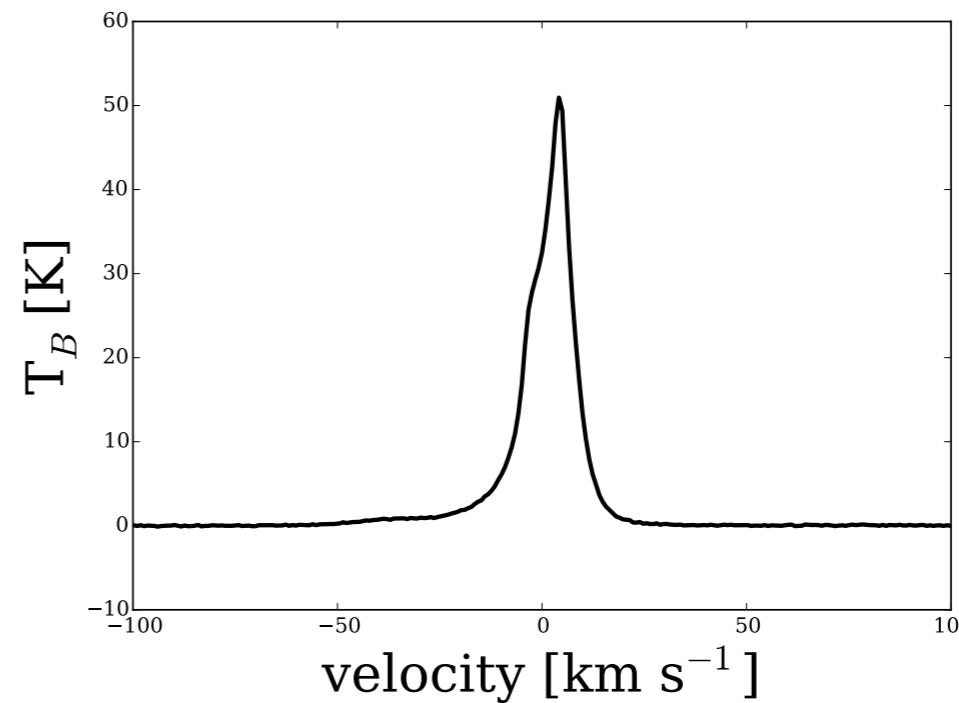
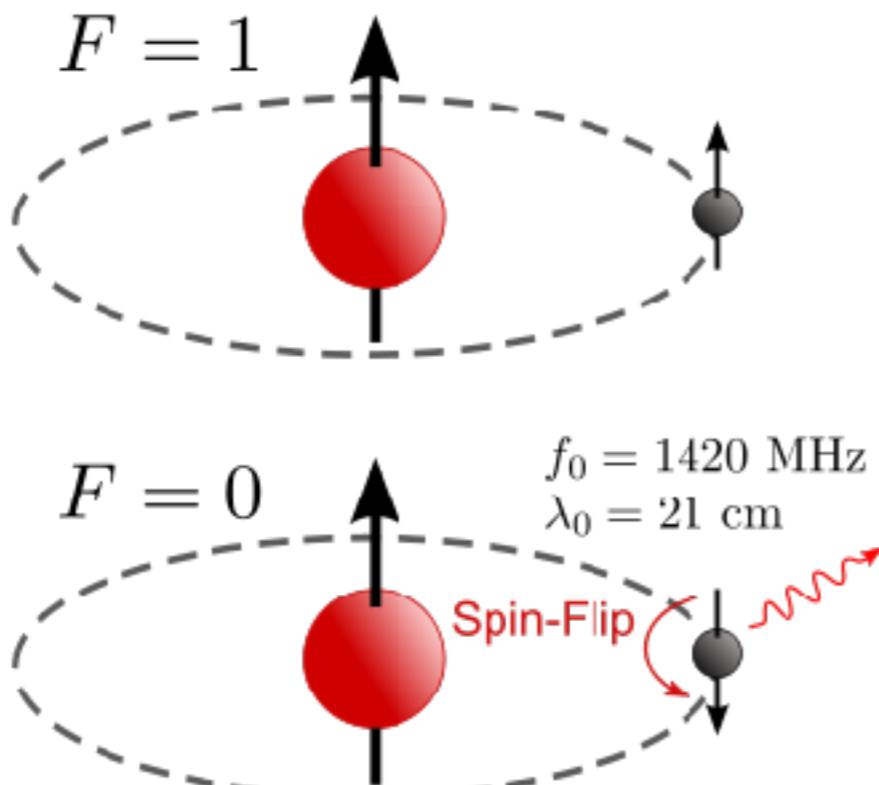




