

Postdoc seminar
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From indoor localization to galaxy cluster detection

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Room 64

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Background and present

- PhD and postdoc at the Data Processing and Simulation Group, Universidad Politécnica de Madrid (Spain)
 - Indoor localization



- Postdoc at SAp, CEA
 - Detection of galaxy clusters



Outline

- Indoor localization



- Detection of galaxy clusters



Introduction

- Localization: Estimate the position of an object or person
- Localization and tracking of people and objects is fundamental for numerous applications
 - Emergency management
 - Domotic control
 - Ambient-assisted living
 - Augmented reality



Localization technologies

- **Outdoor** loc. technologies

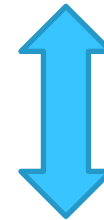
- GPS
- Cellular

- **Indoor** loc. technologies

- Radio
- Inertial sensors
- Ultrasound
- Infrared
- Computer vision
- Pressure sensors
- Electromagnetic sensing



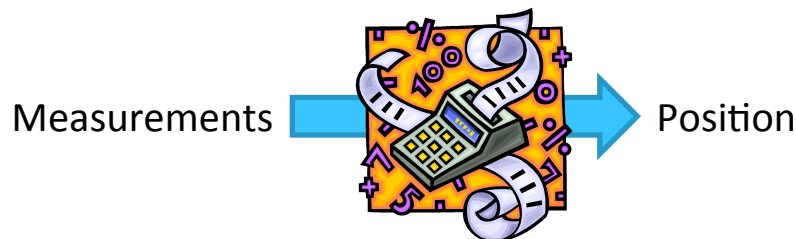
Localization accuracy



Energy consumption

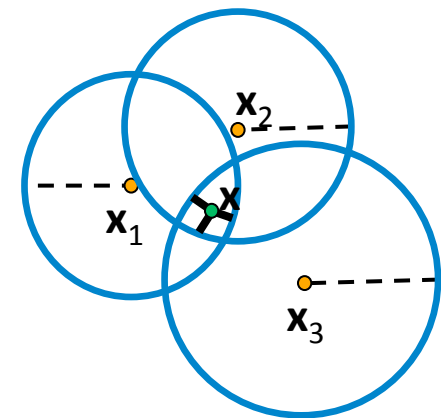
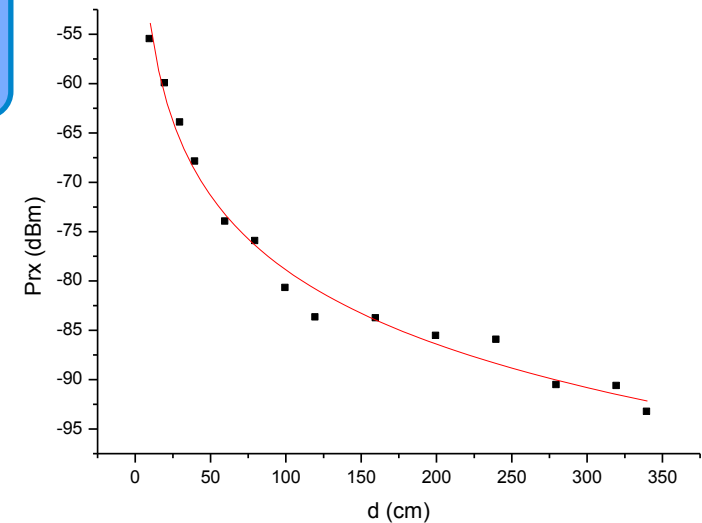
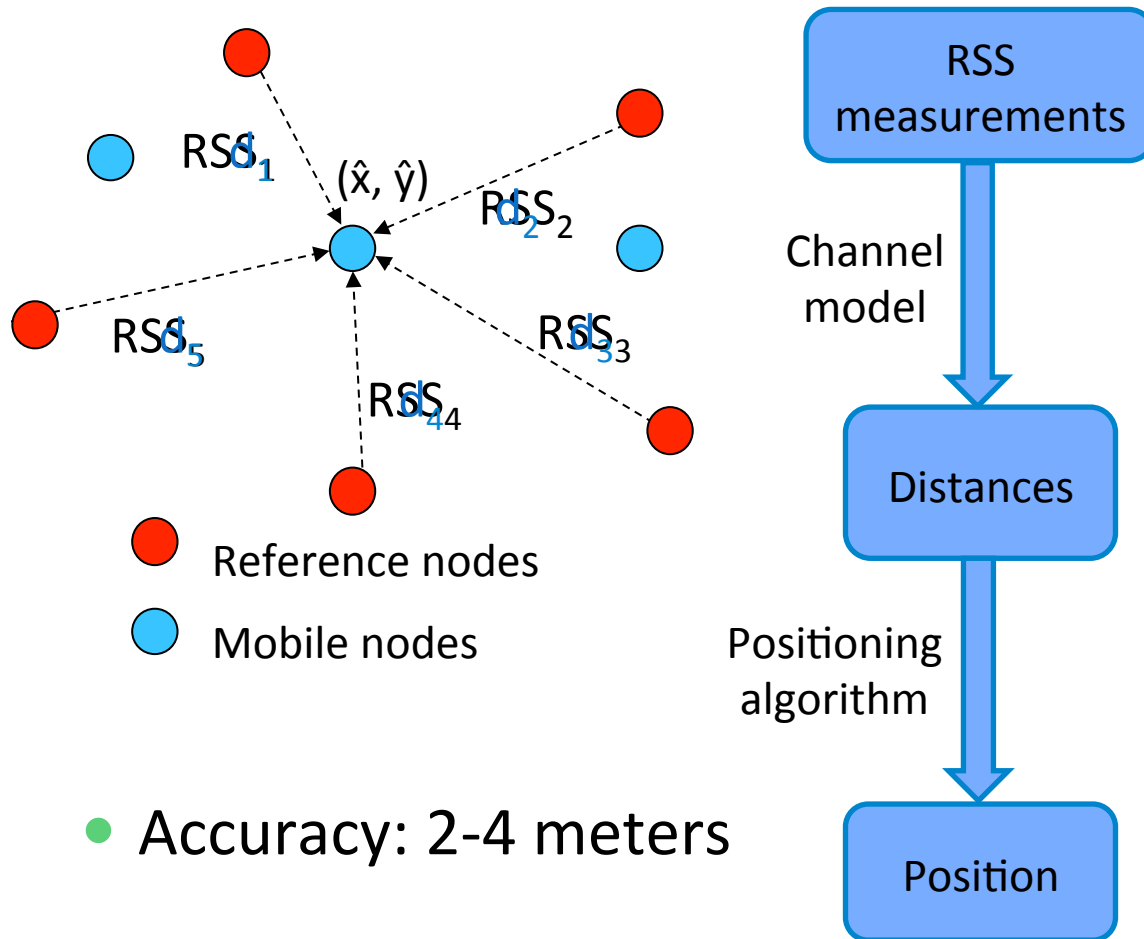
Radio-localization

