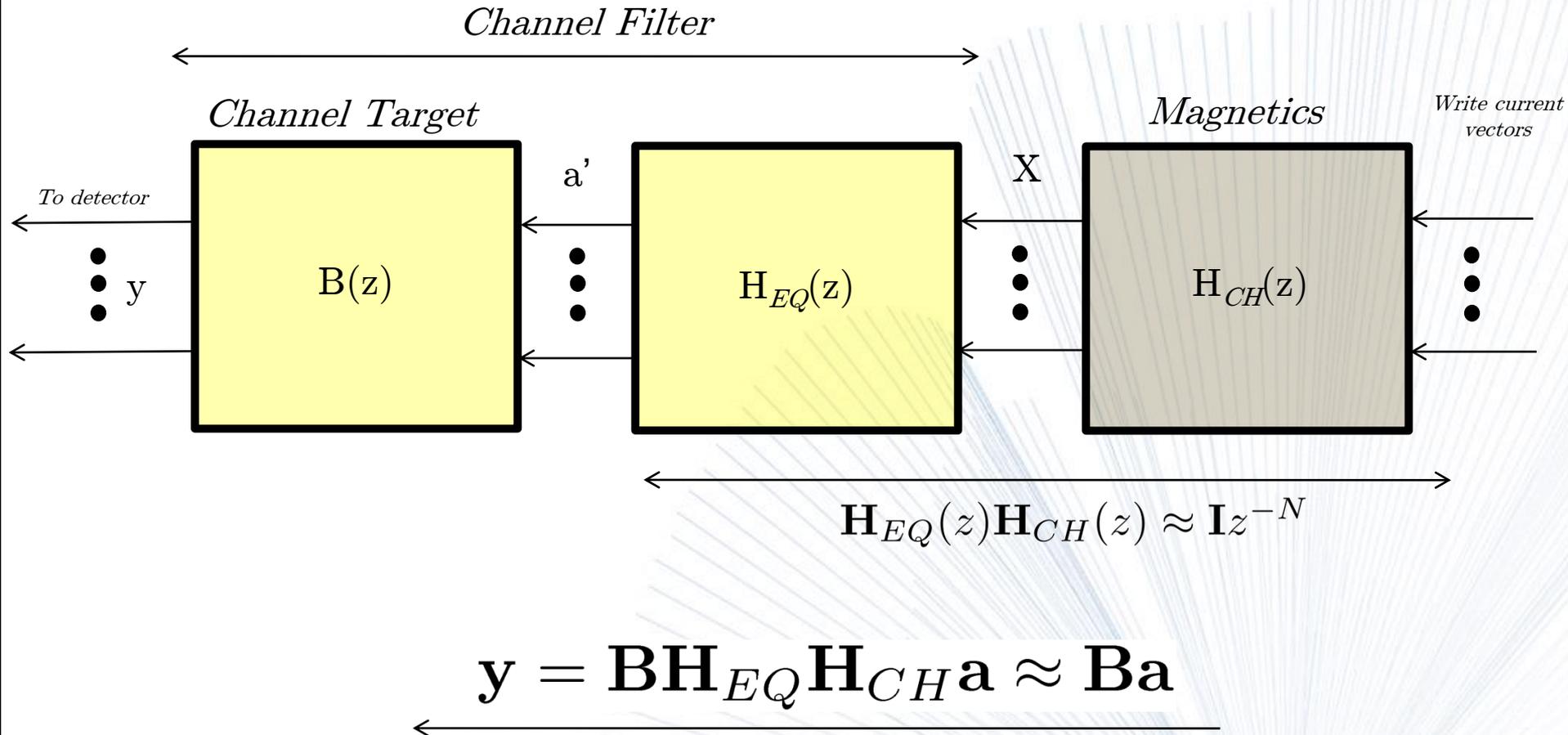
Note that this signal processing step can be written in matrix notation:

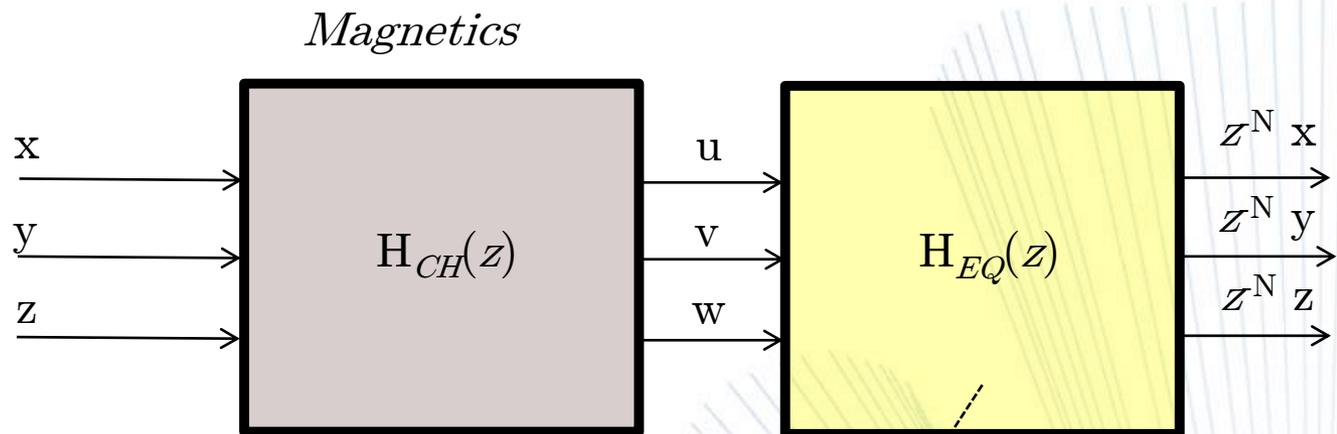
$$\hat{X}_m = \begin{pmatrix} \frac{\sigma_2^2}{\sigma_1^2 + \sigma_2^2} & \frac{\sigma_1^2}{\sigma_1^2 + \sigma_2^2} \end{pmatrix} \begin{pmatrix} X_1 \\ X_2 \end{pmatrix}$$

Suggested reading: H.L. Van Trees, “*Detection, Estimation, and Modulation Theory Part I*”, any edition.