HELGA DÉNES

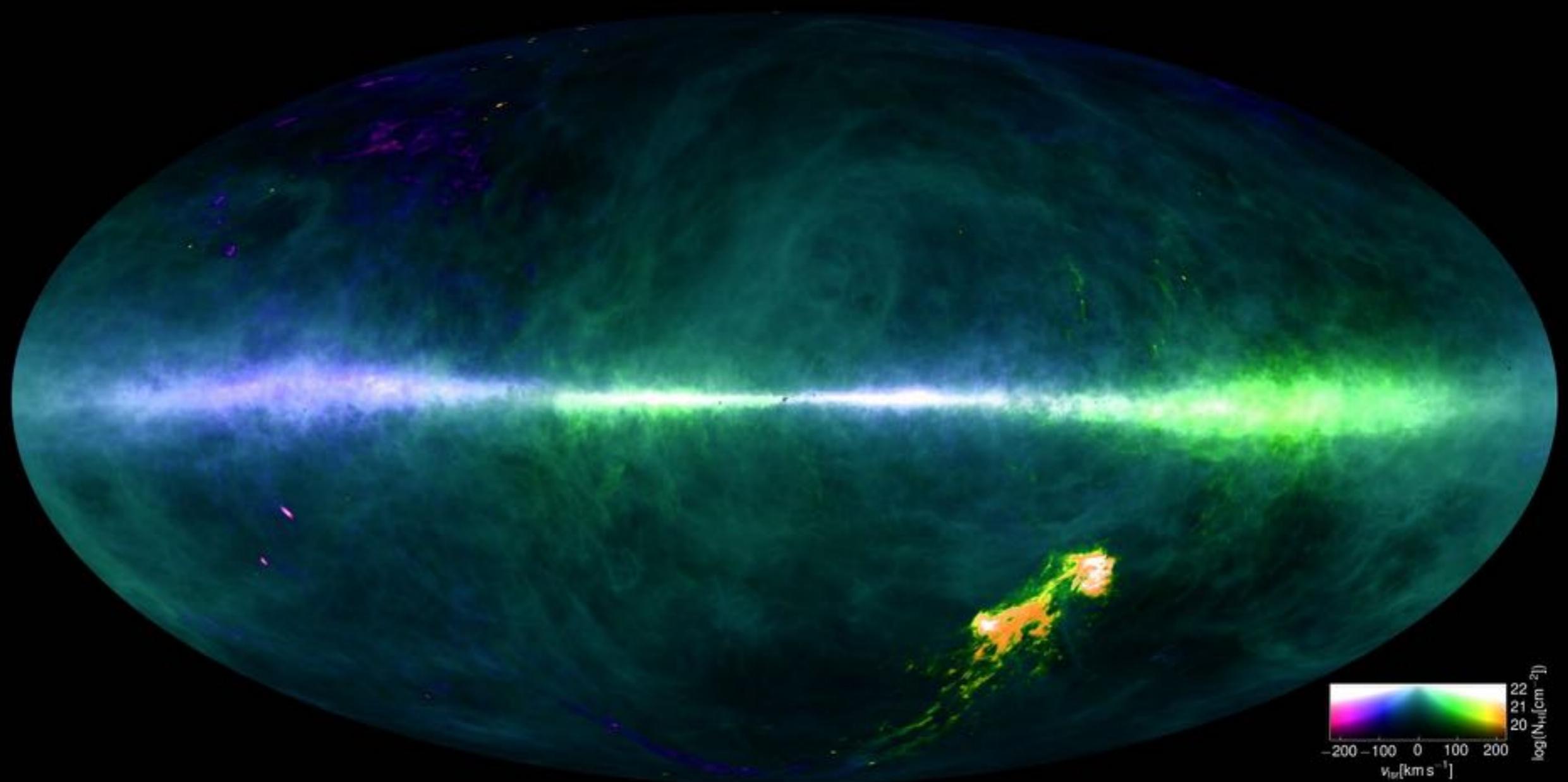
**GAUSSIAN DECOMPOSITION OF
SPECTRAL LINES
AND NEUTRAL HYDROGEN IN GALAXIES**

NEUTRAL ATOMIC HYDROGEN (HI) EMISSION



- Hydrogen is the most common element in the Universe
- The neutral atomic hydrogen (HI) emits at 21 cm
- HI is the fuel for potential star formation
- HI is also the lightest atom and very sensitive to disturbances

ATOMIC COLD GAS - HI IN THE MILKY WAY



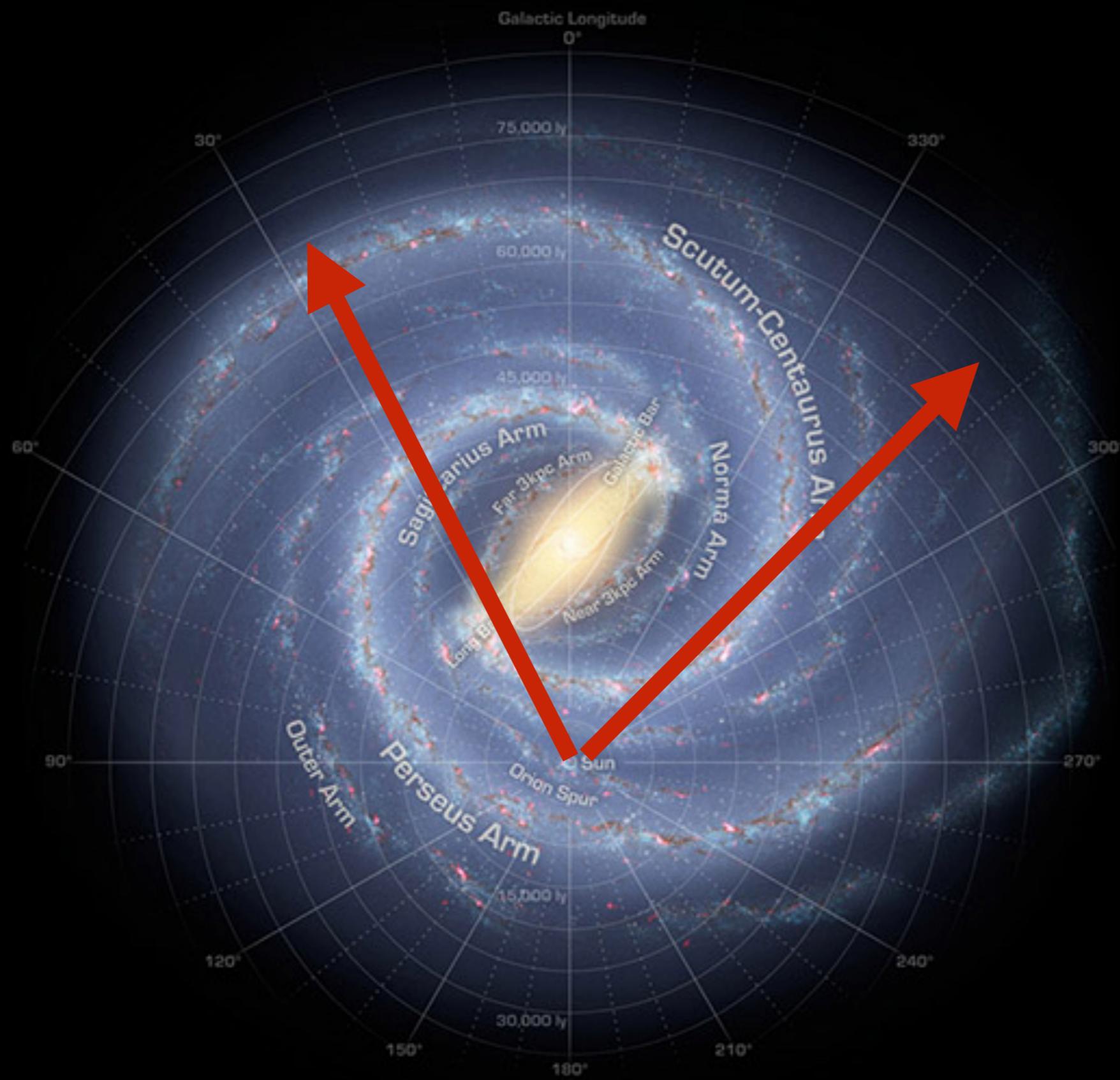
HI4PI, <http://www.icrar.org/hi4pi/>

THE GALFA HI SURVEY



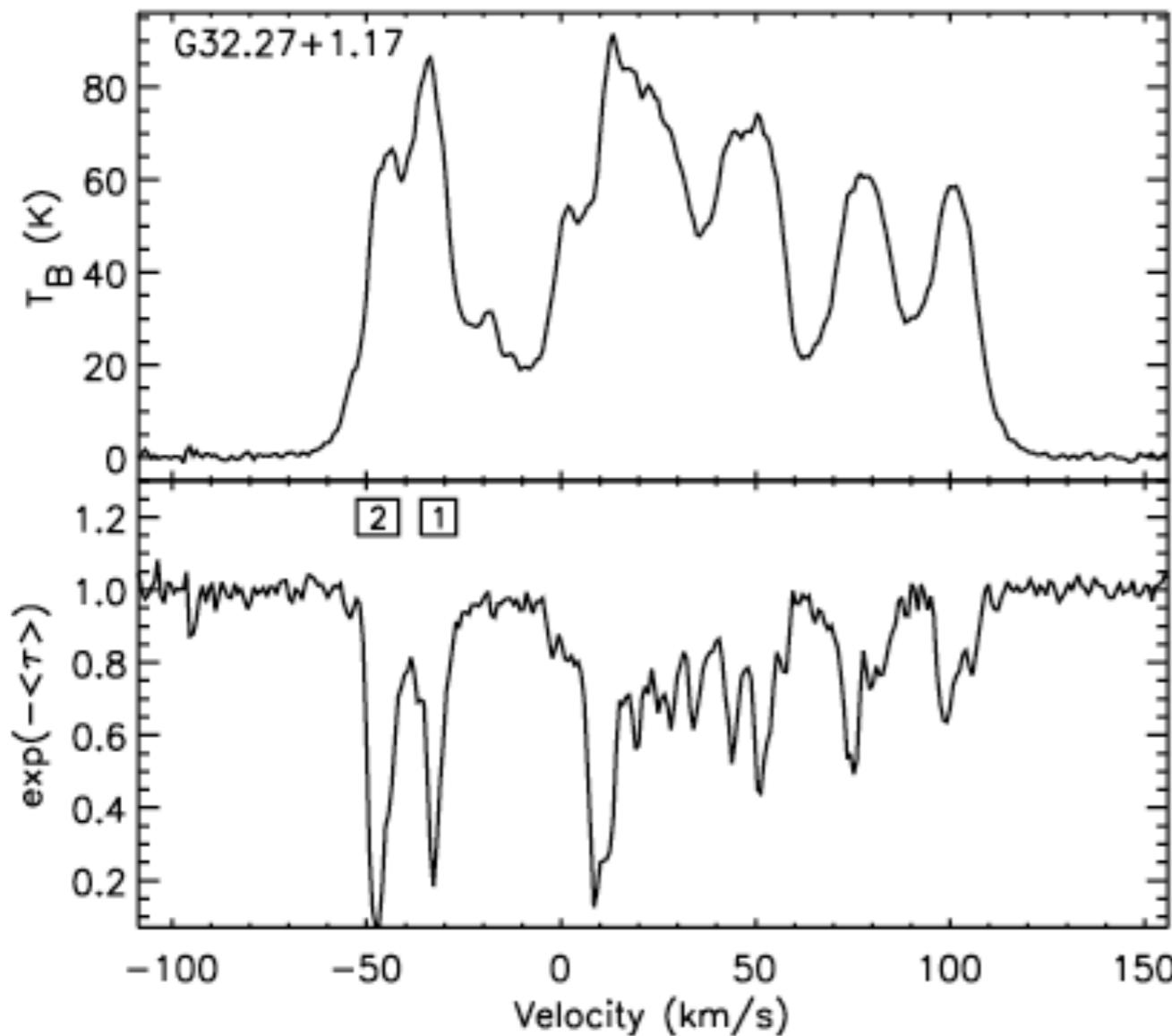
GALFA DR2, Peek et al. in prep, (3' resolution with Arecibo)

