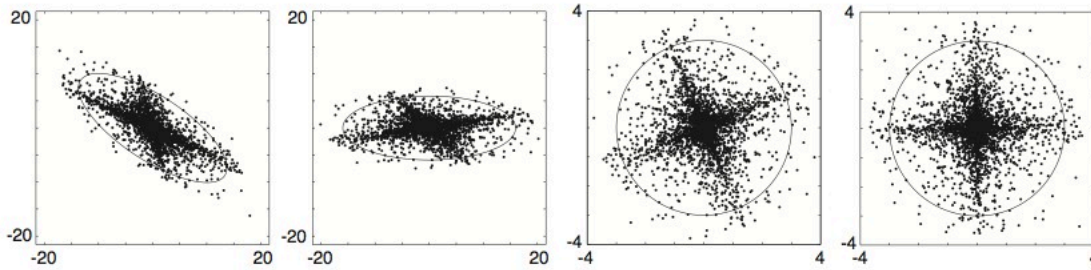


# “Independent” Components Analysis (ICA)

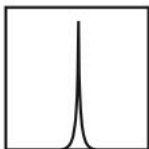
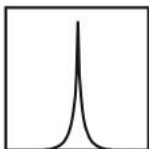
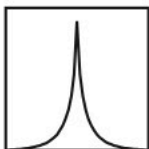


For Linearly Transformed Factorial (LTF) sources:  
guaranteed independence  
(with some minor caveats)

[Comon 94; Cardoso 96; Bell/Sejnowski 97; ...]

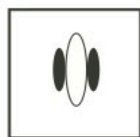
## Model II (LTF)

Coefficient  
density:



x

Basis set:



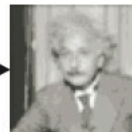
x



x



Image:



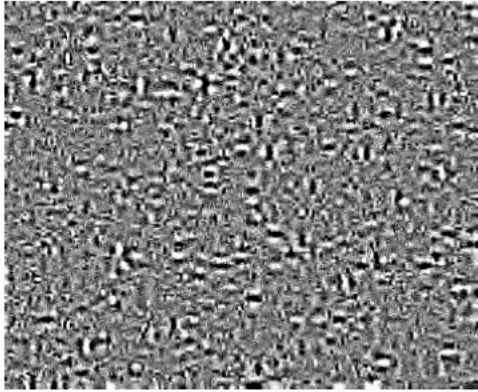
# Trouble in paradise

- Statistics: Images don't obey ICA source model
  - Image subband coefficients are clearly not independent samples (by visual inspection!)
  - The responses of ICA filters are highly dependent [Wegmann & Zetsche 90, Simoncelli 97]
  - All bandpass filters give sparse marginals [Baddeley 96] => Oriented filters are a shallow optimum [Bethge 06; Lyu & Simoncelli 08]

# Trouble in paradise

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- Biology: Visual system uses a cascade
  - Where's the retina? The LGN?
  - What happens after V1? Why don't responses get sparser? [Baddeley et al 97; Chechik et al 06]

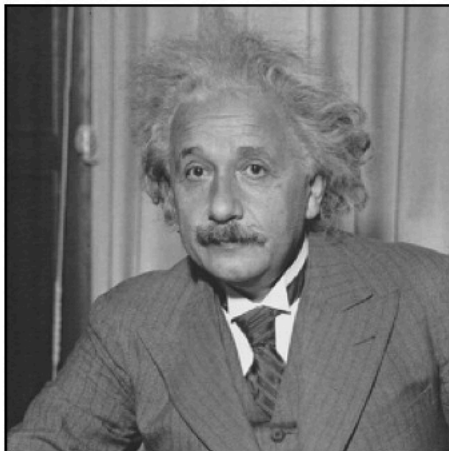
# Indications that the model is weak...