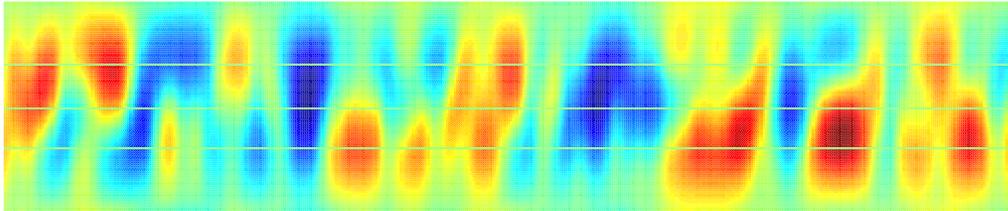
$$\sigma'^2 = \frac{\sigma_1^2 \sigma_2^2}{\sigma_1^2 + \sigma_2^2}$$



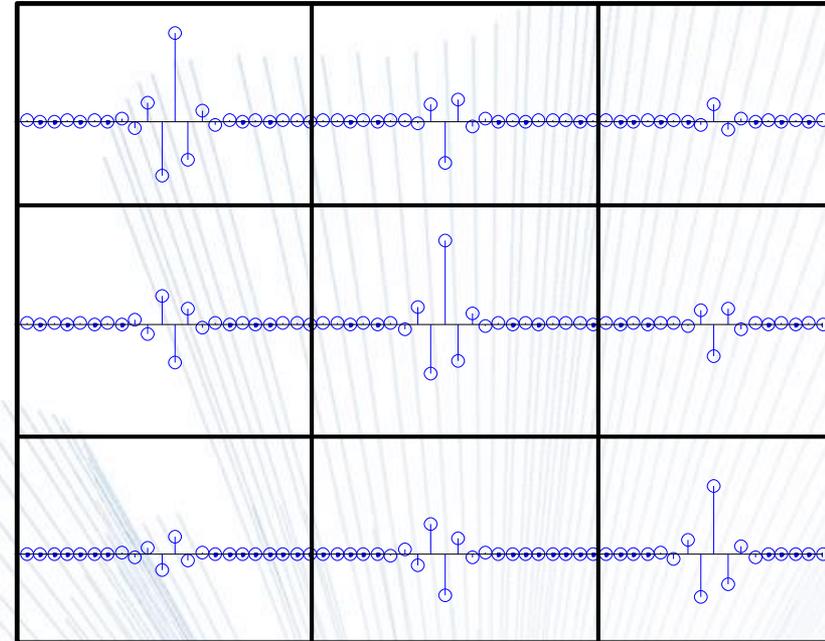


$$\begin{array}{c} \text{3 x Ntaps} \end{array} \begin{array}{c} \left(\begin{array}{ccc} \mathbf{R}_{uu} & \mathbf{R}_{uv} & \mathbf{R}_{uw} \\ \mathbf{R}_{vu} & \mathbf{R}_{vv} & \mathbf{R}_{vw} \\ \mathbf{R}_{wu} & \mathbf{R}_{wv} & \mathbf{R}_{ww} \end{array} \right) \begin{array}{c} \left(\begin{array}{ccc} \mathbf{h}_{ux} & \mathbf{h}_{vx} & \mathbf{h}_{wx} \\ \mathbf{h}_{uy} & \mathbf{h}_{vy} & \mathbf{h}_{wy} \\ \mathbf{h}_{uz} & \mathbf{h}_{vz} & \mathbf{h}_{wz} \end{array} \right) = \begin{array}{c} \left(\begin{array}{ccc} \mathbf{r}_{xu} & \mathbf{r}_{yu} & \mathbf{r}_{zu} \\ \mathbf{r}_{xv} & \mathbf{r}_{yv} & \mathbf{r}_{zv} \\ \mathbf{r}_{xw} & \mathbf{r}_{yw} & \mathbf{r}_{zw} \end{array} \right) \end{array} \begin{array}{c} \text{3 x Ntaps} \end{array}
 \end{array}$$

