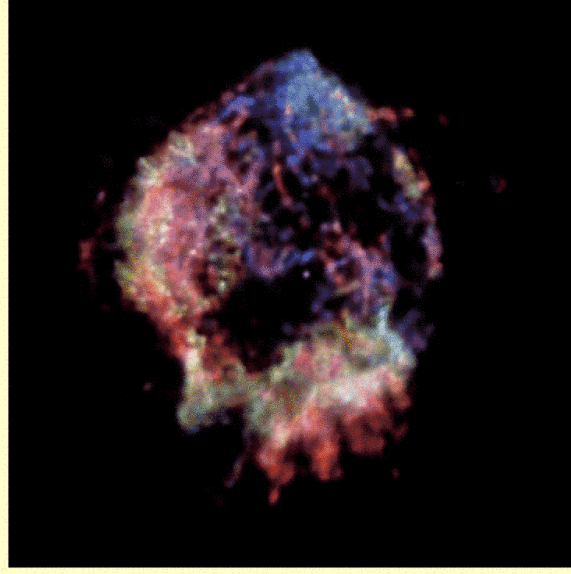


A Principal Components Analysis Applied to the Cas A Megasecond

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Jack Hughes
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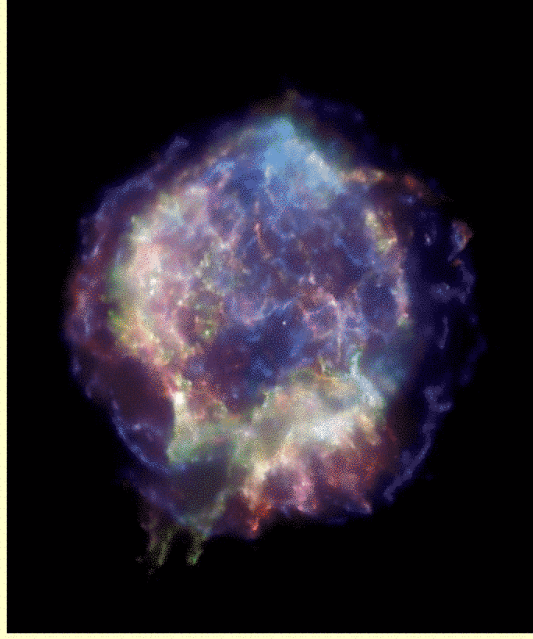
Cas A – 5,000 sec

- Hughes et al. 2000
 - Evidence for mixing, overturning of ejecta
 - Fe exterior to Si



Cas A – 50,000 sec

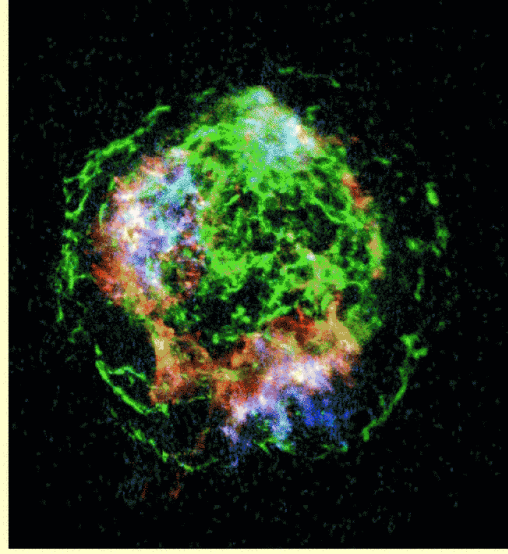
- Hwang et al. 2000
 - Map X-ray emission
- Gotthelf et al. 2001
 - Forward & reverse shocks
- Laming & Hwang 2003
 - Ejecta profile
- Hwang & Laming 2003
 - Pure Fe knots
- DeLaney et al. 2004
 - Kinematics
- And others...



Note: green areas are Si-rich

Cas A – 1,000,000 sec

- Hwang et al. 2004
 - Jet
- “... extract spectra with sufficient signal to noise to fit the spectra of the ACIS CCD...”
 - Laming & Hwang 2005



Note: red areas are Si-rich

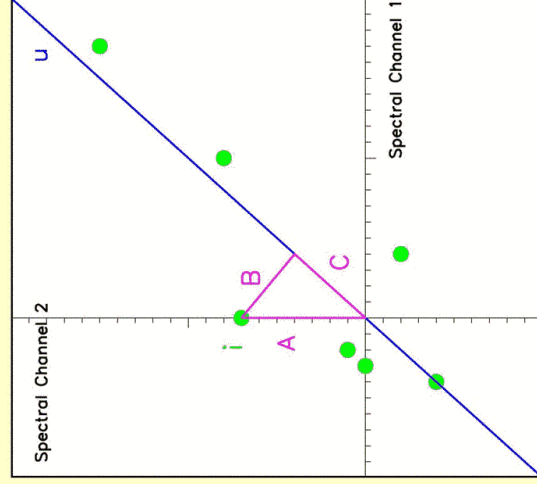
yikes!

Some help, please...

One way to get a handle on this rich dataset is to use a **Principal Components Analysis (PCA)**:

- A statistical technique often used to **reduce the dimensionality** of a dataset
- Relatively unbiased technique for identifying spectral variations in a statistically quantifiable way