Sample from model



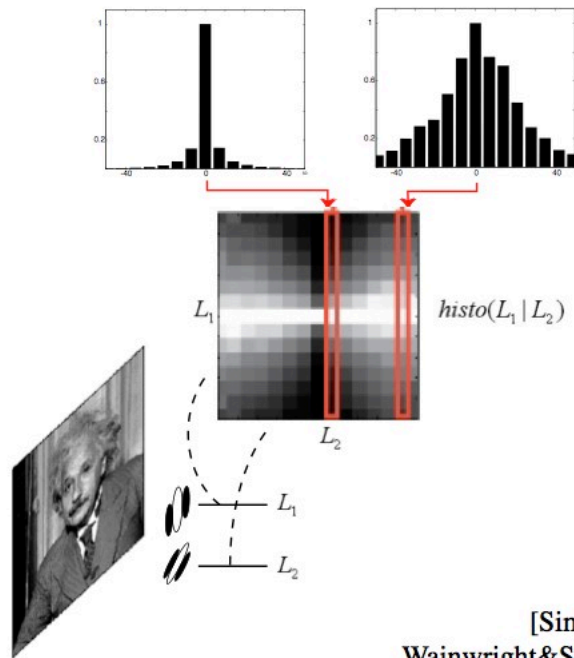
Image, ICA-transformed  
and Gaussianized



- Large-magnitude subband coefficients are found at neighboring positions, orientations, and scales.

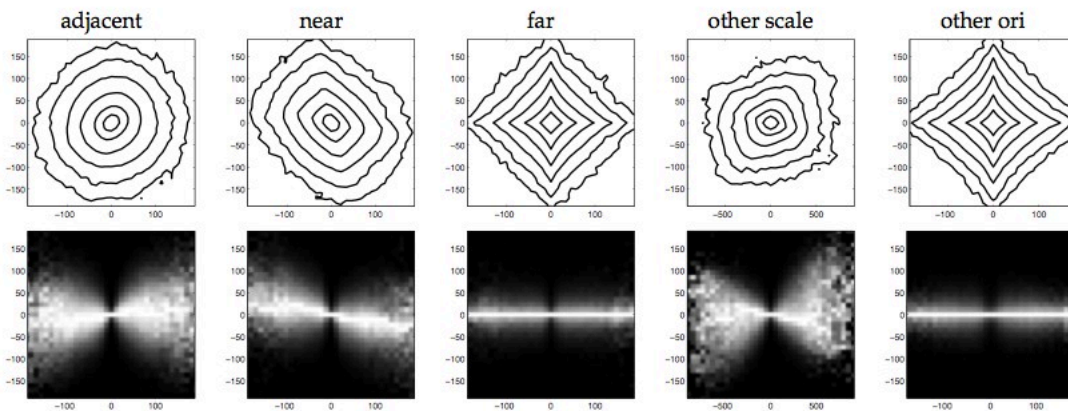


# Conditional densities reveal nonlinear dependencies



[Simoncelli 97; Buccigrossi&Simoncelli 99; Wainwright&Simoncelli 99; Schwartz&Simoncelli 01]

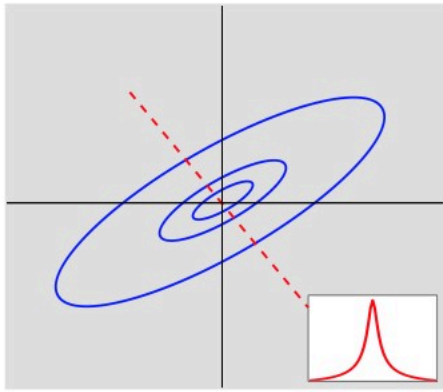
# Joint densities



- Nearby: densities are approximately circular/elliptical
- Distant: densities are approximately factorial

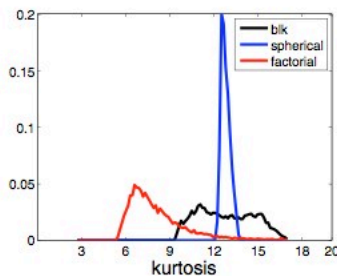
[Simoncelli, '97; Wainwright&Simoncelli, '99]