- Typical deployment:
 - Reference nodes at know (fixed) positions
 - Mobile nodes whose position we want to calculate
- Communication phase: Measure some parameter of the radio signal
 - Connectivity (signal/no signal)
 - TOA/TDOA
 - AOA
 - RSS
- Processing phase



RSS-based localization system

- Channel modeling approach:



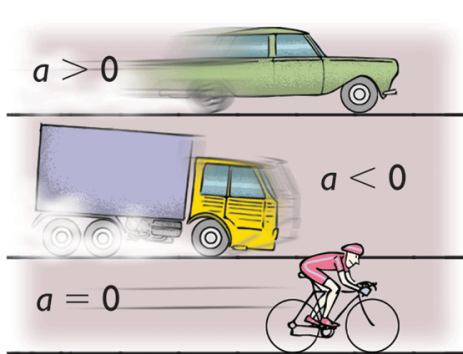
- Accuracy: 2-4 meters

Inertial sensors

- What are inertial sensors?

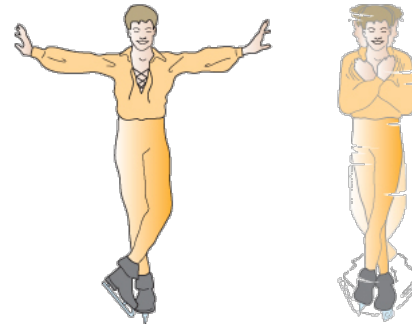
- Accelerometers

(measure acceleration)



- Gyroscopes

(measure angular speed)



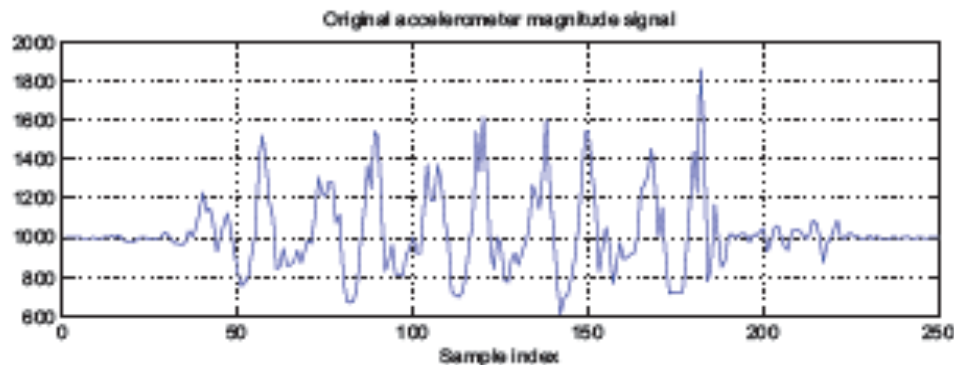
- Recently lightweight and small

- Thanks to MEMS technology
 - Used in mobile phones, smartphones, etc.



Inertial tracking systems

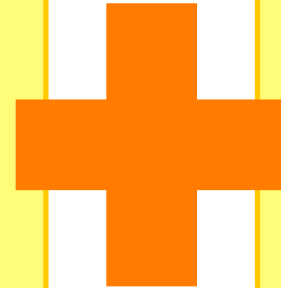
- Typical deployment:
 - The person/object to be tracked carries the inertial sensors
 - No fixed infrastructure is needed
- Localization is done via **dead reckoning**: calculating current position from
 - a previous position/speed
 - measurements collected by the sensors



- Problem: **Drift**: position error grows with time

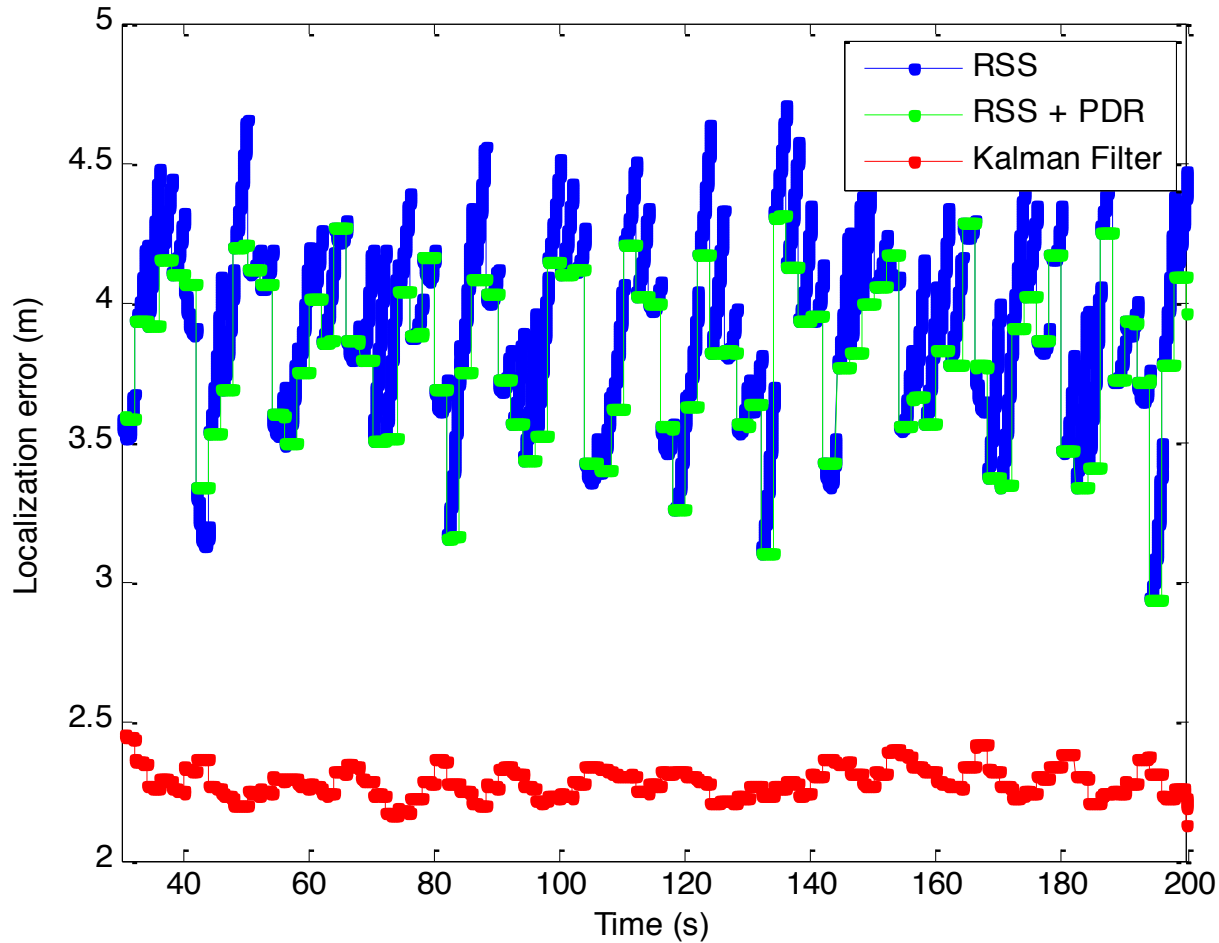
Hybrid localization systems

- RSS-based localization system
 - Medium accuracy
 - constant with time
- Inertial tracking system
 - Good accuracy
 - growing with time
- Others ...