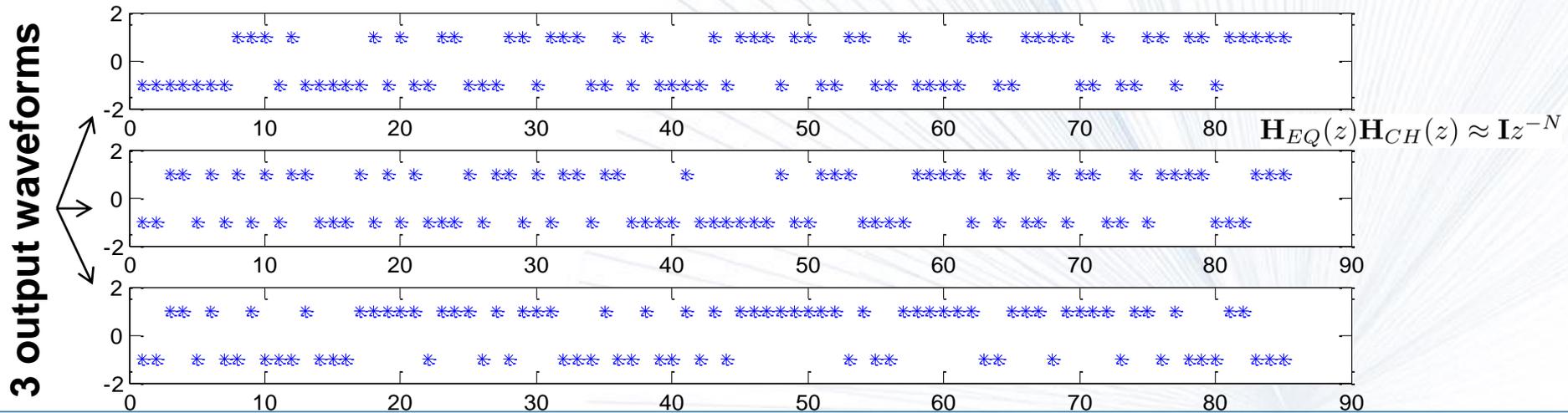
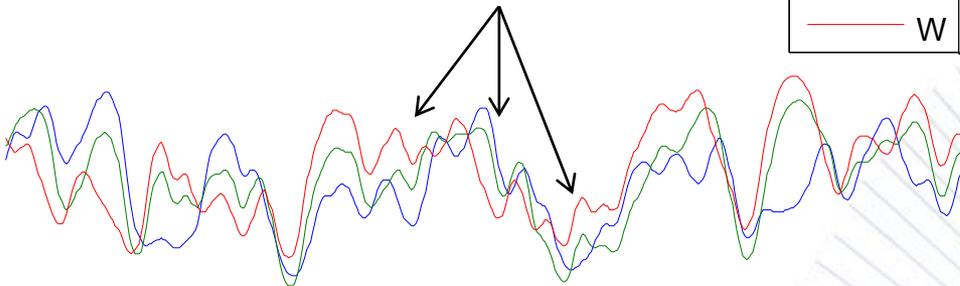
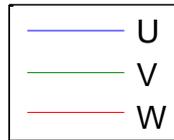
$\xleftrightarrow{3 \times \text{Ntaps}}$ $\xleftrightarrow{3}$ $\xleftrightarrow{3}$