Principal Components Analysis

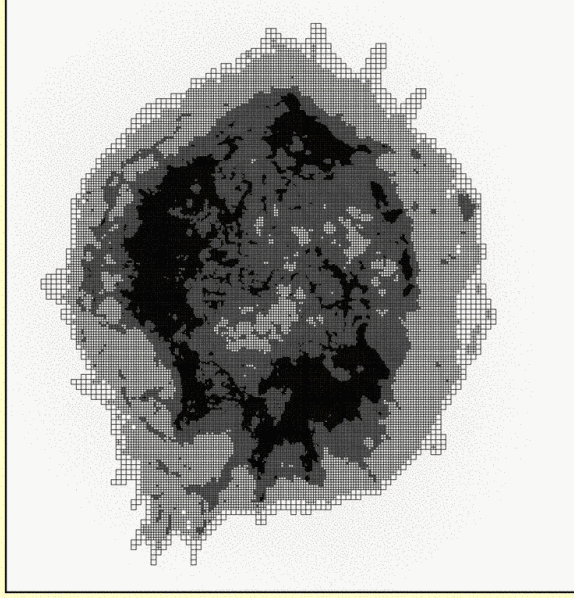


Cartoon illustration of PCA

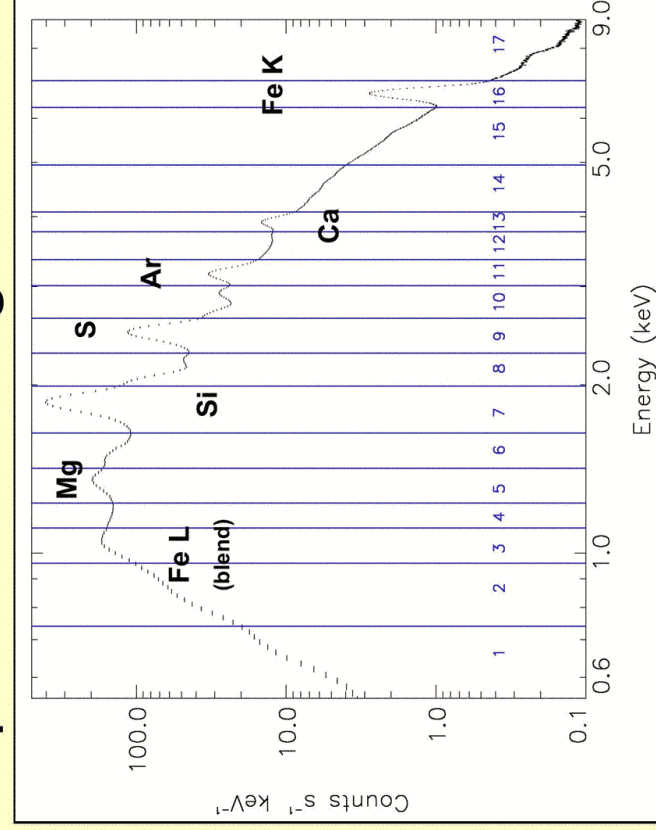
- PCA finds new variables (**U**) that maximize the variance (**C**) of the data (●)
 - The new variables (axes) are orthogonal to each other
- **U** is a linear combination of the old variables (**Spectral channels 1 and 2**)
 - **U** is an *eigenvector* of the data matrix

PCA applied to SNRs

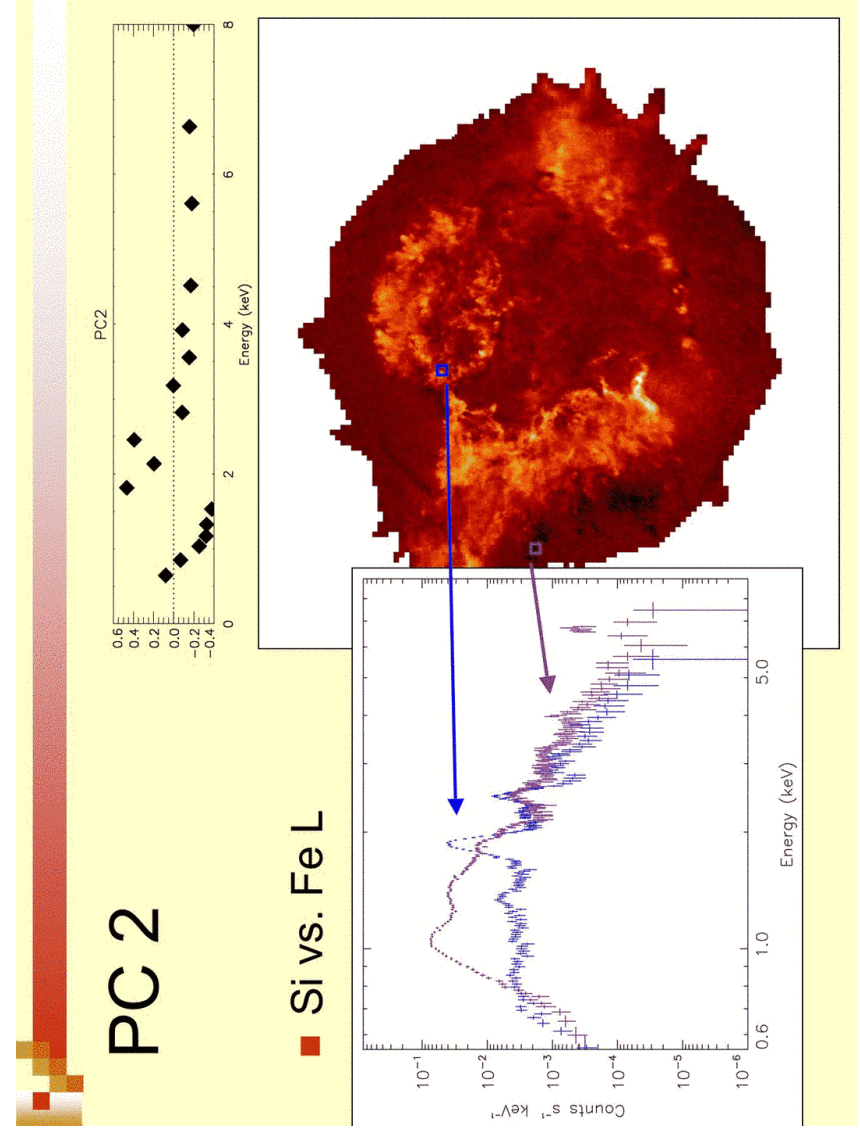
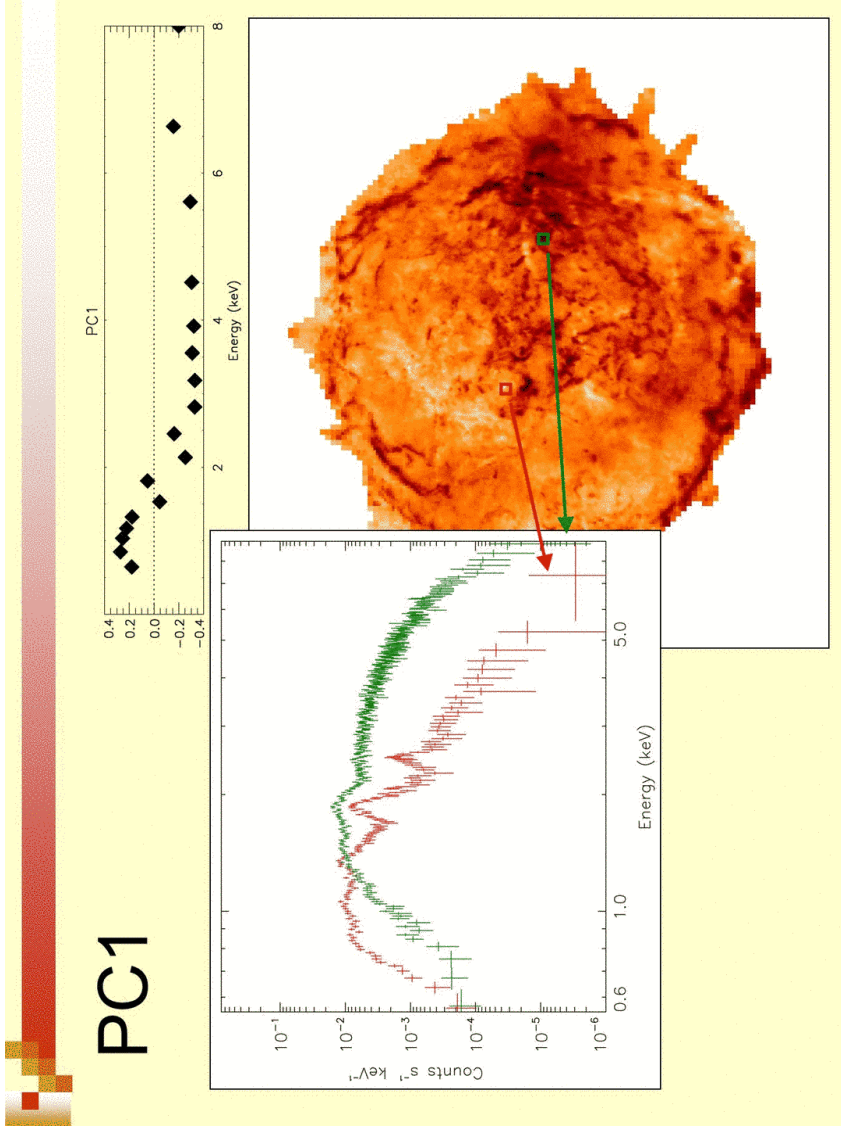
- Our data: **spectra** from many spatial **regions** in the SNR
- Divide remnant into square regions
 - Each with at least 1000 counts
 - $\sim 0.5'' \times 0.5''$ to $4'' \times 4''$
 - 125,181 regions