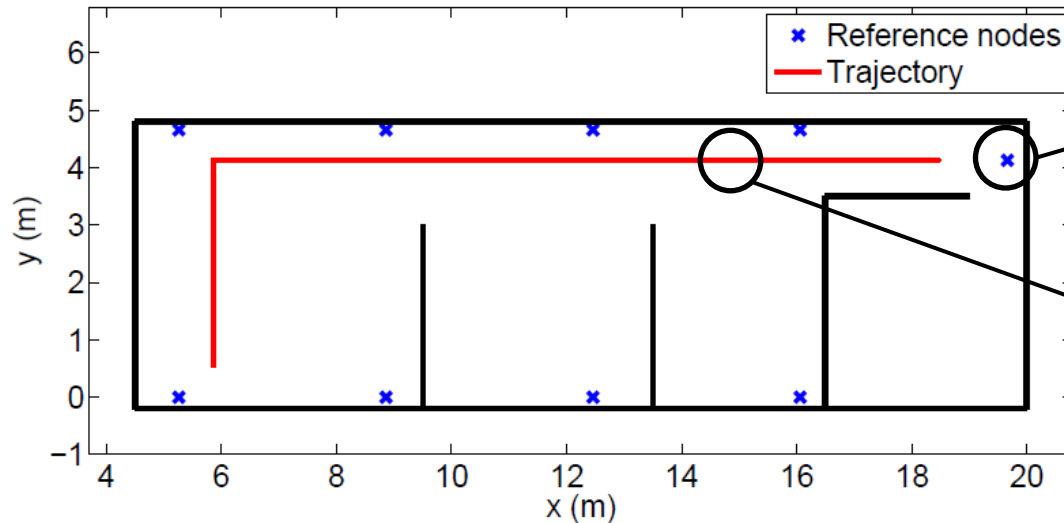
- Fusion algorithm
 - Kalman filter
 - Particle filter
 - ...

Simulation results



9 reference nodes
400 m² area

Experimental results



Average
localization
error

User	1	2	3	4	5	6	7	8
RSS	3.31	2.68	2.48	3.83	2.59	2.82	3.72	2.70
RSS + PDR	3.05	2.54	2.32	3.60	2.47	2.48	3.42	2.61
KF	2.39	2.26	1.64	2.92	2.30	2.22	2.63	2.22

Maximum
localization
error

User	1	2	3	4	5	6	7	8
RSS	11.06	6.45	8.03	10.60	6.58	5.98	7.75	5.87
RSS + PDR	11.17	6.20	8.03	9.93	6.33	5.97	7.67	5.61
KF	3.91	4.36	4.02	4.30	4.65	4.33	3.92	4.62

Outline

- Indoor localization

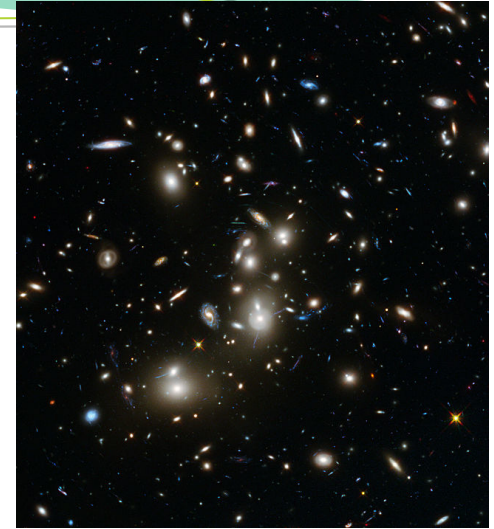


- Detection of galaxy clusters



Galaxy clusters

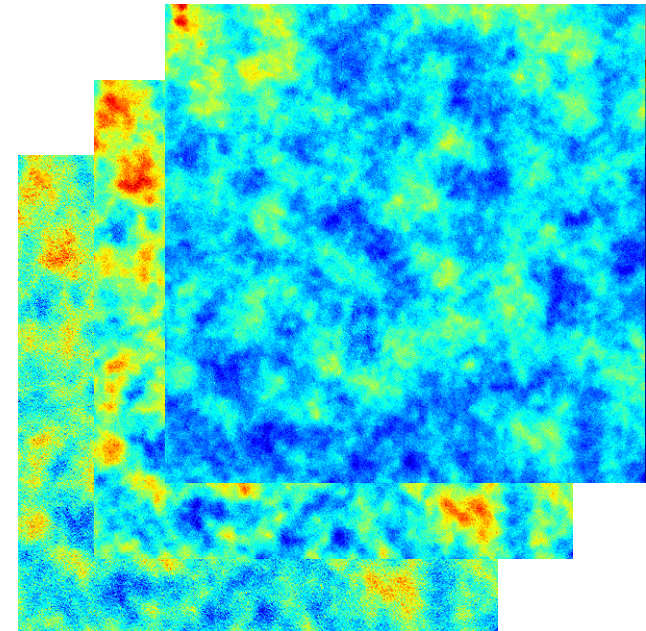
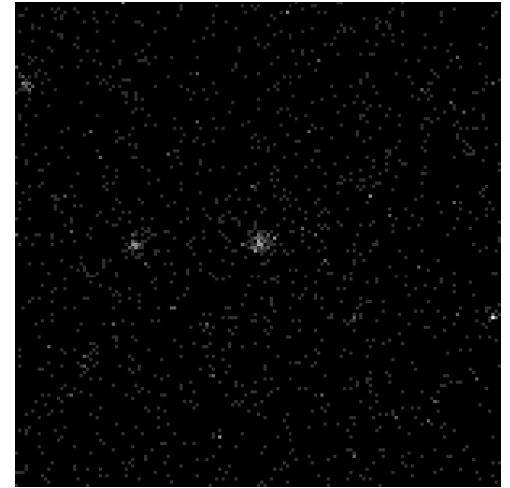
- Galaxy clusters:
 - 100s/1000s galaxies bound by gravity
 - Hot ionized gas
 - Dark matter
- The history of structure formation in the Universe depends on the cosmology -> studying cluster samples at different redshifts can constrain cosmological parameters
- Goal: build a sample of massive clusters up to $z \sim 1$