THE MILKY WAY



THE INTERSTELLAR MEDIUM

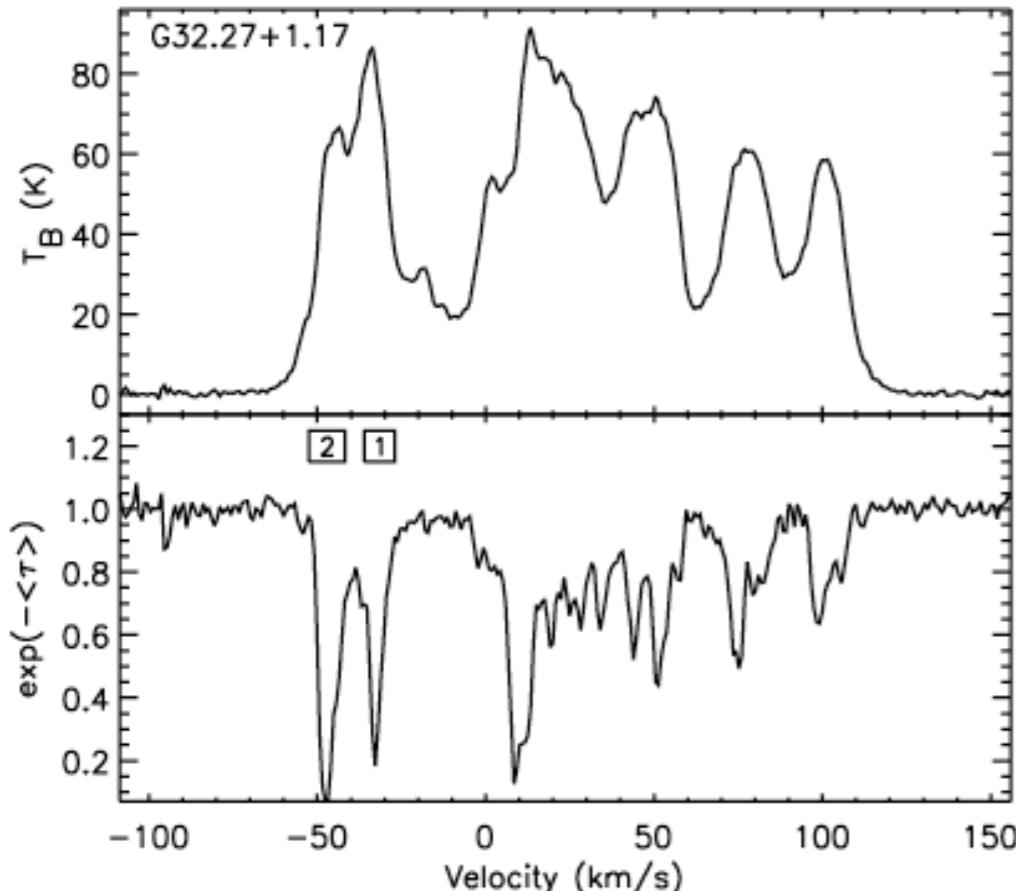
HI emission



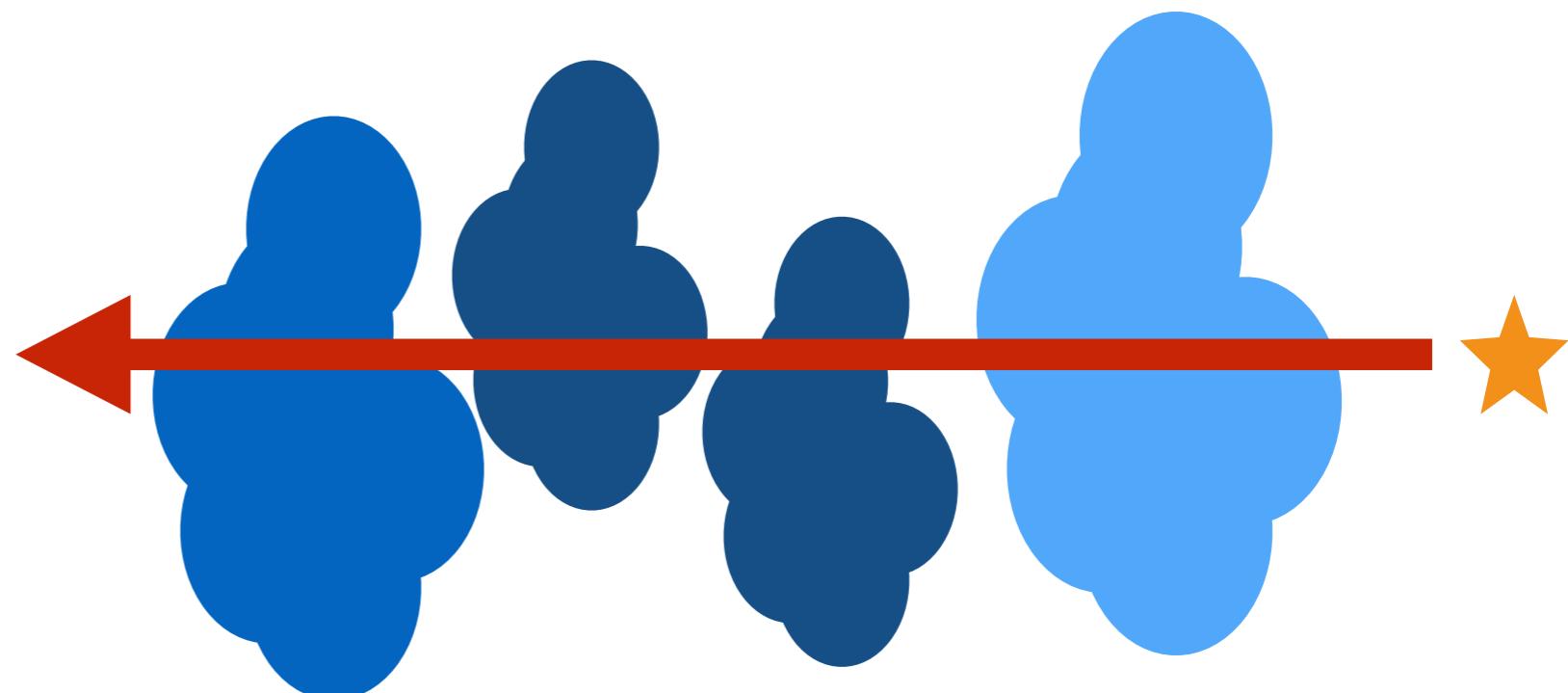
- ▶ 1. **WNM** (warm neutral medium) optically thin, 21 cm emission, $T > 200\text{K}$
- ▶ 2. **CNM** (cold neutral medium) optically thick, 21 cm absorption, $T \sim 50\text{ K}$, high volume-gas density, at 10^{20} cm^{-2} HI is saturating and converted into H₂