## Non-Gaussian elliptical observations and models of natural images:



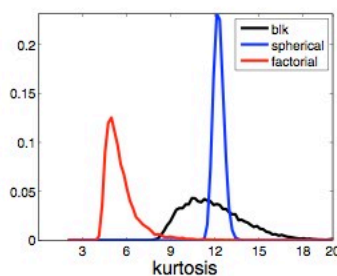
- Zetsche & Krieger, 1999;
- Huang & Mumford, 1999;
- Wainwright & Simoncelli, 1999;
- Hyvärinen and Hoyer, 2000;
- Parra et al., 2001;
- Srivastava et al., 2002;
- Sendur & Selesnick, 2002;
- Teh et al., 2003;
- Gehler and Welling, 2006
- Lyu & Simoncelli, 2008
- etc.

## Spherical vs Sparse



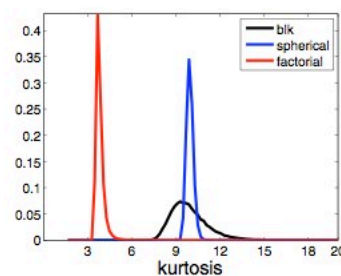
3x3

data (ICA'd): —



7x7

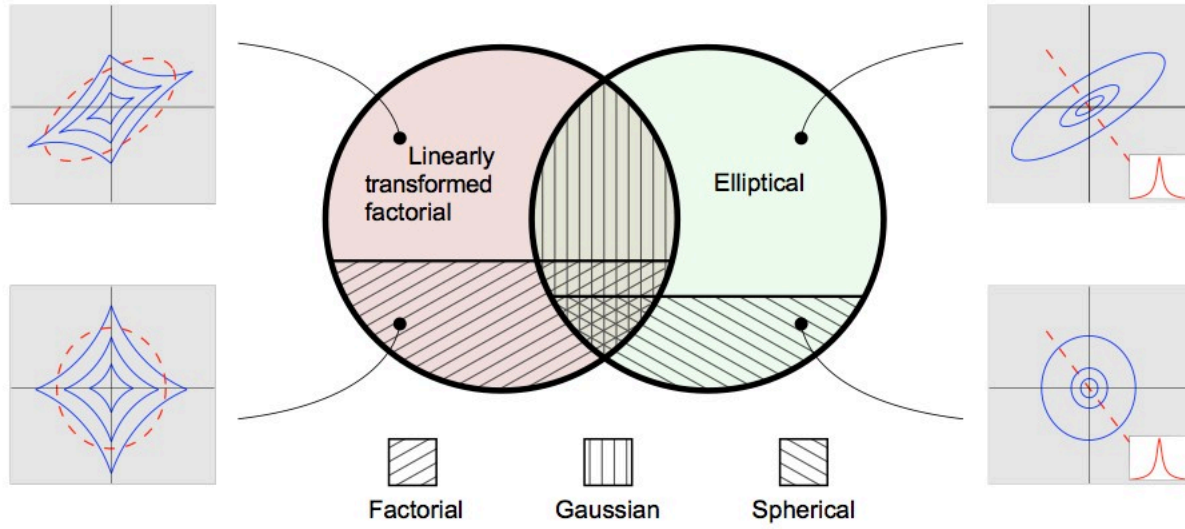
sphericalized: —



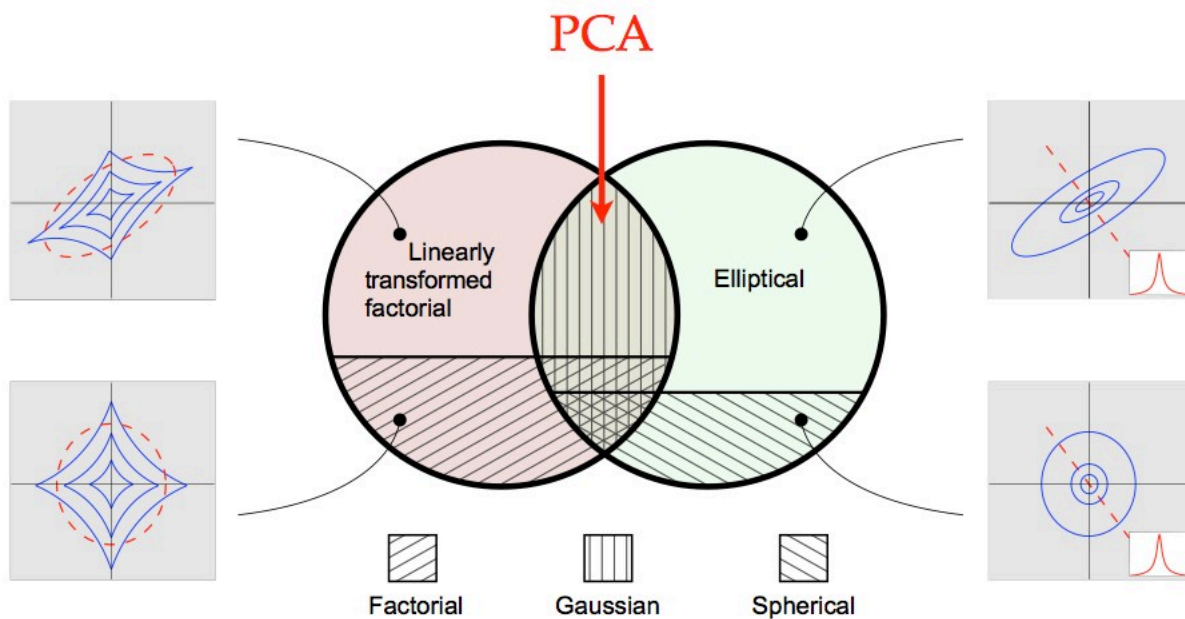