Detection of galaxy clusters

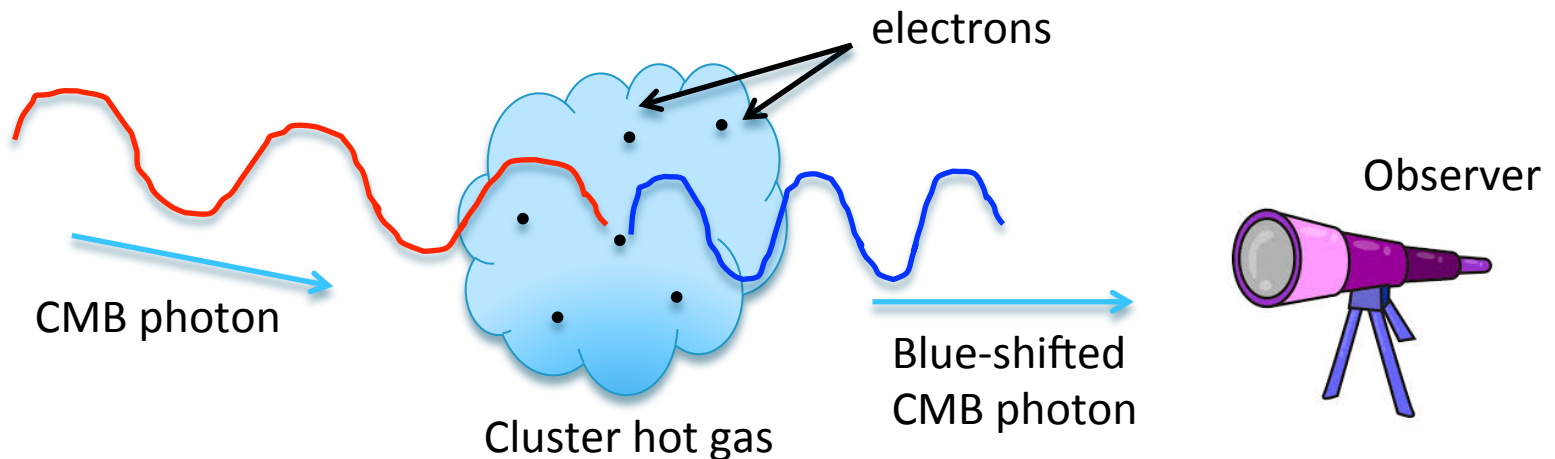
- The hot ionized gas of the inter-cluster medium
 - emits thermal radiation in the **X-ray** band
 - distorts the cosmic microwave radiation via the **Sunyaev-Zeldovich** effect
- We are developing an algorithm to combine these two complementary sources of information
 - improve the cluster detection rate
 - reduce the number of false detections
- SZ cluster detection is undertaken via a Matched Multi-Filter (MMF)*

* Melin et al. 2006

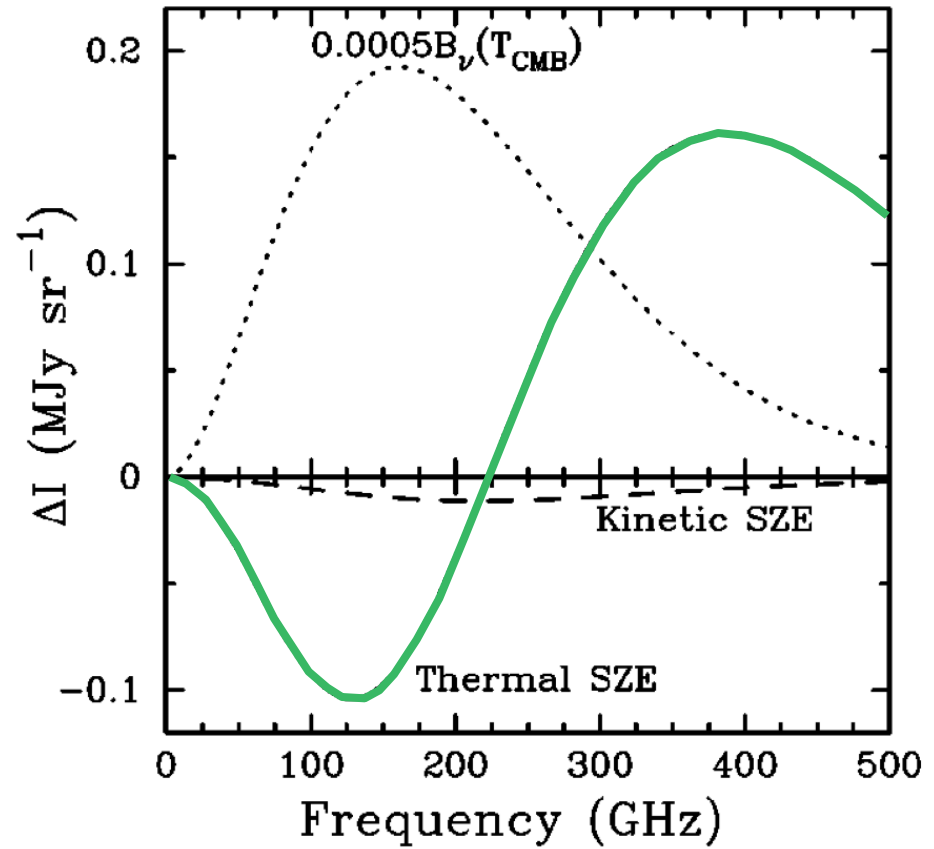
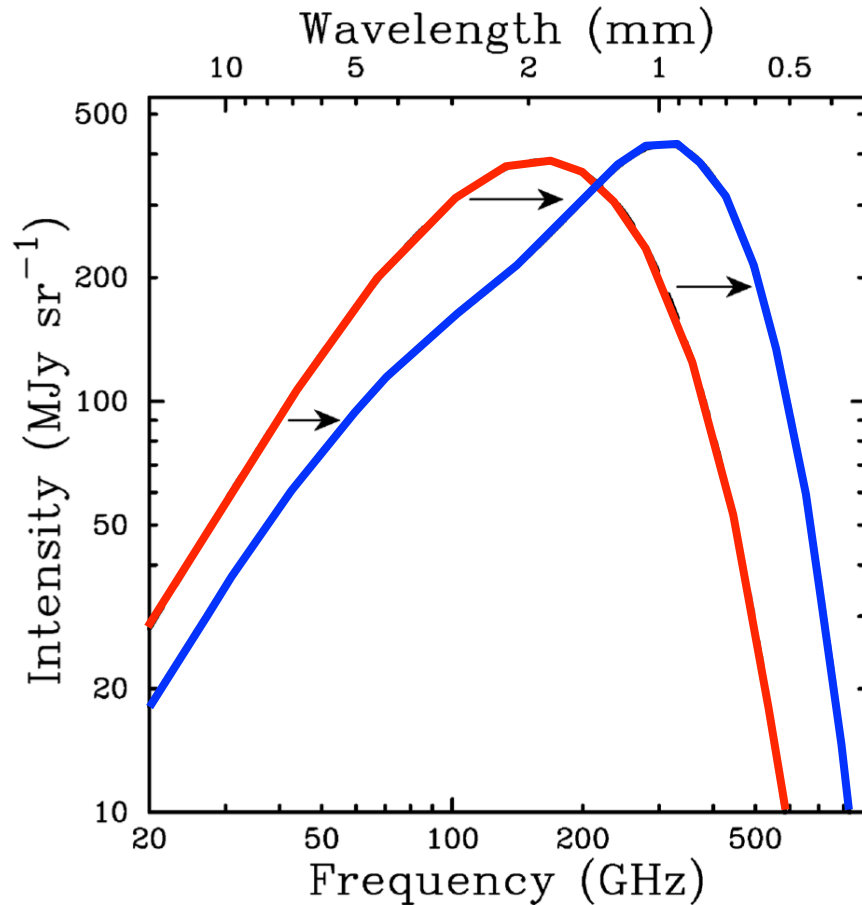