HI absorption

HI ABSORPTION SPECTRA



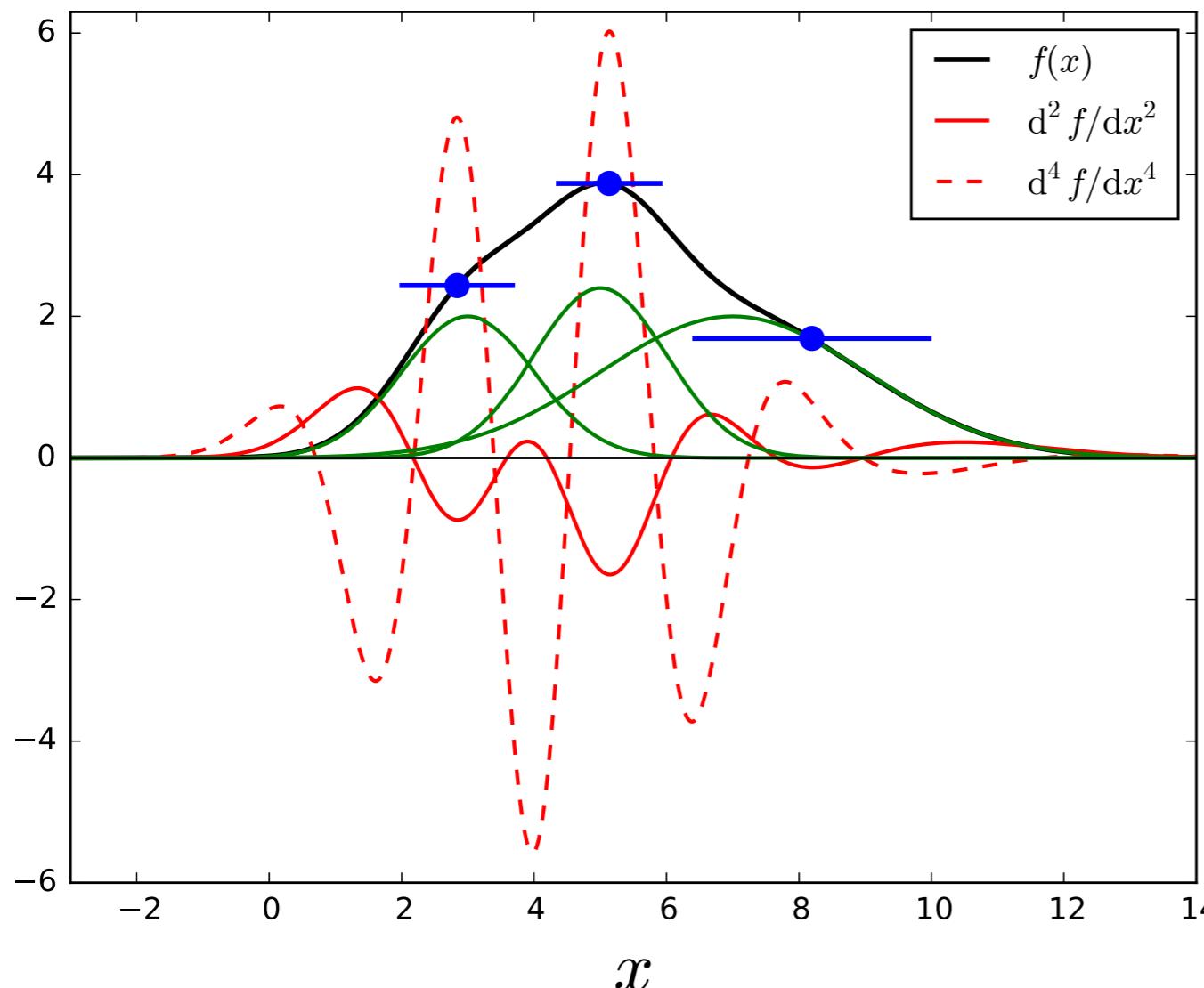
distant continuum
source



cold HI clouds in absorption

- ▶ each cloud corresponds to a Gaussian line
- ▶ strongly blended multiple component profiles
- ▶ automated decomposition is needed!

GAUSSIAN DECOMPOSITION



Guess criteria:

$$f > 0$$

$$d^2 f / df^2 < 0$$

$$d^3 f / df^3 = 0$$

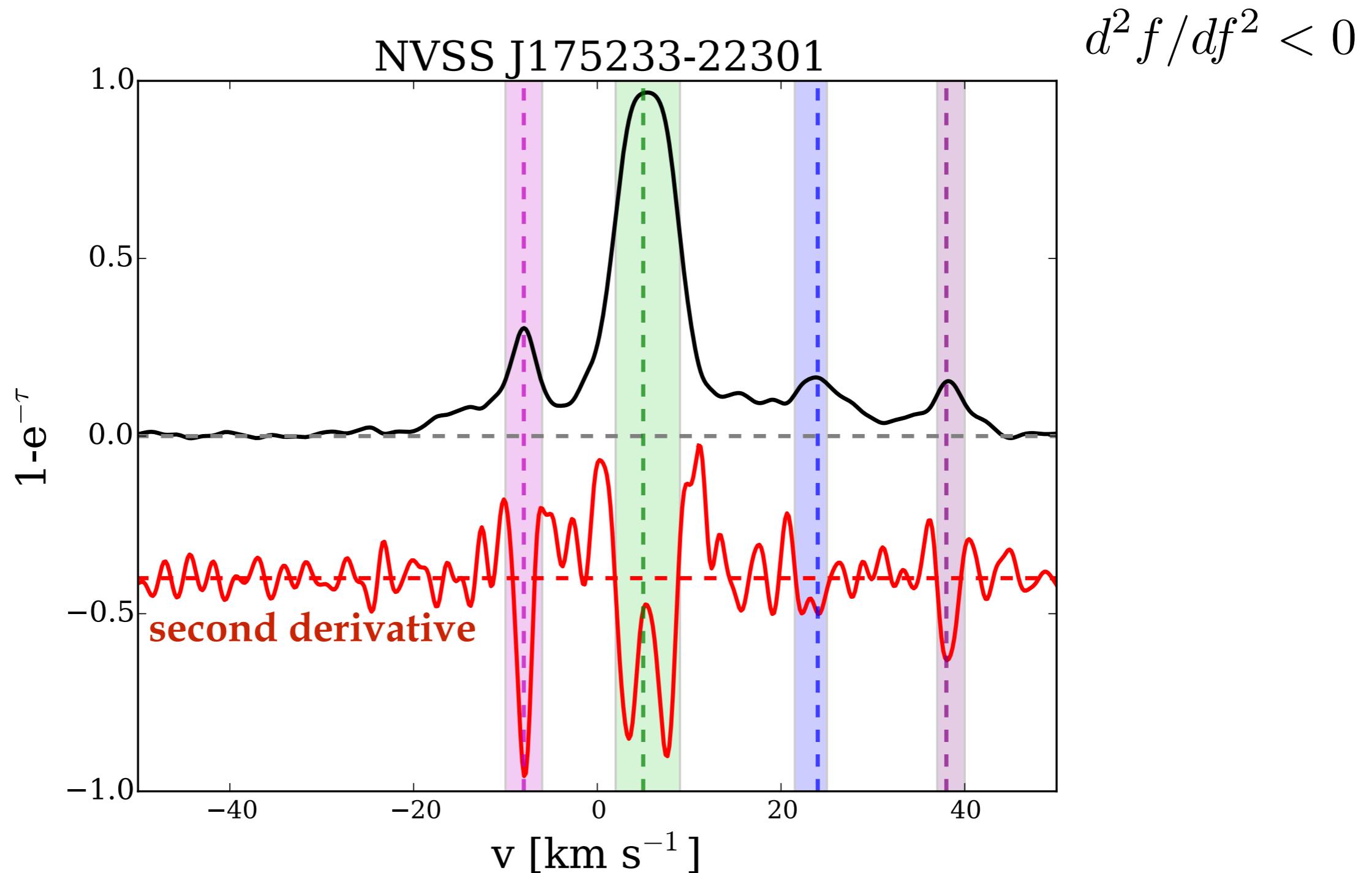
$$d^4 f / df^4 > 0$$

- ▶ Autonomous Gaussian decomposition (GaussPy, Lindner et al. 2015)
 - ▶ based on derivative spectroscopy and machine learning
 - ▶ no ‘human’ input needed
 - ▶ ideal for large data sets

GAUSSPY

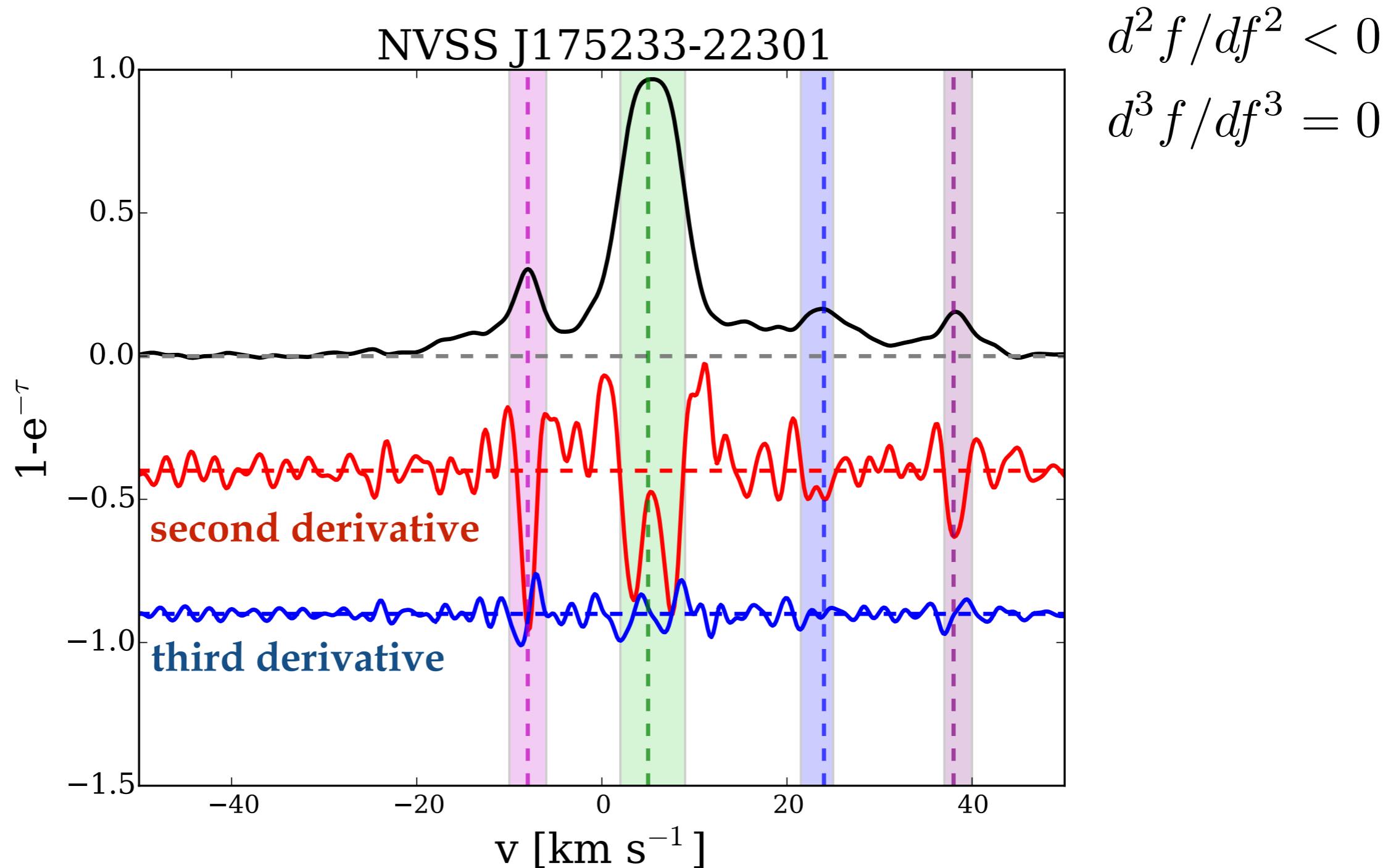
- ▶ GaussPy is available on Github: <https://github.com/gausspy/gausspy>
- ▶ can be used to decompose any spectra into Gaussian components
- ▶ New, optimised version: <https://github.com/mriener/gausspyplus>