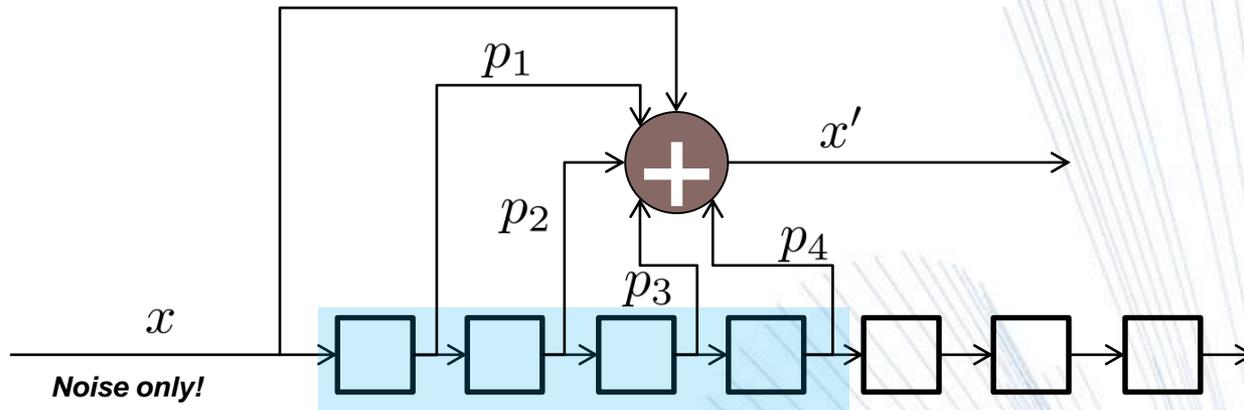
$$H_{EQ}(z)$$



3 input waveforms

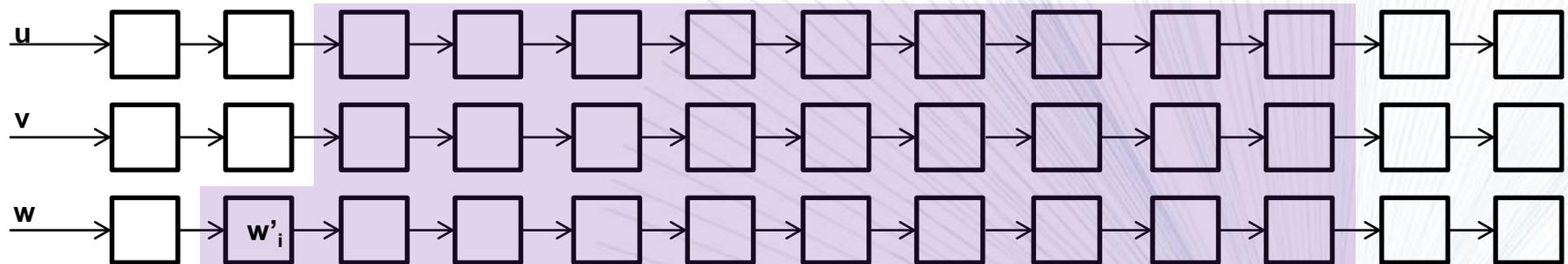
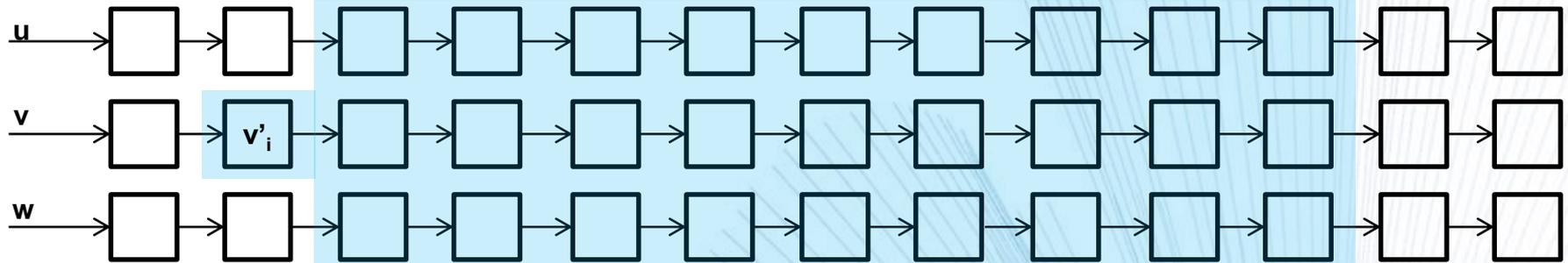
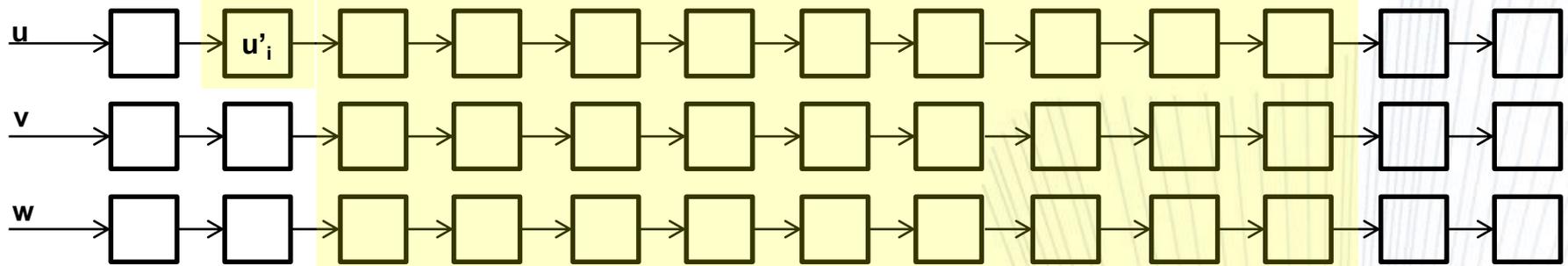


Noise Prediction and Minimization from Past Samples in One Dimension



$$B(z^{-1}) = 1 + p_1 z^{-1} + p_2 z^{-2} + p_3 z^{-3} + p_4 z^{-4} + \dots$$

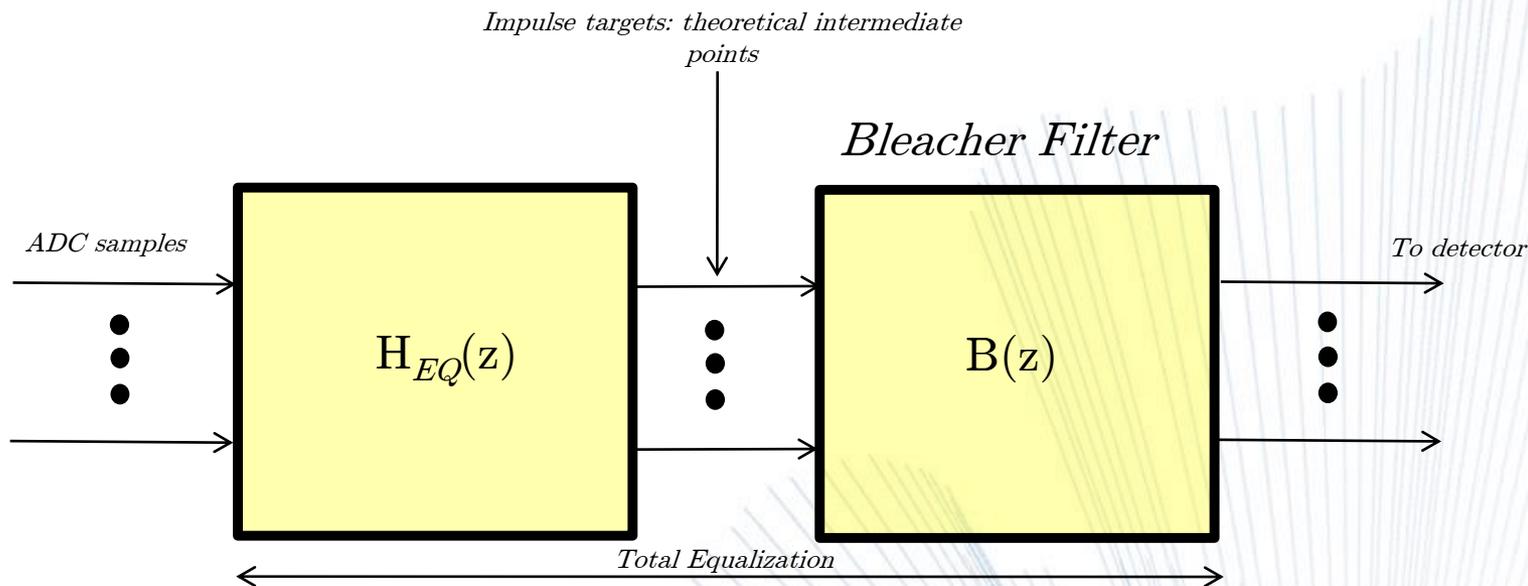
Suggested reading: Coker et al, *IEEE Trans. on Magnetics*, Vol. 34 p 110-177, 1998