The SZ effect

- The SZ effect is caused by the hot gas in galaxy clusters
- CMB photons gain energy when they interact with the electrons of the gas



Spectral distortion



Spectral distortion

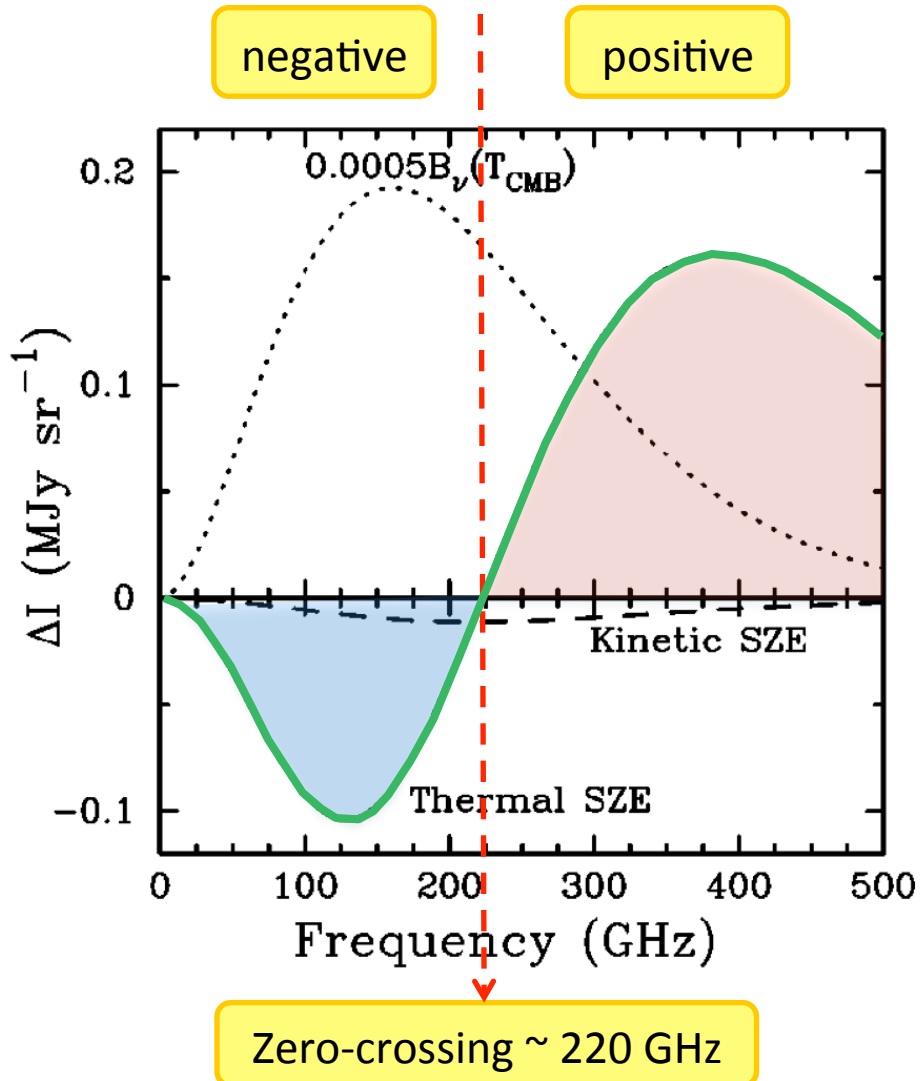
- Universal form
- Amplitude given by the Compton y parameter (integral of the gas pressure along the line-of-sight)

Compton y parameter

Universal shape

$$\Delta i_\nu = y \cdot j_\nu$$

$$y \propto \int_{los} n_e \cdot T_e \cdot dl$$



Observed signal

- Cluster brightness profile at observation frequency ν :

$$\Delta i_\nu(\mathbf{x}) = y(\mathbf{x}) \cdot j_\nu$$

Compton y parameter
at position \mathbf{x}

SZ spectral function at
frequency ν

$$y(\mathbf{x}) = y_0 \cdot T_{\theta_c}(\mathbf{x})$$

Normalized
cluster profile

Integral of p_e
 $p_e \sim \text{NFW profile}$

- Observed maps contain noise + cluster at \mathbf{x}_0

$$M_\nu(\mathbf{x}) = y_0 \cdot j_\nu \cdot T_{\theta_c}(\mathbf{x} - \mathbf{x}_0) + N_\nu(\mathbf{x})$$

$$\mathbf{M}(\mathbf{x}) = y_0 \cdot \mathbf{j}_\nu \cdot T_{\theta_c}(\mathbf{x} - \mathbf{x}_0) + \mathbf{N}(\mathbf{x})$$

known signal
(if size known)