GAUSSIAN DECOMPOSITION



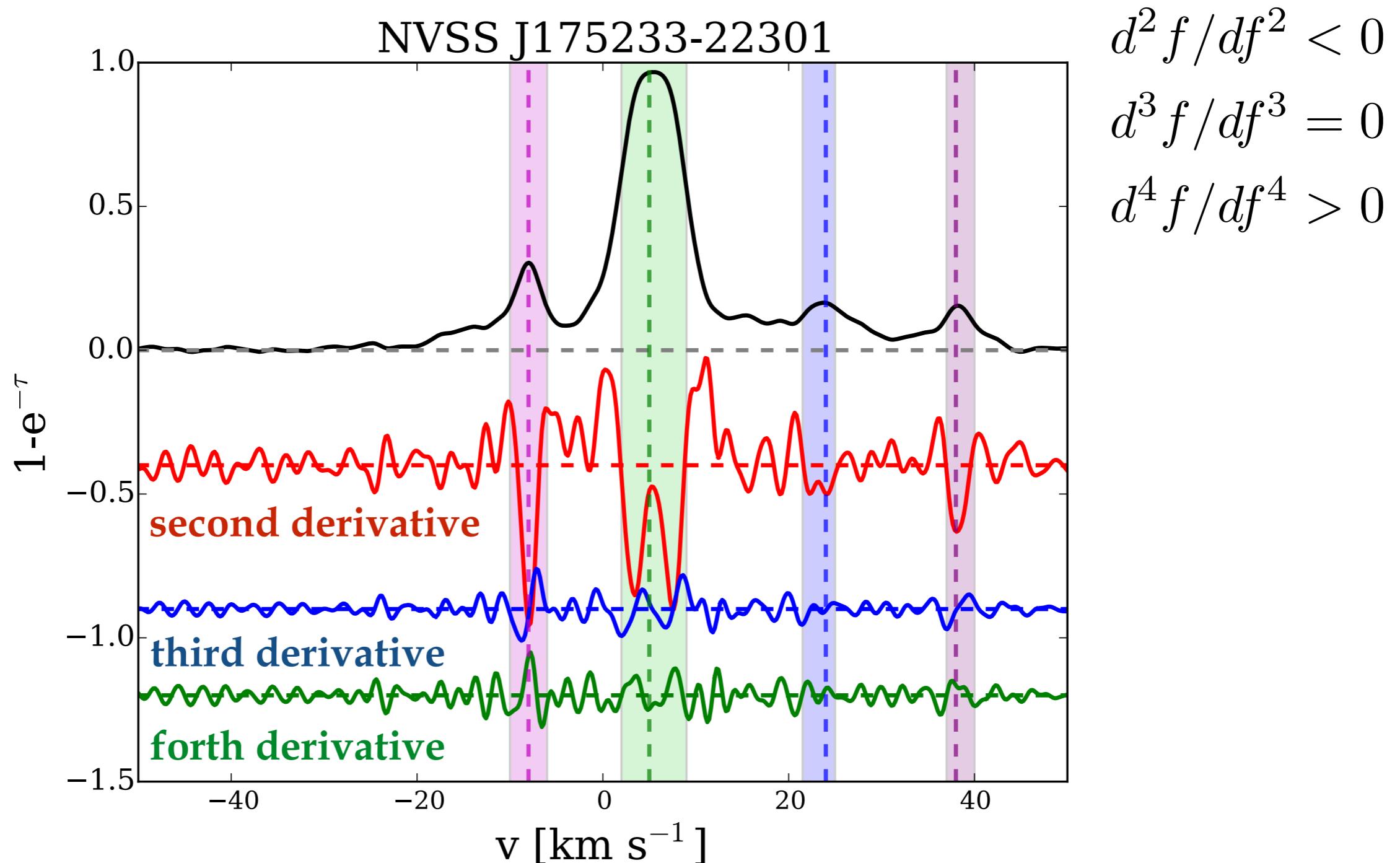
- the derivative is used to identify the positions of the Gaussian components

GAUSSIAN DECOMPOSITION



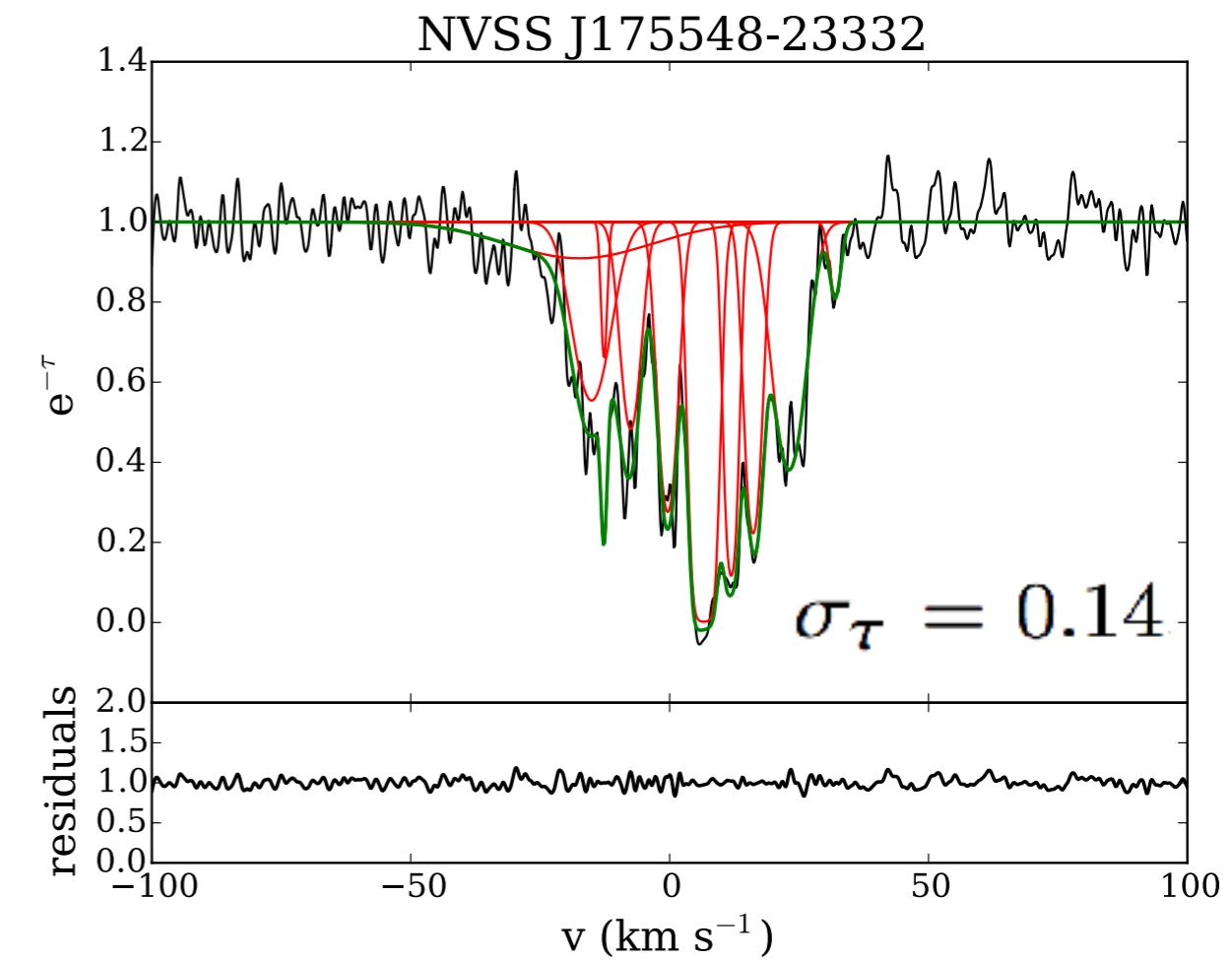
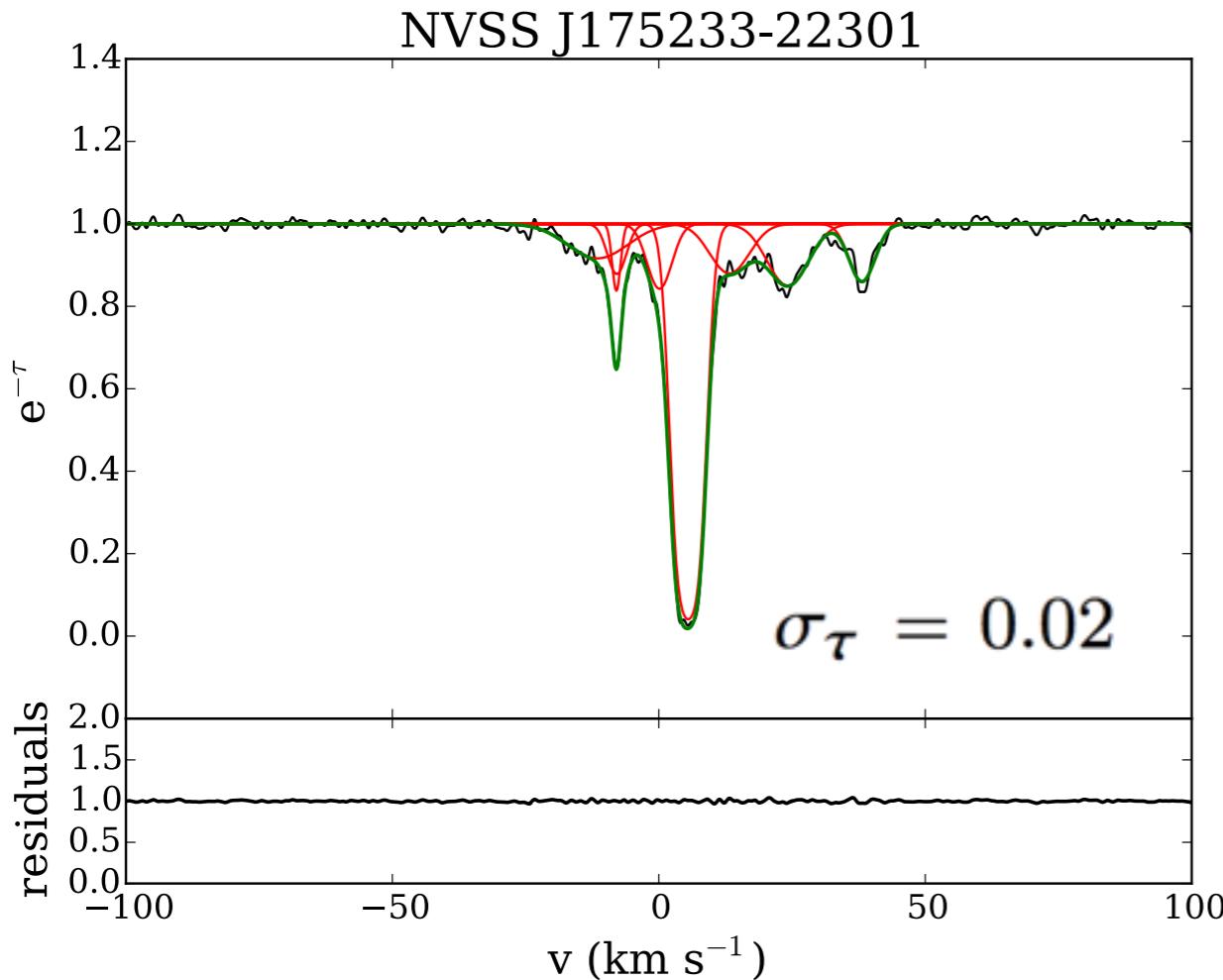
- the derivative is used to identify the positions of the Gaussian components