Spectral Binning

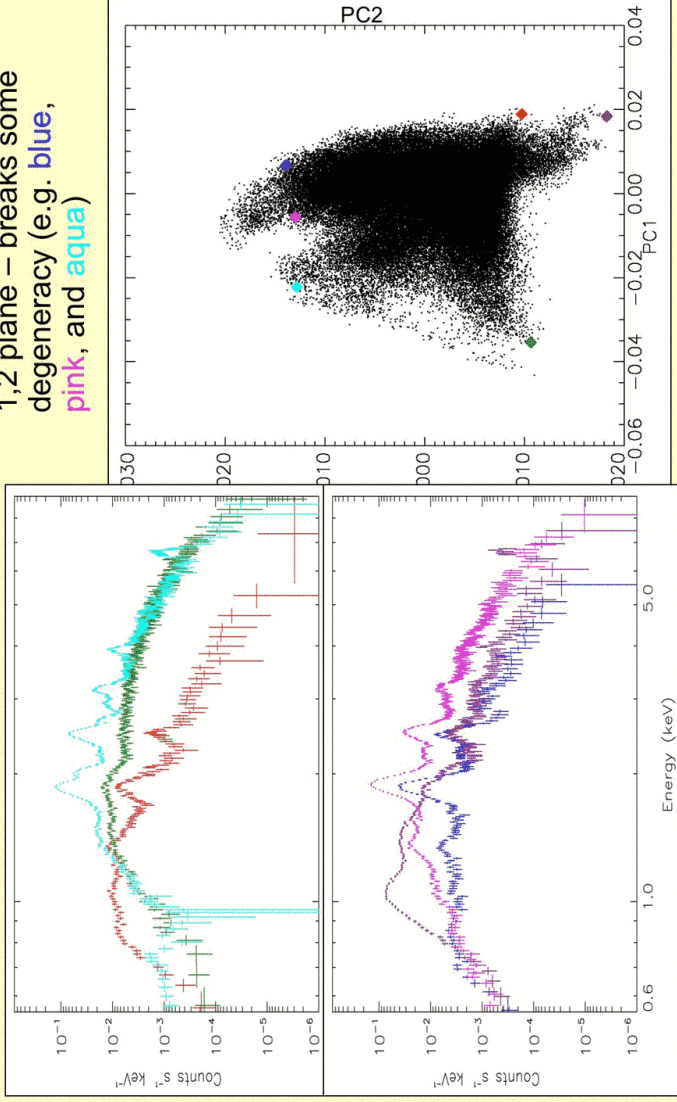


- Isolate certain features from integrated spectrum
- Each regions' spectrum is binned in this way