2D predictor filter banks from past samples

***Thm:* u' , v' , and w' are mutually orthogonal (uncorrelated) everywhere when power-minimized, except when aligned in the same time instant.**

Design equations for 2D noise prediction from *past* samples based on noise power minimization/whitening



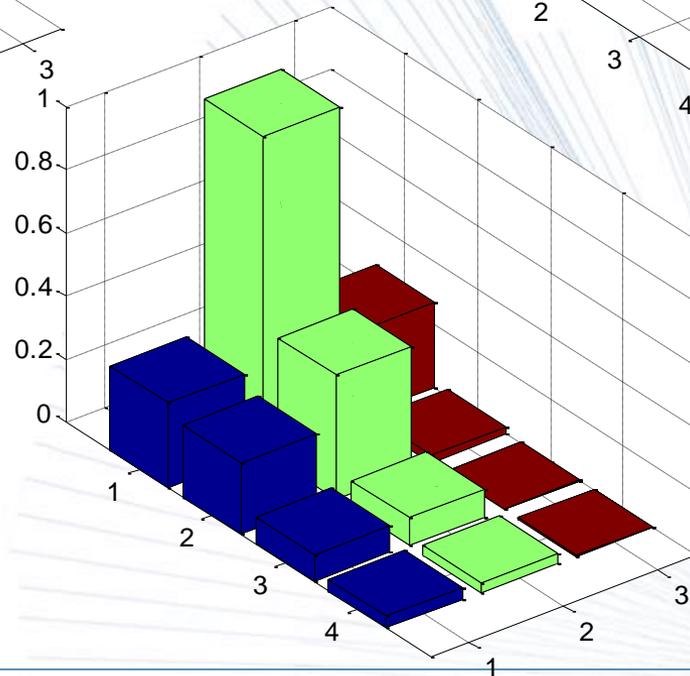
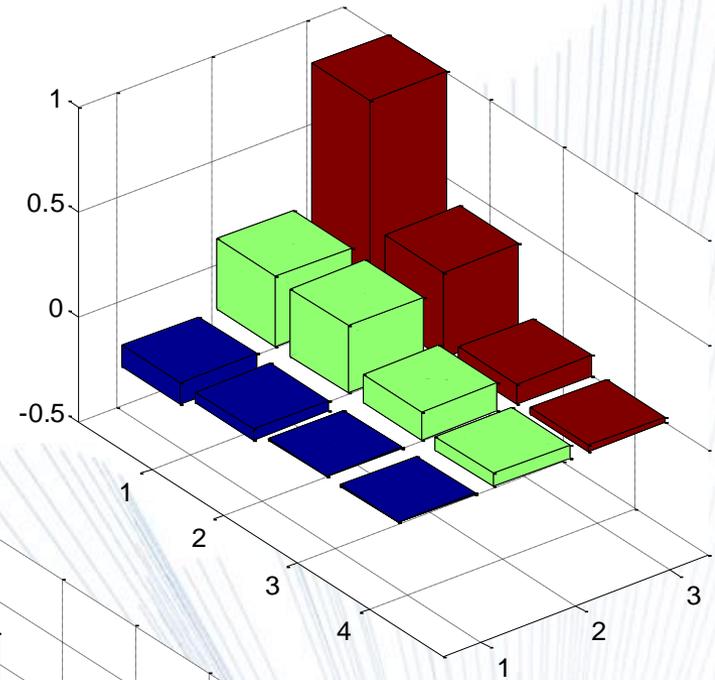
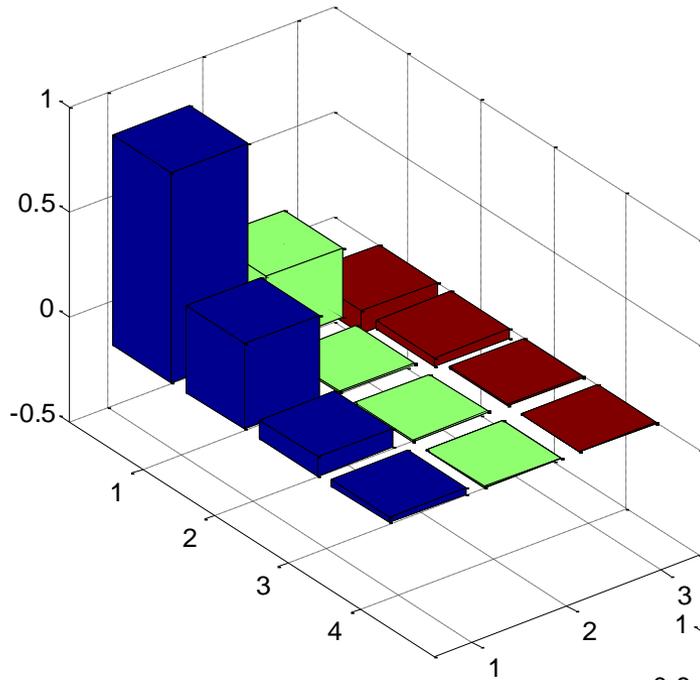
3x3 form example for minimization filter:

$$\mathbf{B}(z) = \begin{pmatrix} 1 + p_{11}z^{-1} + p_{12}z^{-2} + \dots & q_{11}z^{-1} + q_{12}z^{-2} + \dots & r_{11}z^{-1} + r_{12}z^{-2} \dots \\ p_{21}z^{-1} + p_{22}z^{-2} + \dots & 1 + q_{21}z^{-1} + q_{22}z^{-2} + \dots & r_{21}z^{-1} + r_{22}z^{-2} \dots \\ p_{31}z^{-1} + p_{32}z^{-2} + \dots & q_{31}z^{-1} + q_{32}z^{-2} + \dots & 1 + r_{31}z^{-1} + r_{32}z^{-2} \dots \end{pmatrix}$$

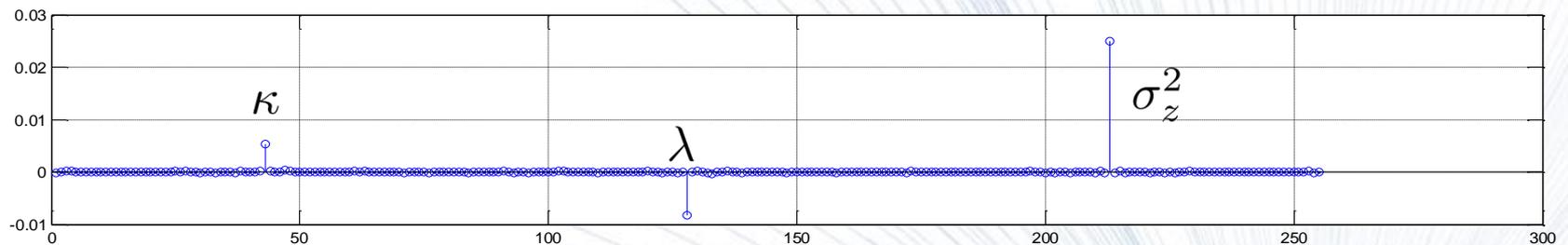
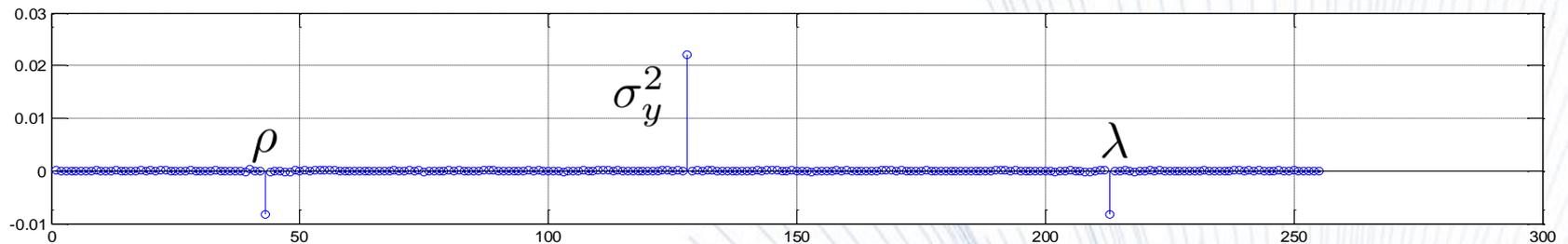
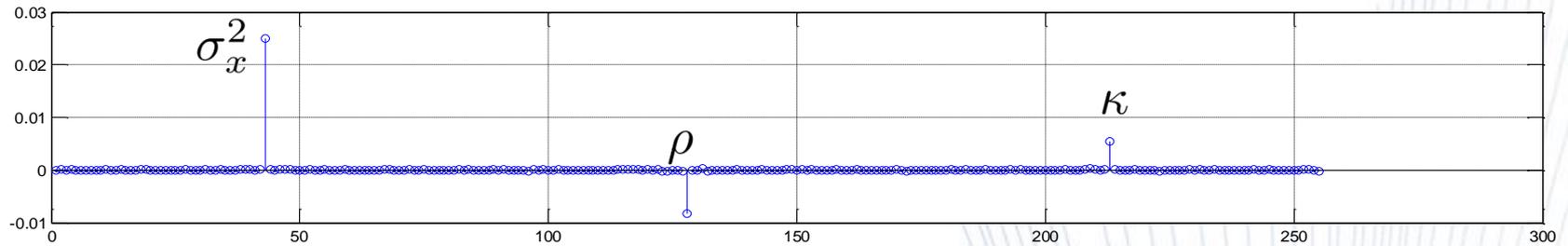
3x3 Design equations:

$$\begin{pmatrix} \mathbf{R}_{uu} & \mathbf{R}_{uv} & \mathbf{R}_{uw} \\ \mathbf{R}_{vu} & \mathbf{R}_{vv} & \mathbf{R}_{vw} \\ \mathbf{R}_{wu} & \mathbf{R}_{wv} & \mathbf{R}_{ww} \end{pmatrix} \begin{pmatrix} \mathbf{p}_1 & \mathbf{p}_2 & \mathbf{p}_3 \\ \mathbf{q}_1 & \mathbf{q}_2 & \mathbf{q}_3 \\ \mathbf{r}_1 & \mathbf{r}_2 & \mathbf{r}_3 \end{pmatrix} = - \begin{pmatrix} \mathbf{r}_{uu}(1:N) & \mathbf{r}_{vu}(1:N) & \mathbf{r}_{wu}(1:N) \\ \mathbf{r}_{vu}(1:N) & \mathbf{r}_{vv}(1:N) & \mathbf{r}_{vw}(1:N) \\ \mathbf{r}_{wu}(1:N) & \mathbf{r}_{wv}(1:N) & \mathbf{r}_{ww}(1:N) \end{pmatrix}$$

An Example Detector Target Filter B



Noise Correlation Example after Noise Prediction with Past-Sample minimization



$$K = \begin{pmatrix} \sigma_x^2 & \rho & \kappa \\ \rho & \sigma_y^2 & \lambda \\ \kappa & \lambda & \sigma_z^2 \end{pmatrix}$$

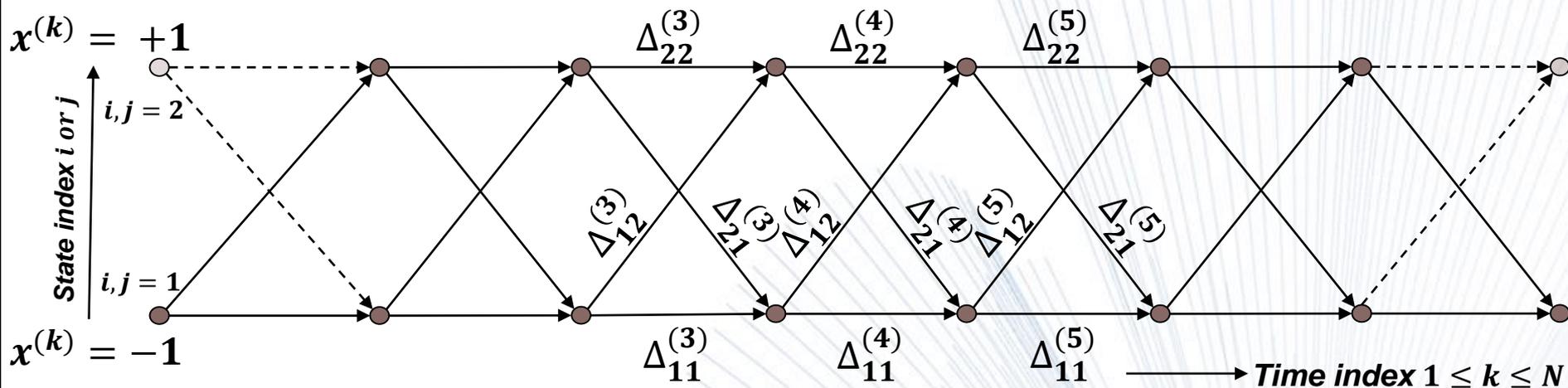
Maximum-Likelihood Detection of a Single Stream

“Noise” in the branch from state i to state j in the k^{th} instant.

Expected observation on the branch from state i to state j during the k^{th} instant. A function of a presumed input sequence x_0^N

$$\Delta_{ij}^{(k)} = Y_{ij}^{(k)} - y^{(k)}$$

Actual observation in the k^{th} instant.