GAUSSIAN DECOMPOSITION



- the derivative is used to identify the positions of the Gaussian components

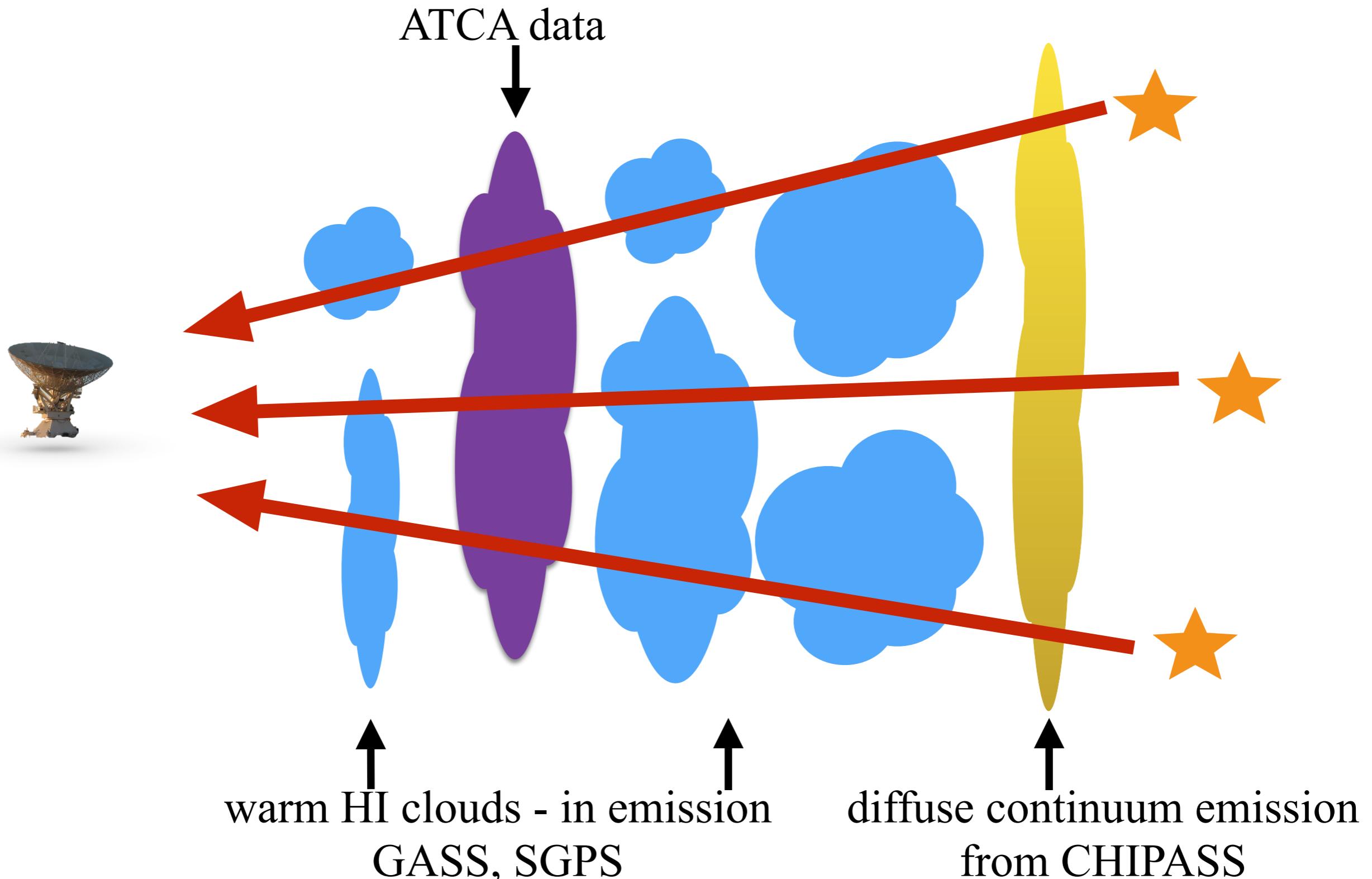
GAUSSIAN DECOMPOSITION



- ▶ strongly blended multiple component profiles
- ▶ automated Gaussian decomposition with **GaussPy** (Lindner et al. 2015)