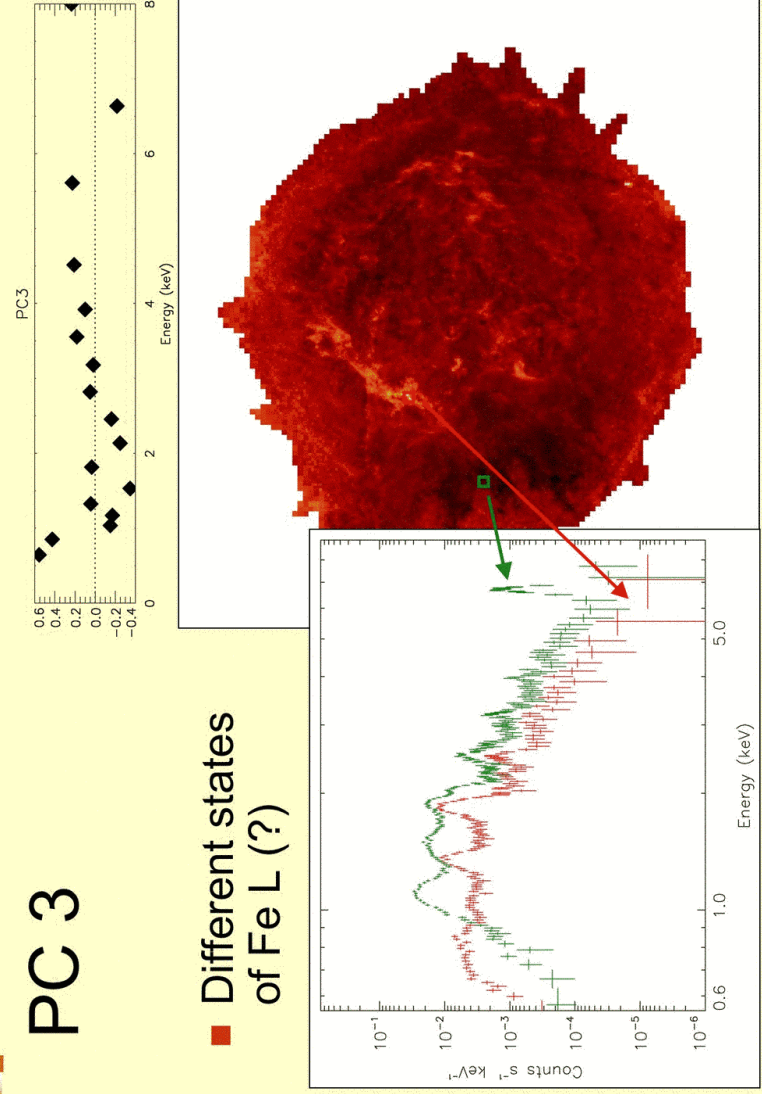
PC 1,2 Plane

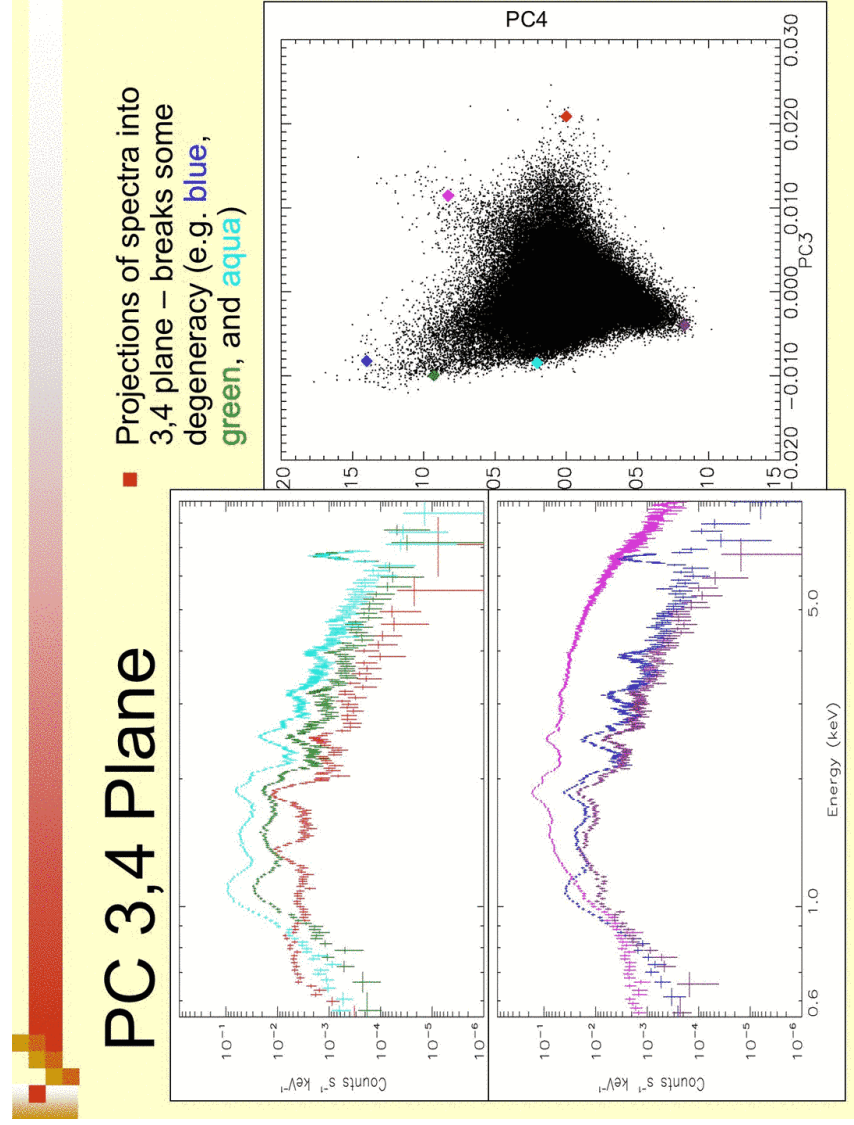
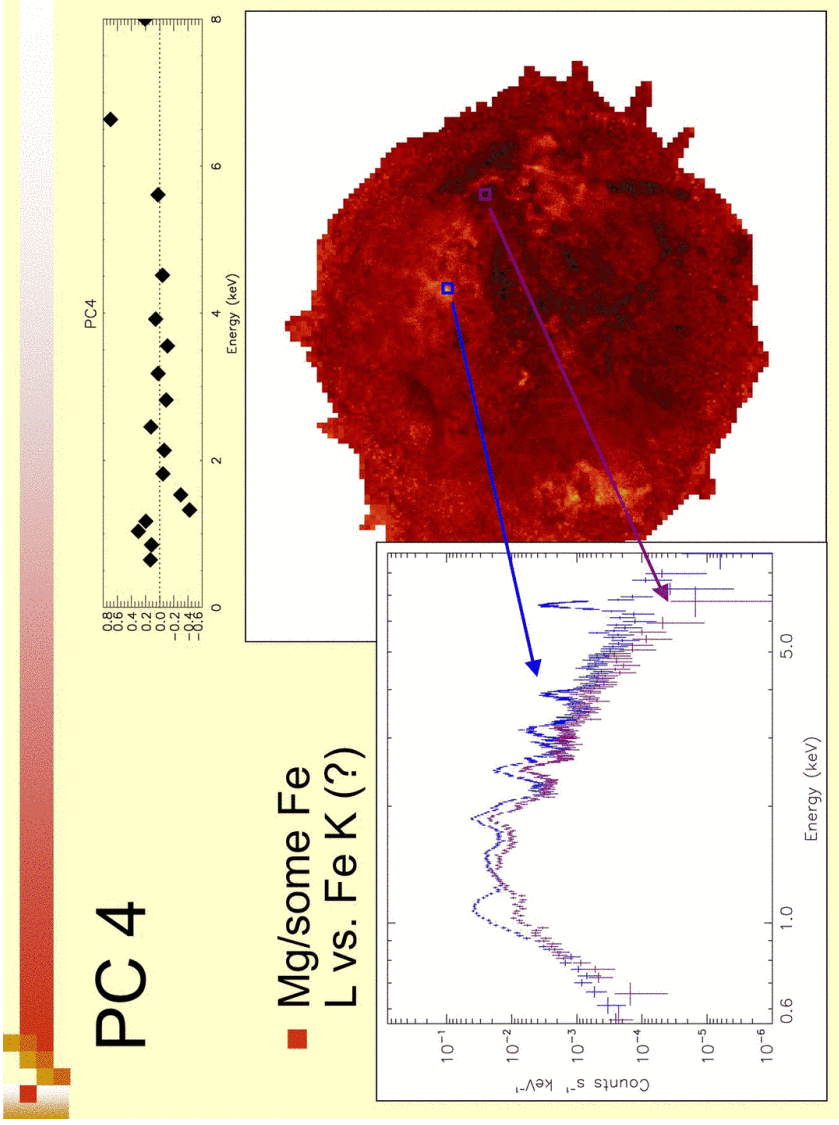
- Projections of spectra into 1,2 plane – breaks some degeneracy (e.g. blue, pink, and aqua)