15x15

factorialized: —

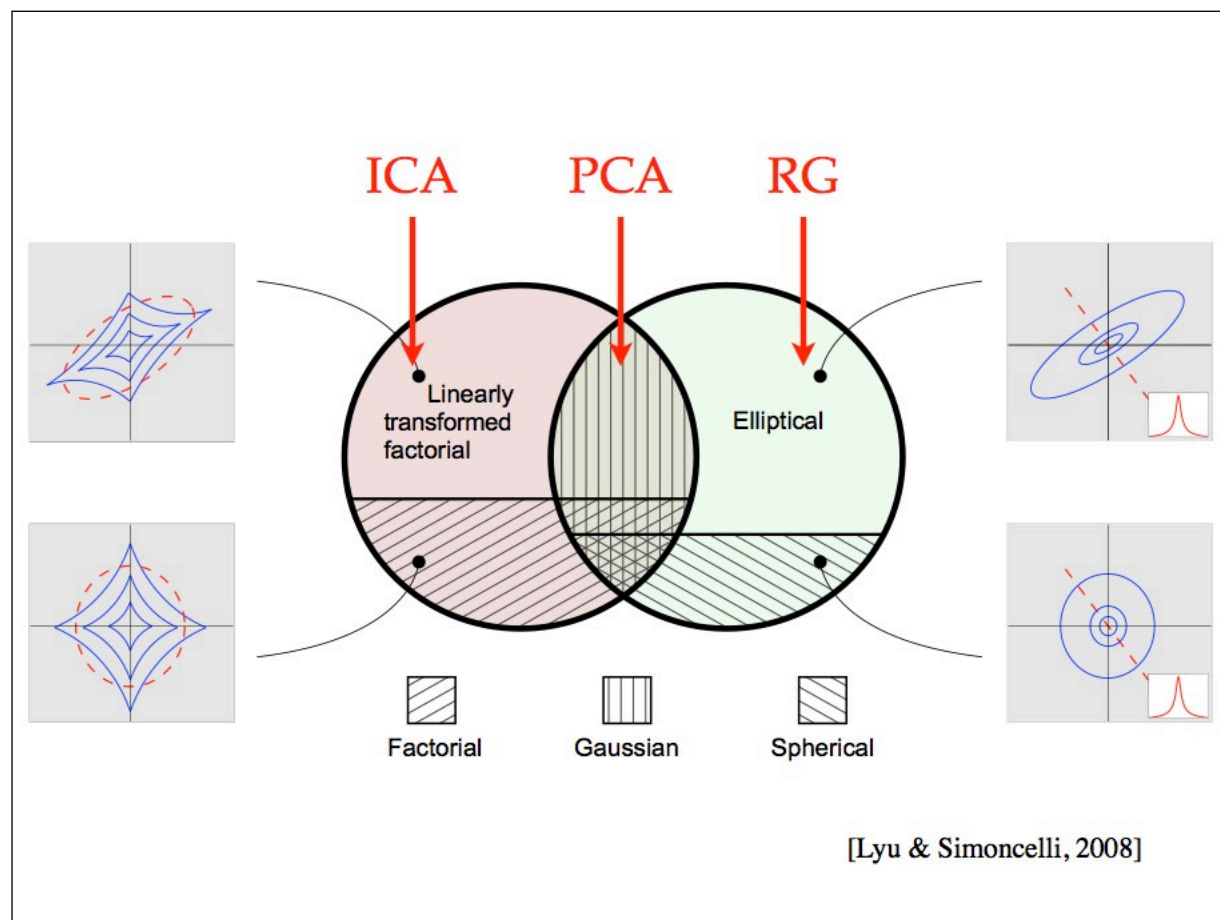
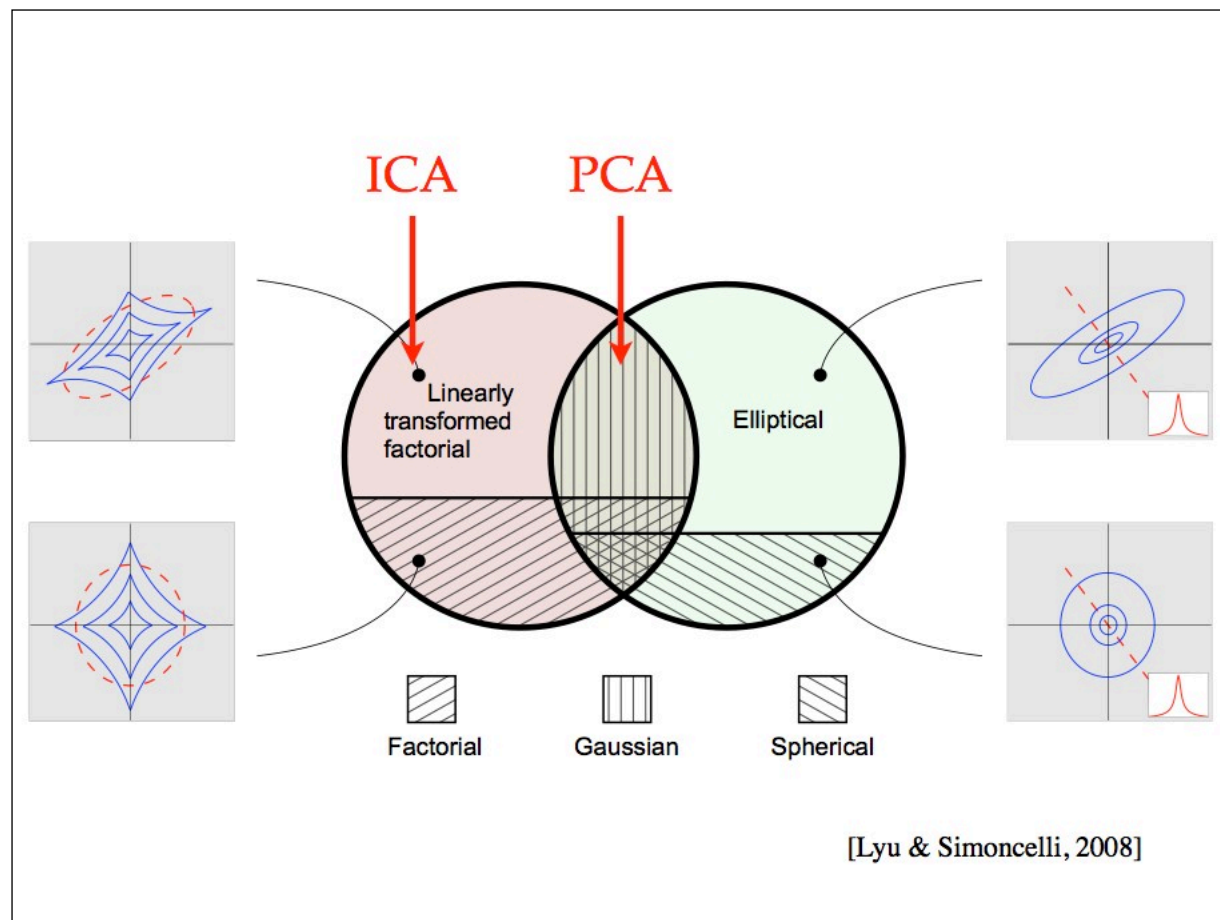
- Histograms, kurtosis of projections of image blocks onto random unit-norm basis functions.
- These imply data are closer to spherical than factorial