noise

Toy example

- 2 frequencies

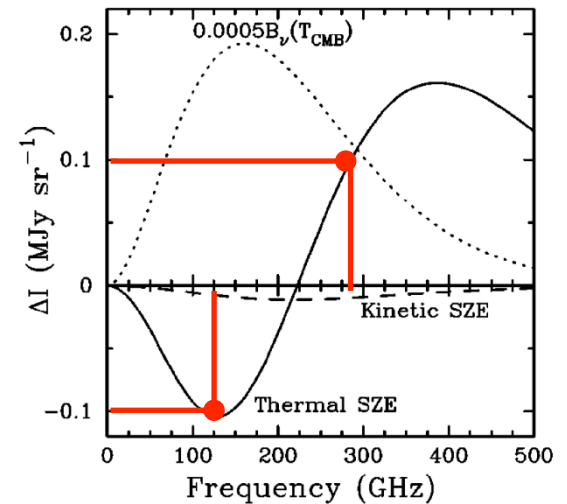
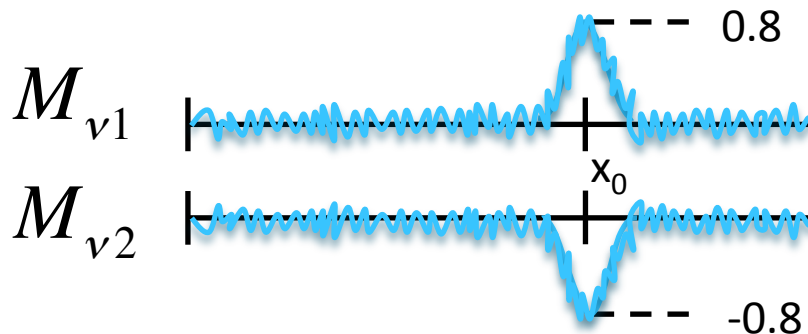
$$j_{\nu 1} = +0.1$$

$$j_{\nu 2} = -0.1$$

- Integration of p_e :

$$\int_0^8 p_e \, dx = 8 * \int_0^1 p_e \, dx$$

- Observed maps:



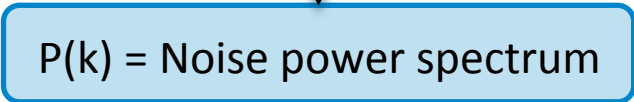
Matched filter

- The minimum variance unbiased estimator of the profile amplitude y_0 is:


$$\hat{y}_0(\mathbf{x}_0) = \sum \sum \Psi_{\theta_c}^t(\mathbf{x} - \mathbf{x}_0) \cdot \mathbf{M}(\mathbf{x}) \cdot d\mathbf{x}$$

where:

$$\Psi_{\theta_c}(\mathbf{k}) = \frac{\mathbf{P}^{-1}(\mathbf{k}) \cdot \mathbf{j}_v \cdot T_{\theta_c}(\mathbf{k})}{\sum \sum [\mathbf{j}_v \cdot T_{\theta_c}(\mathbf{k})]^t \cdot \mathbf{P}^{-1}(\mathbf{k}) \cdot \mathbf{j}_v \cdot T_{\theta_c}(\mathbf{k}) \cdot d\mathbf{k}}$$



$P(k)$ = Noise power spectrum



Known signal

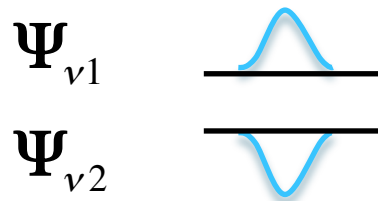
Toy example

- Suppose uncorrelated IID Gaussian noise

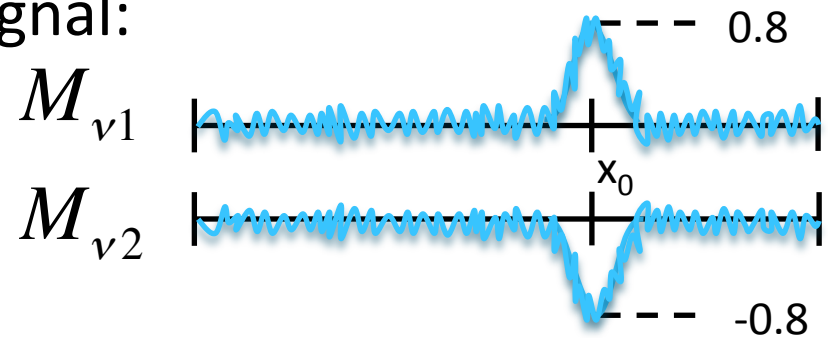
$$\sim \mathcal{N}(0, \sigma^2) \Rightarrow \mathbf{P}(\mathbf{k}) = \sigma^2 \cdot \mathbf{I} \Rightarrow \Psi_{\theta_c}(\mathbf{k}) \propto \mathbf{j}_v \cdot T_{\theta_c}(\mathbf{k})$$

Known signal

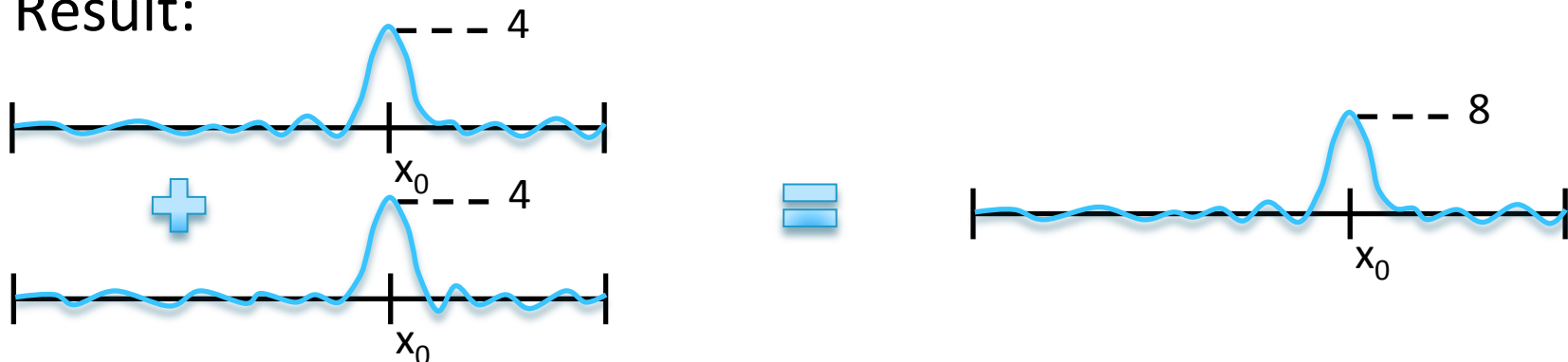
- Filter:



- Signal:

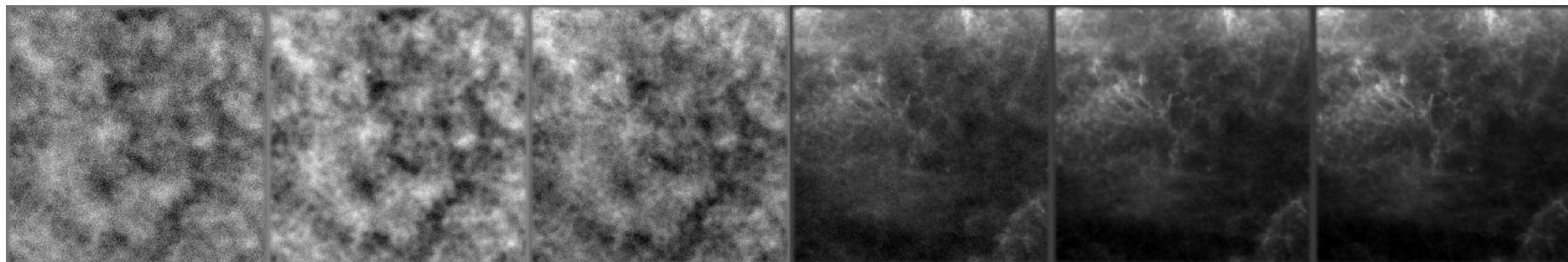


- Result:

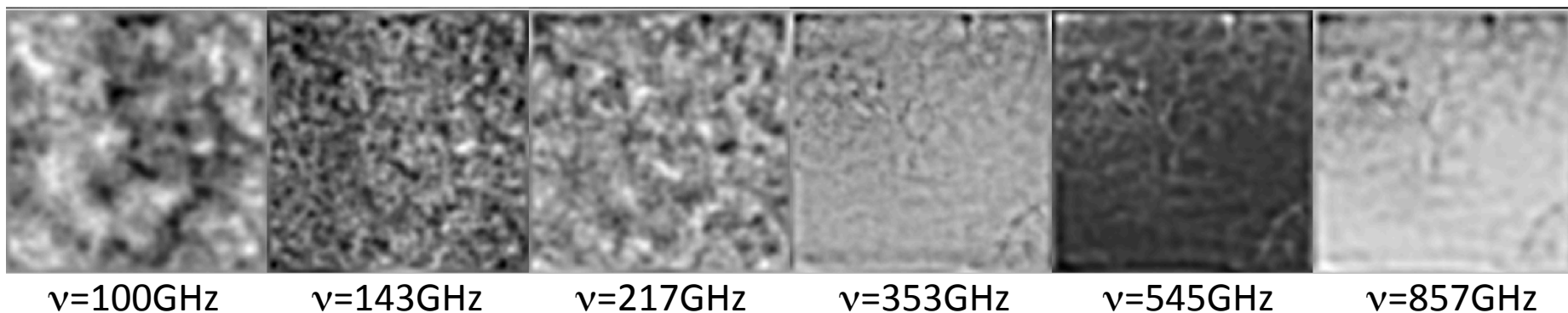


Example

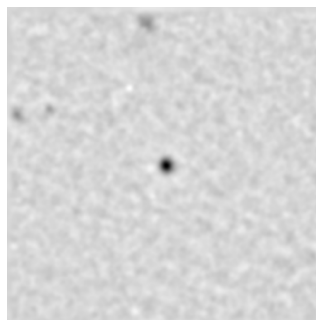
Observed maps at each frequency



Filtered maps at each frequency

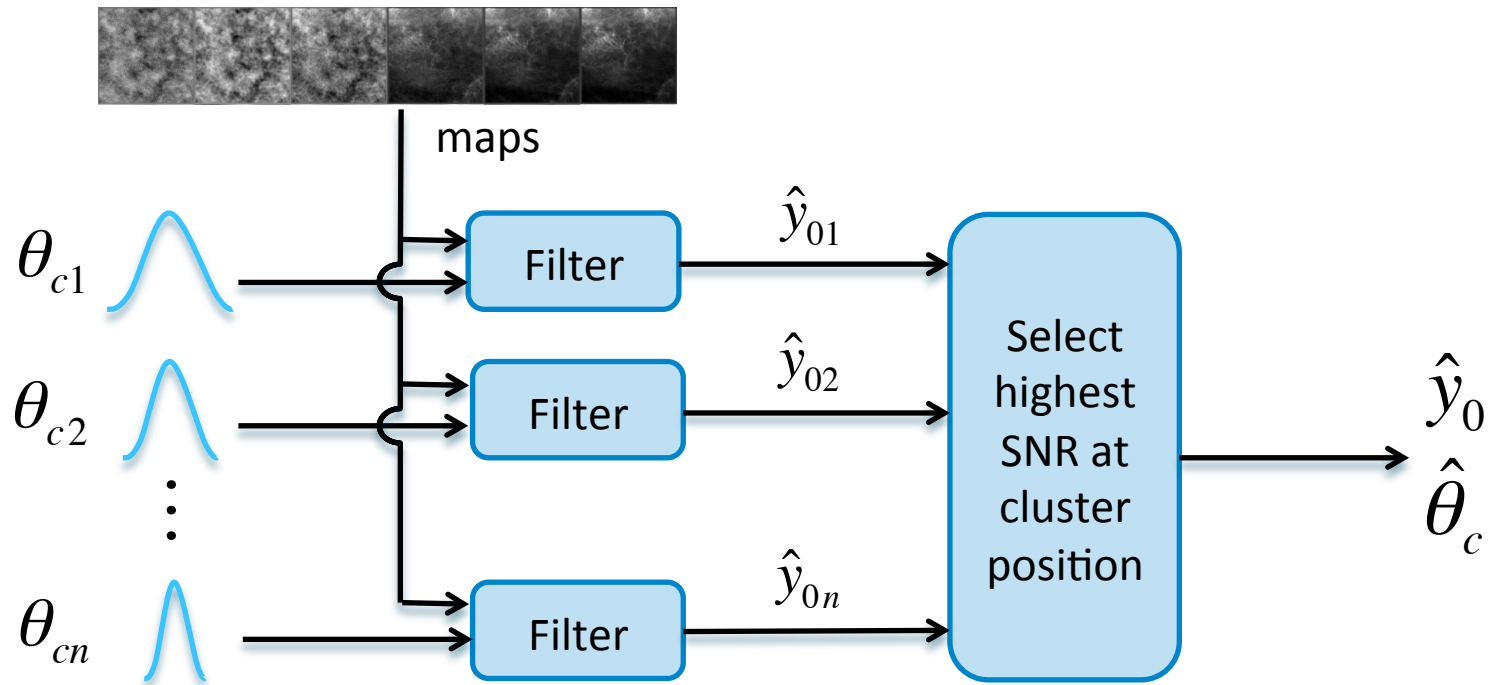


Filtered map



Unknown cluster size

- Create normalized cluster profiles $T_{\theta_{ci}}$ for various cluster sizes
- Filter image for each cluster size
- Select result with higher SNR