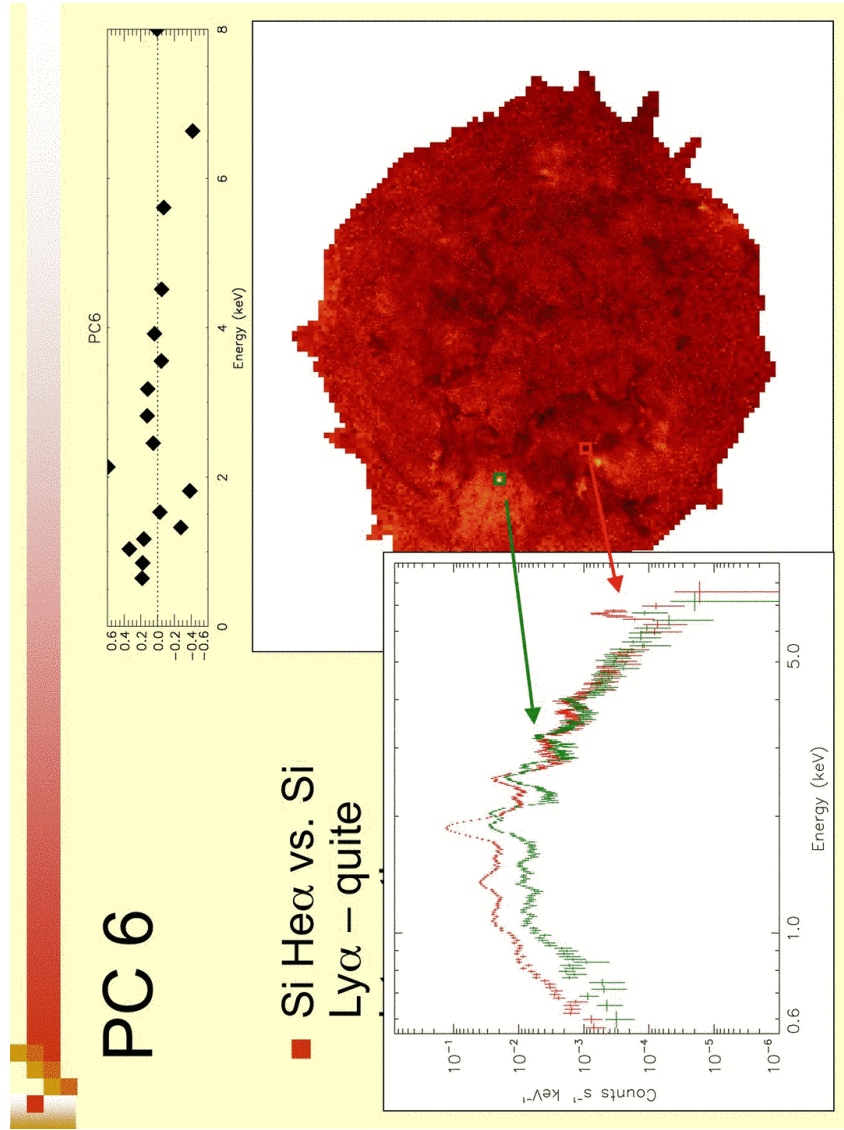
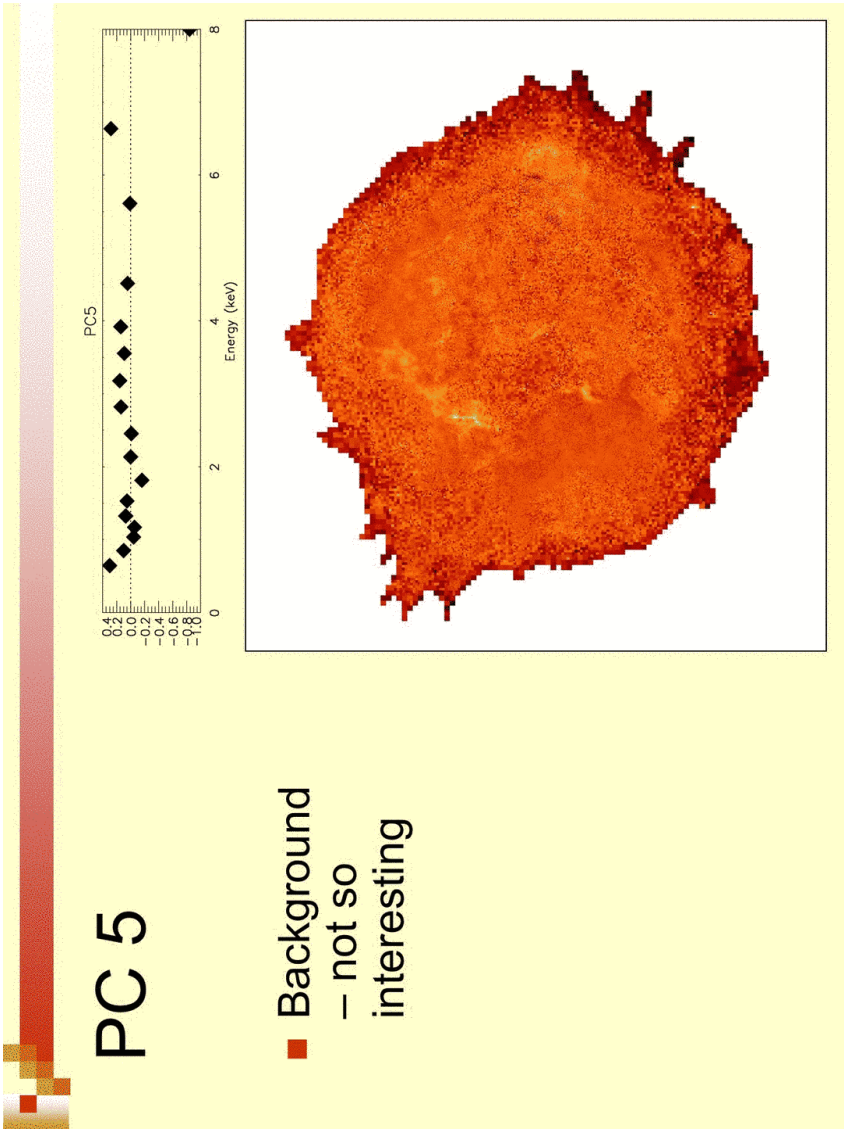