[Lyu & Simoncelli, 2008]



[Lyu & Simoncelli, 2008]

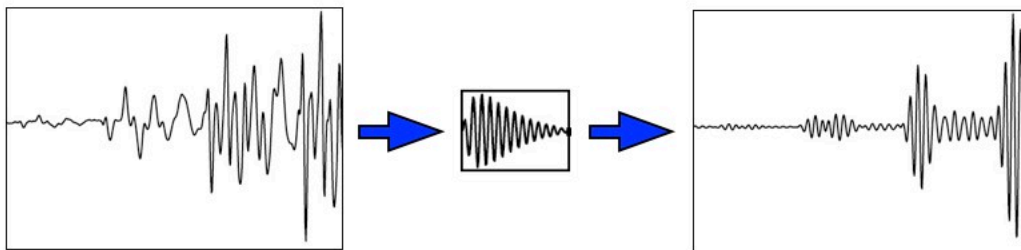




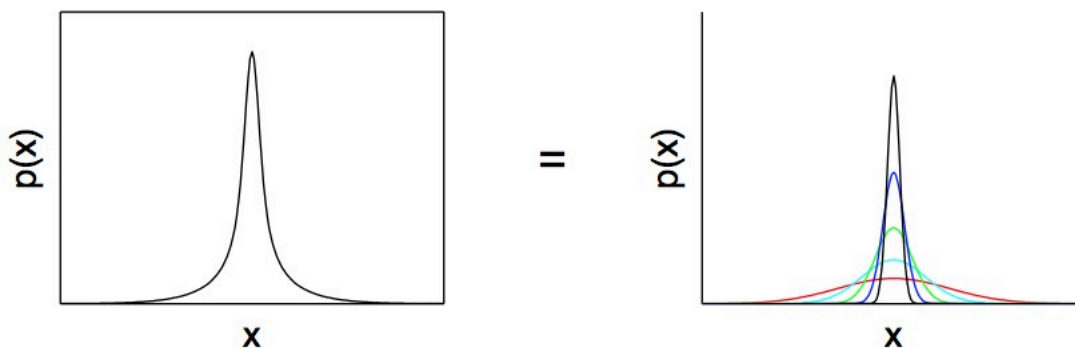
# Putting it all together...