The X-ray map

- Emission measure is proportional to n_e^2

- Cluster brightness profile:

$$m(\mathbf{x}) = cte \cdot T_{\theta_c}(\mathbf{x})$$

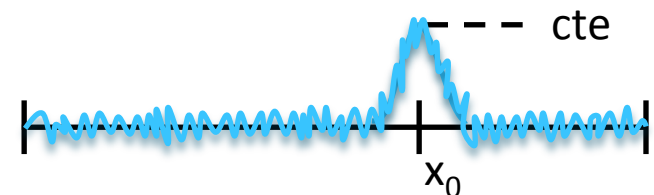
Integral of n_e^2
 $n_e^2 \sim \text{NFW profile}$

- Observed maps contain noise + cluster at \mathbf{x}_0

$$M(\mathbf{x}) = cte \cdot T_{\theta_c}(\mathbf{x} - \mathbf{x}_0) + N(\mathbf{x})$$

known signal
(if size known)

noise



MMF for X-ray map

- The X-ray detection problem is equivalent to the SZ detection problem (known signal with unknown amplitude under noise)
- We can apply the same filter
- But we need to take into account the differences
 - The cluster profile is different
 - We have Poisson noise

Combined MMF

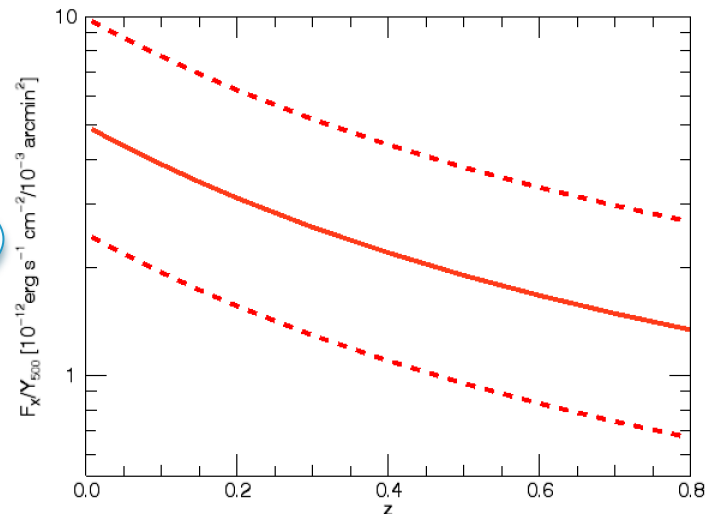
- The SZ MMF takes as input 6 frequency maps
- We add the X-ray map as a 7th frequency map
- To do this we need to express the two maps in the same units

X-ray flux
[erg/s/cm²]



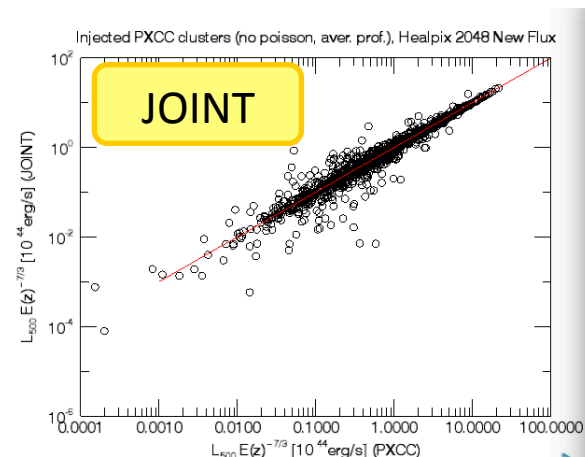
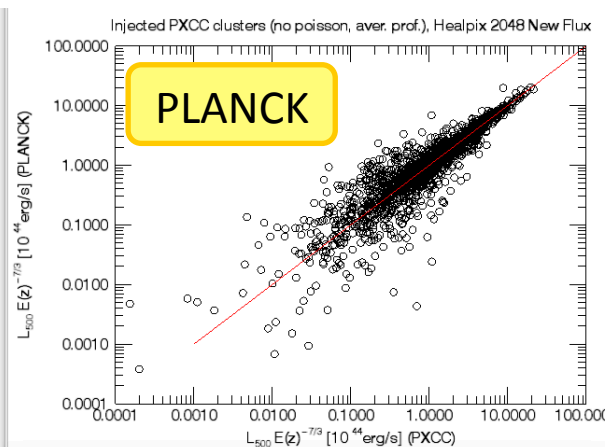
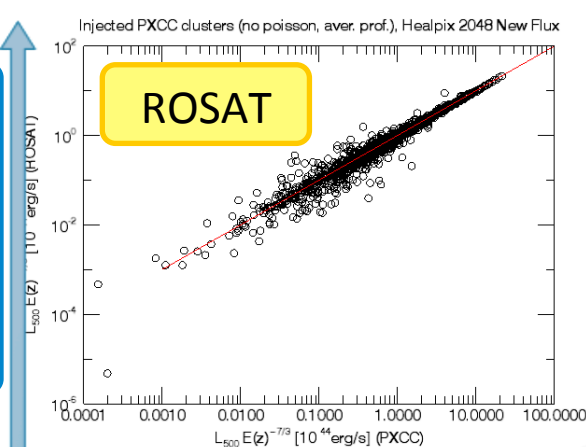
Y₅₀₀ [arcmin²]

$$\frac{F_X [\text{erg} \cdot \text{s}^{-1} \cdot \text{cm}^{-2}]}{Y_{500} [\text{arcmin}^2]} = 4.95 \cdot 10^{-9} \cdot E(z)^{5/3} \cdot (1+z)^{-4} \cdot K(z)$$



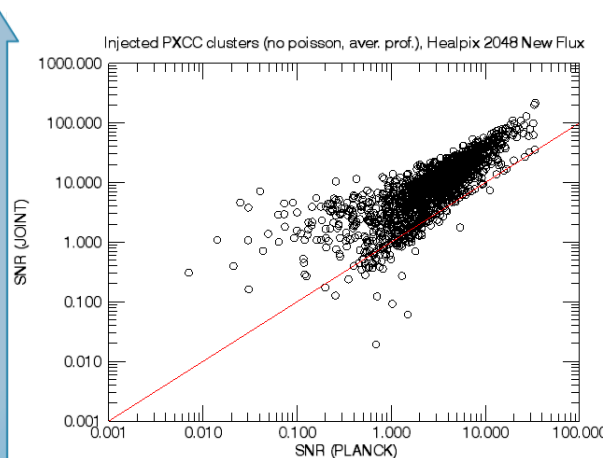
Ideal simulation of PXCC clusters

Extracted flux



Injected flux

SNR (JOINT)



SNR (PLANCK only)

Ideal PXCC clusters injected

- using average profile
- no Poisson fluctuations

Questions?