4 COMPONENT MODEL

Riegel-Crutcher cloud
cold HI - in absorption



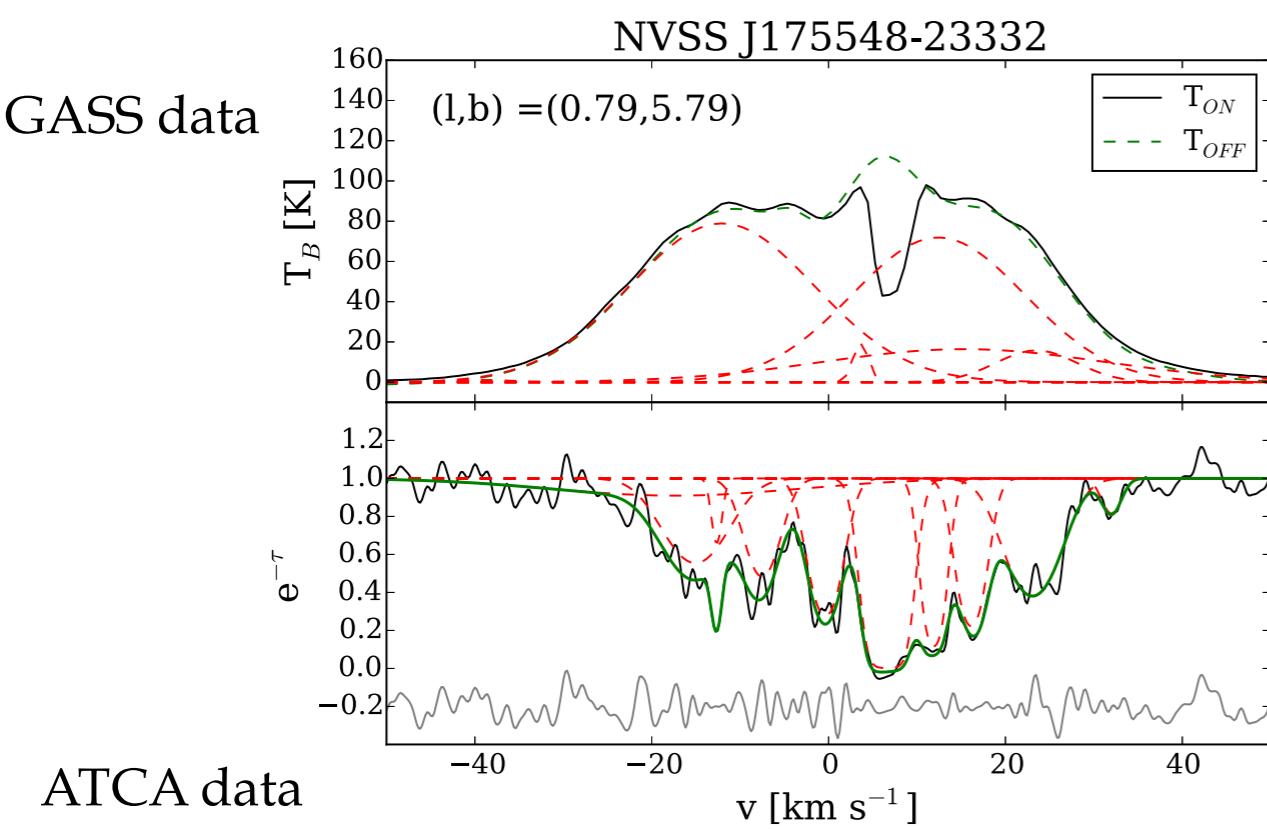
SPIN TEMPERATURES

4 component model:

$$T_B = T_{s,fg}(1 - e^{-\tau_{fg}}) + T_{s,HISA}(1 - e^{-\tau_{HISA}})e^{-\tau_{fg}} \\ + T_{s,bg}(1 - e^{-\tau_{bg}})e^{-(\tau_{fg} + \tau_{HISA})} + T_c e^{-(\tau_{fg} + \tau_{HISA} + \tau_{bg})}$$

$$T_{ON} - T_{OFF} = (T_{s,HISA} - T_c - pT_{OFF})(1 - e^{-\tau_{HISA}})$$

$$T_s = \frac{T_{ON} - T_{OFF}}{1 - e^{-\tau}} + T_c + pT_{OFF}$$

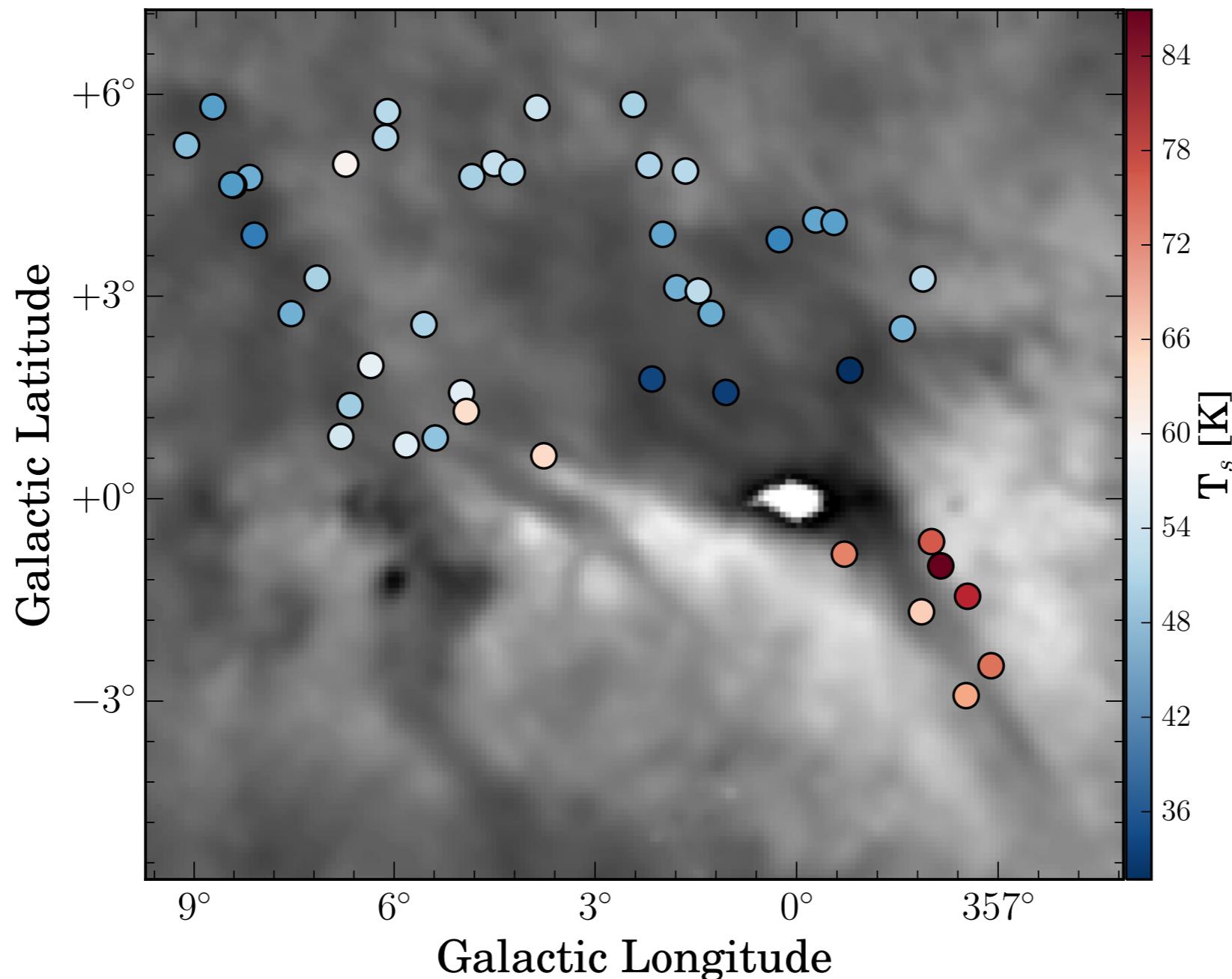


→ T_{ON}

→ T_{OFF}

→ $e^{-\tau}$

SPIN TEMPERATURES



- ▶ T_s median 48 K
- ▶ Temperature gradient across the cloud
- ▶ Molecular cloud formation may start



THANK YOU!