- Subband coefficients are marginally non-Gaussian
- Coefficient pairs, or local clusters, are approximately elliptical
- Image subbands contain a small number of very large coefficients (that's what lets us separate them from noise), and these tend to occur near each other
- So suppose coefficients are locally Gaussian, but the variance is *fluctuating* over the image (known as heteroscedasticity)

## Marginal statistics - sound



Signal is *heteroskedastic* (has varying variance):





# Modeling heteroscedasticity

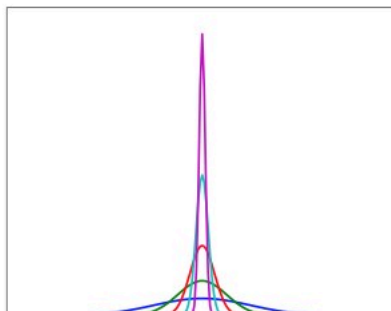
Assume a hidden scaling variable for each patch

Gaussian scale mixture (GSM)

[Andrews & Mallows 74]:

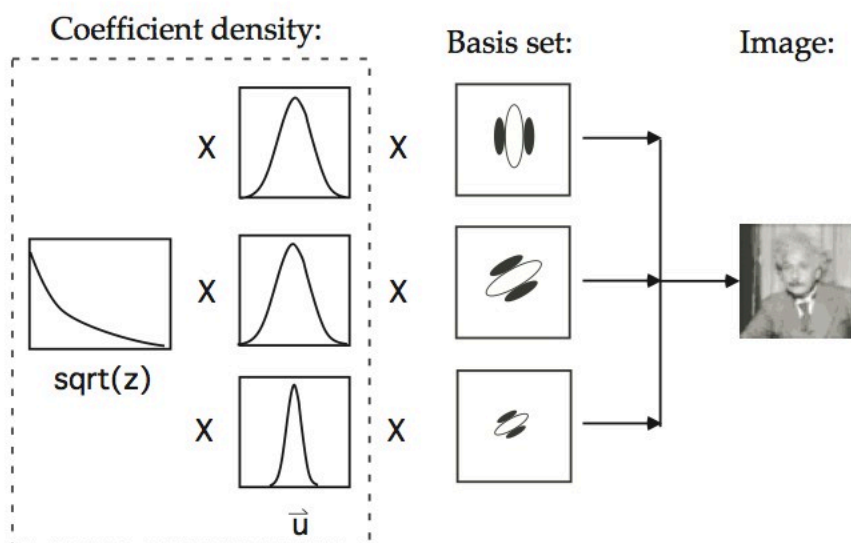
$$\vec{x} = \sqrt{z}\vec{u}$$

- $\vec{u}$  is Gaussian,  $z > 0$
- $z$  and  $\vec{u}$  are independent
- $\vec{x}$  is elliptically symmetric, with covariance  $\propto C_u$
- marginals of  $\vec{x}$  are leptokurtotic



[Wainwright&Simoncelli 99]

## Model III (GSM)



# Denoising: Joint

$$\begin{aligned}\mathbb{E}(x|\vec{y}) &= \int dz \mathcal{P}(z|\vec{y}) \mathbb{E}(x|\vec{y}, z) \\ &= \int dz \mathcal{P}(z|\vec{y}) \left[ zC_u(zC_u + C_w)^{-1}\vec{y} \right]_{\text{ctr}}\end{aligned}$$

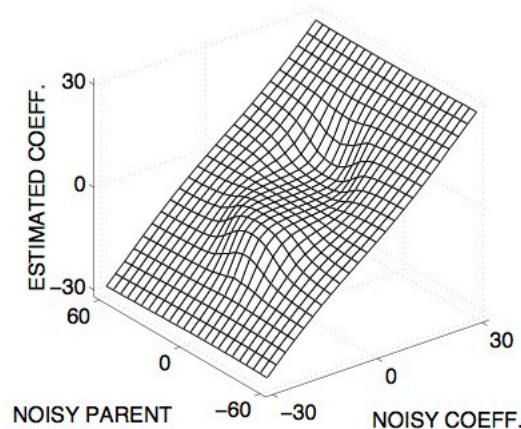
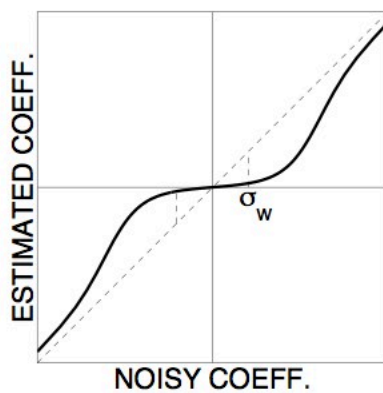
where