PC 3

- Different states of Fe L (?)

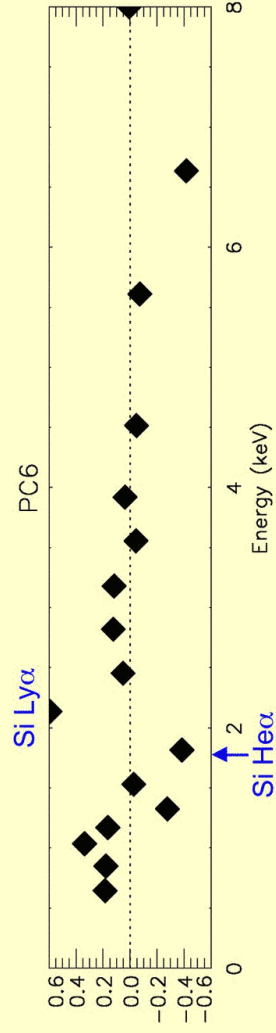






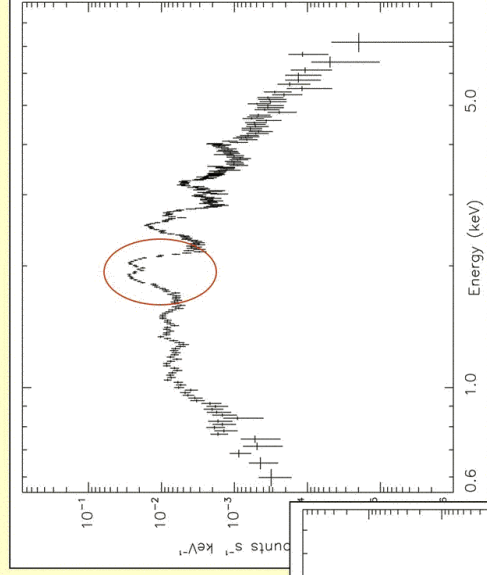
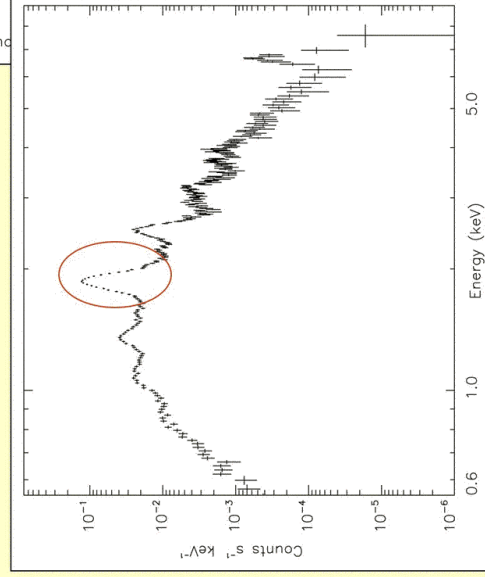
Where do we go from here?