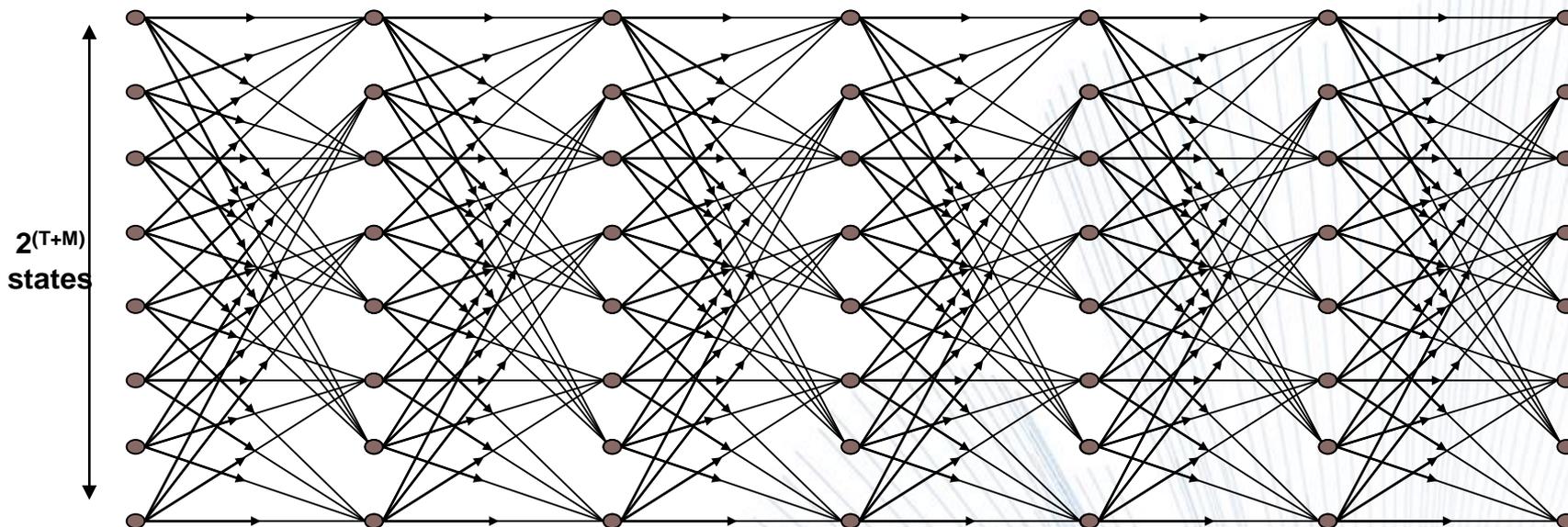


ML Sequence Detection: maximize

$$P(x_0^N | y_1^N) \propto \prod_k \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{(\Delta_{ij}^{(k)})^2}{2\sigma^2}}$$

over all possible input sequences x_0^N .

Key additional assumption: Gaussian noise



T is the total channel length = maximum span of nonzero taps relating all inputs to any output, and M is the number of channels.

Key relationship: the branch probability

$$p(\Delta_{ij}) = \frac{1}{\sqrt{(2\pi)^M |K|}} e^{-\frac{1}{2} \Delta_{ij}^T K^{-1} \Delta_{ij}}$$

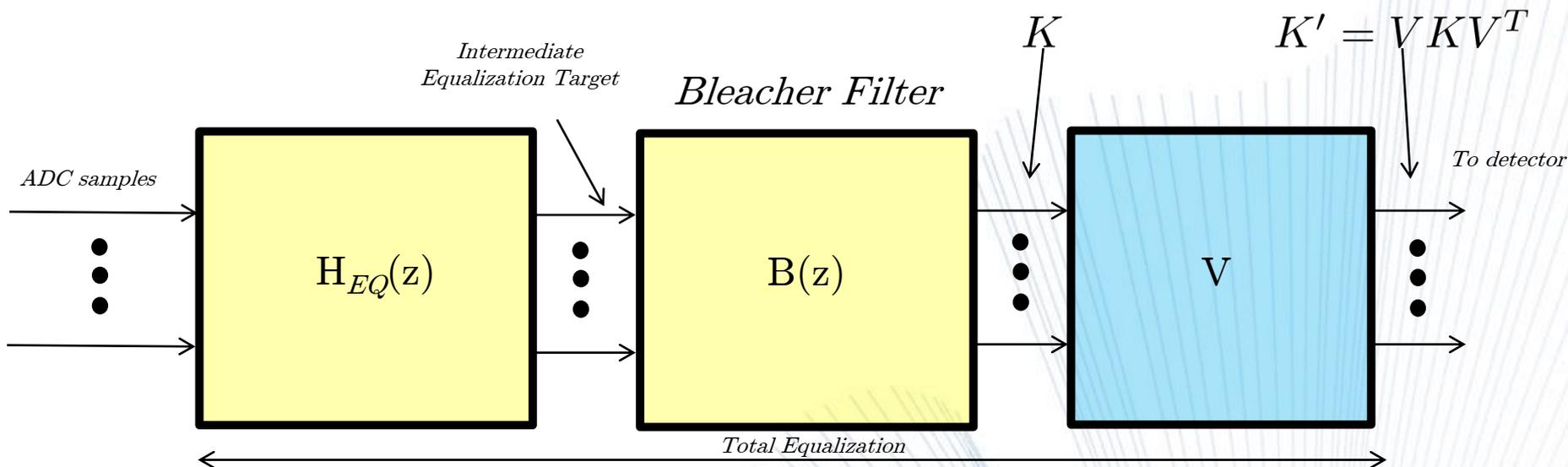
$$K \triangleq E\{\Delta\Delta^T\}$$

Example:

$$\Delta^T K^{-1} \Delta = (\Delta x \ \Delta y) \begin{pmatrix} 1 & \rho \\ \rho & 1 \end{pmatrix}^{-1} \begin{pmatrix} \Delta x \\ \Delta y \end{pmatrix}$$

$$= \frac{\Delta x^2 - 2\Delta x \Delta y \rho + \Delta y^2}{1 - \rho^2}$$

Suggested reading: MacKay, "Information Theory, Inference, and Learning Algorithms", Cambridge, 2003



$$p(\Delta_{ij}) = \frac{1}{\sqrt{(2\pi)^M |K'|}} e^{-\frac{1}{2} \Delta_{ij}^T K'^{-1} \Delta_{ij}}$$

DOES NOT depend on the details of V!

Special case: eigendecomposition of $K = V_E^T \Lambda V_E$ gives diagonal form $K'^{-1} = \Lambda^{-1}$

- **Introduction**
- **Magnetic Models in Two Dimensions**
- **Testing in Two Dimensions**
- **Magnetic Signal Processing in Two Dimensions**
- **Questions**

The logo for HGST features the letters 'HGST' in a bold, blue, sans-serif font. To the left of the 'H', there is a graphic element consisting of several thin, blue lines radiating outwards from a central point, resembling a stylized sunburst or a fan of data lines.

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