$$\mathcal{P}(z|\vec{y}) = \frac{\mathcal{P}(\vec{y}|z) \mathcal{P}(z)}{\mathcal{P}\vec{y}}, \quad \mathcal{P}(\vec{y}|z) = \frac{\exp(-\vec{y}^T(zC_u + C_w)^{-1}\vec{y}/2)}{\sqrt{(2\pi)^N |zC_u + C_w|}}$$

Numerical computation of solution is reasonably efficient if one jointly diagonalizes  $C_u$  and  $C_w$  ...

[Portilla et. al. 2003]

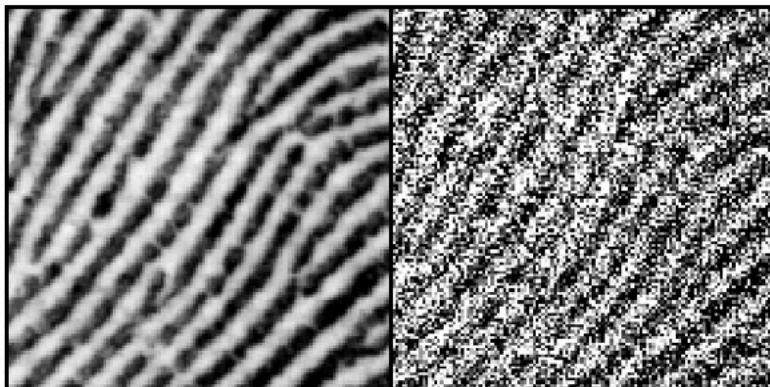
## Example estimators



Estimators for the scalar and single-neighbor cases

[Portilla et. al. 2003]

Original



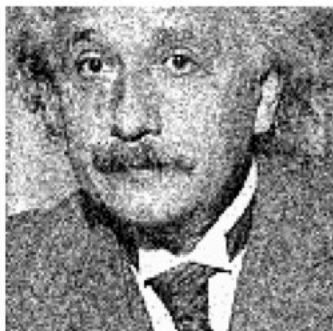
Noisy  
(8.1 dB)

UndWvlt  
Thresh  
(19.0 dB)

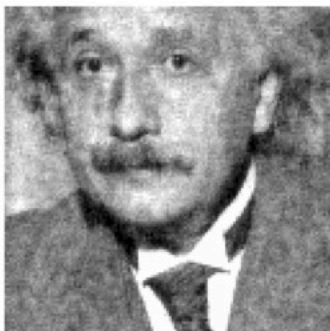
BLS-GSM  
(21.2 dB)

[Portilla et. al. 2003]

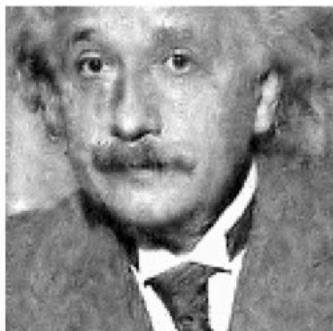
noisy  
(4.8)



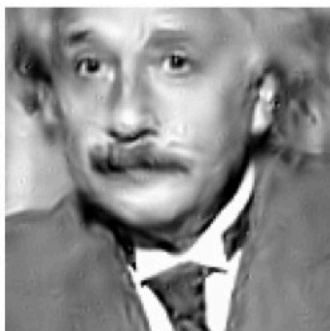
I-linear  
(10.61)



II-marginal  
(11.98)



III-GSM  
nbd:  $5 \times 5 + p$   
(13.60)



[Portilla et. al. 2003]



# Real sensor noise



400 ISO



GSM denoised

[Portilla et. al. 2003]