- PCs are helpful:
 - Organize spectral classes
 - Provide spatial information
 - Indicate regions of potential interest
- PCs are complicated:
 - Not really a simple interpretation
 - Except, perhaps, for PC6 – start here



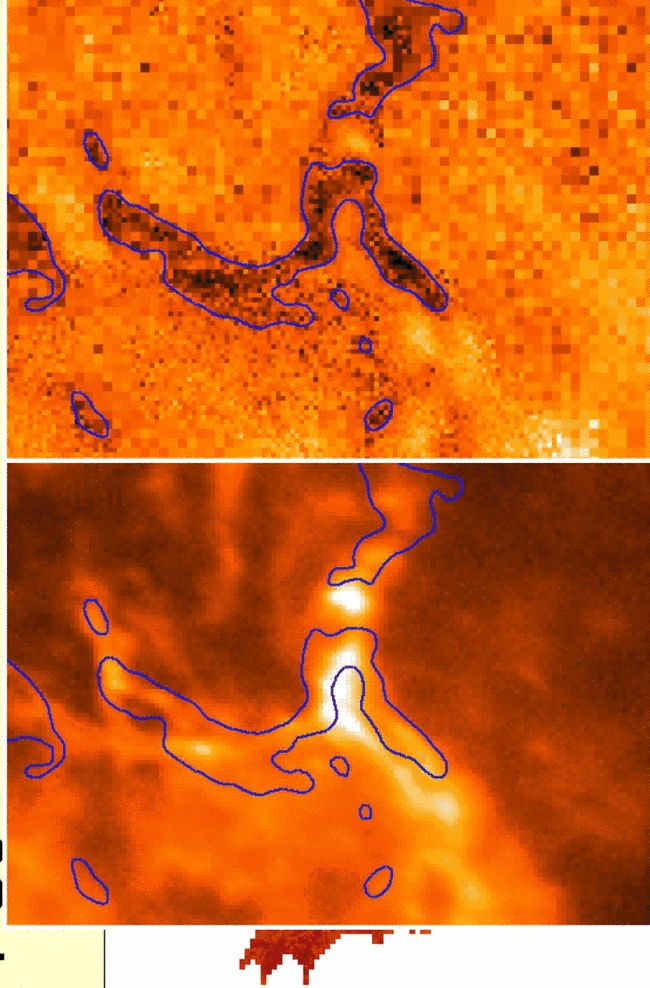
Spectra

- Nearly equal He α and Ly α fluxes (bright PC6 region) →



- He α much **stronger** than Ly α (dark PC6 region) →

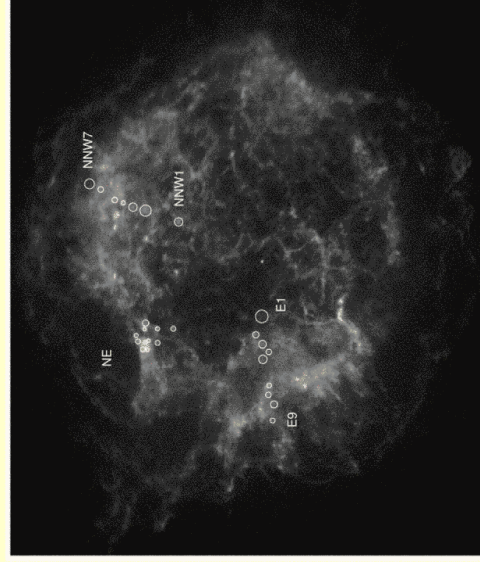
PC6 ■ Main variation in ■ Relate PC6 to



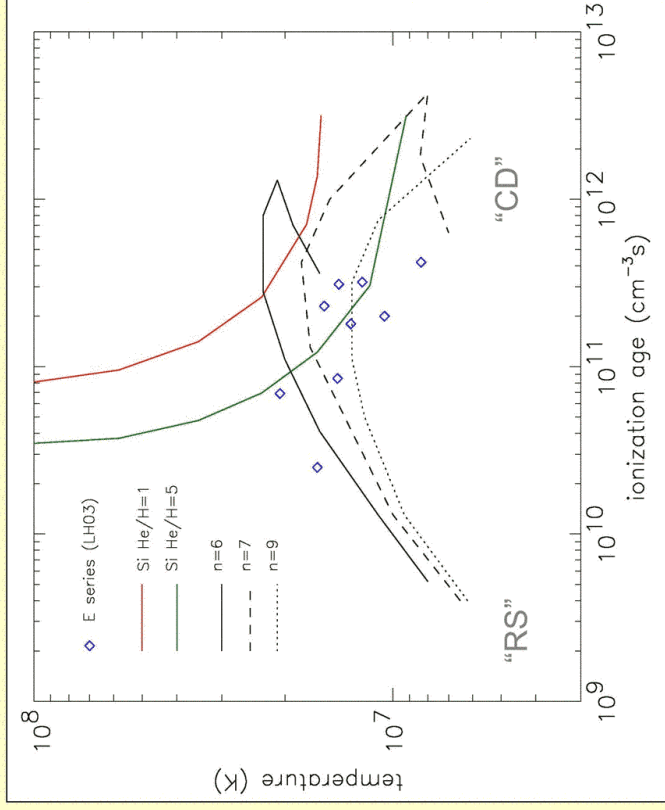
speaking?

Context – Laming & Hwang 2003

- Spectra from radial series of knots
 - Fit for temperature and ionization age
 - Compare to those values from models of a knot's thermodynamic evolution
 - Depends on ejecta density profile power law index

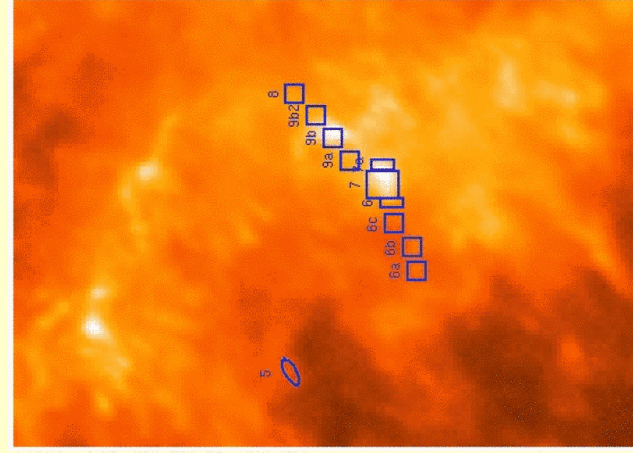


Temperature vs ionization age

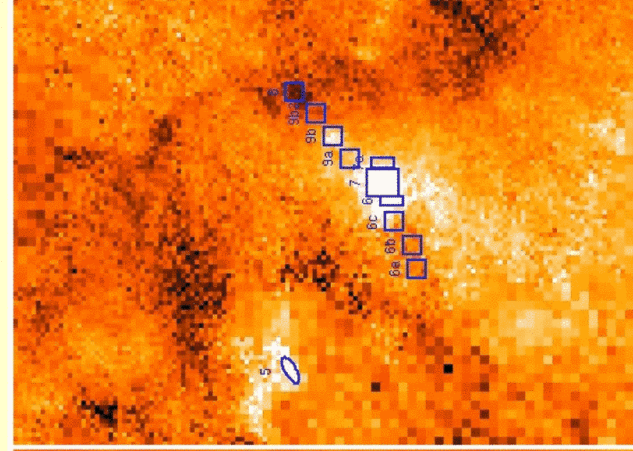


- Adapted from Fig. 6 of LH03
- E series agrees mainly with $n = 9$
- We will use PC6 to choose a similar radial series of knots

Silicon Valley



Broadband

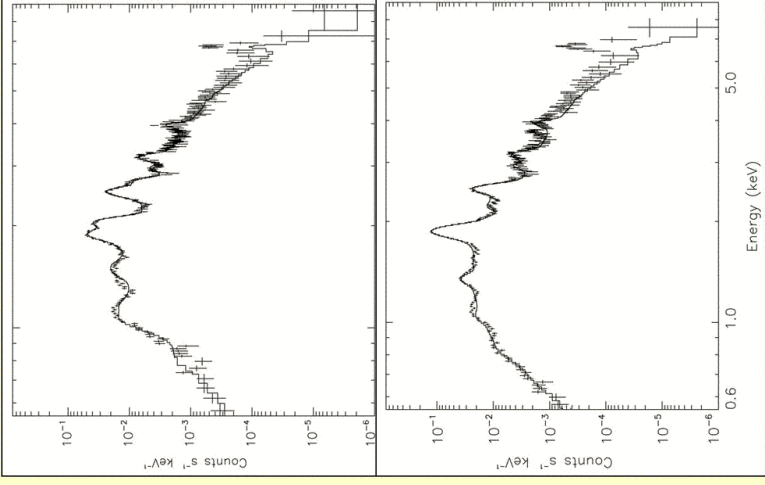


PC6

Spectral Fits

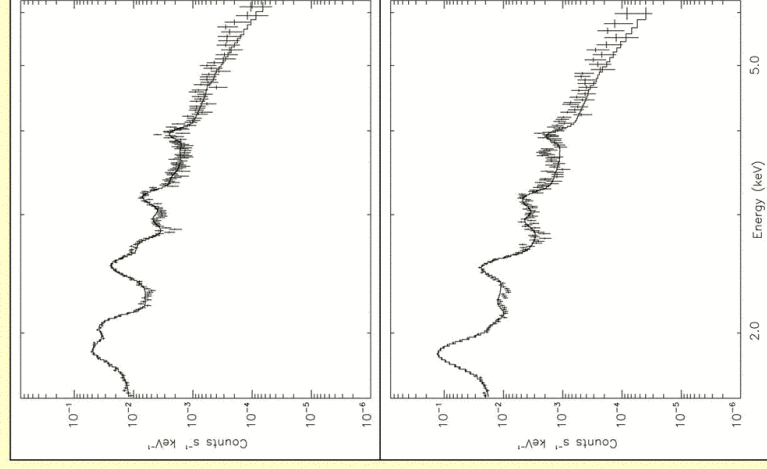
(Follow LH03)

- Single ionization timescale
- Oxygen continuum
- Fe L, K, and continuum not fit so well
- Look at band from Si to just below Fe K (1.6 – 6.25 keV)

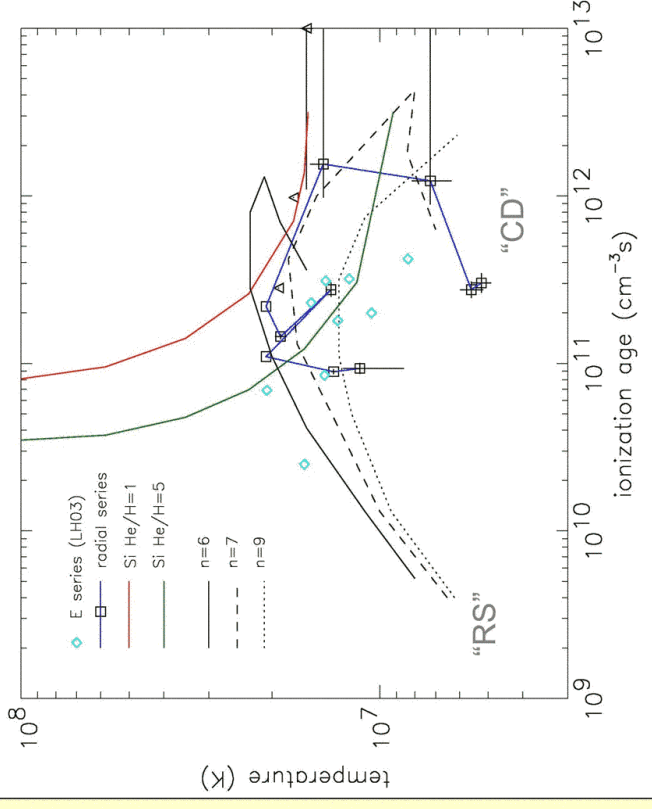


Assumptions

- Fe in spectra is likely due to projection
 - Treat as separate component
 - Ignore in high energy fits
- Temperature from continuum
- Ionization state from lines



Temperature vs ionization age



- Found knots with higher $n_e t$
- Radial series seems to agree better with $n = 7$
- Digesting results...

Preliminary Conclusions

- PC6 shows high vs. low Si ionization states
 - Only a few isolated areas with nearly equal He α and Ly α fluxes – why?
 - Possibly higher ionization state near jet – analyze another radial series there

