

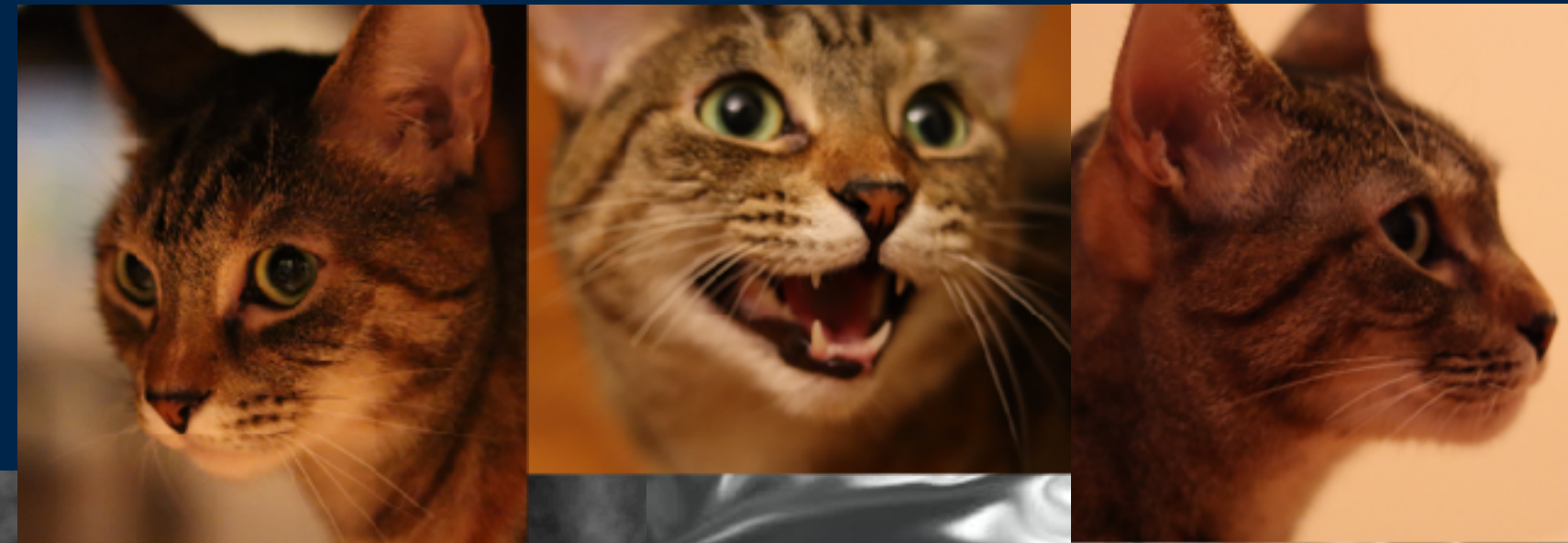
The scattering transform: a new vocabulary for textures

Sihao Cheng (程思浩)
Johns Hopkins University

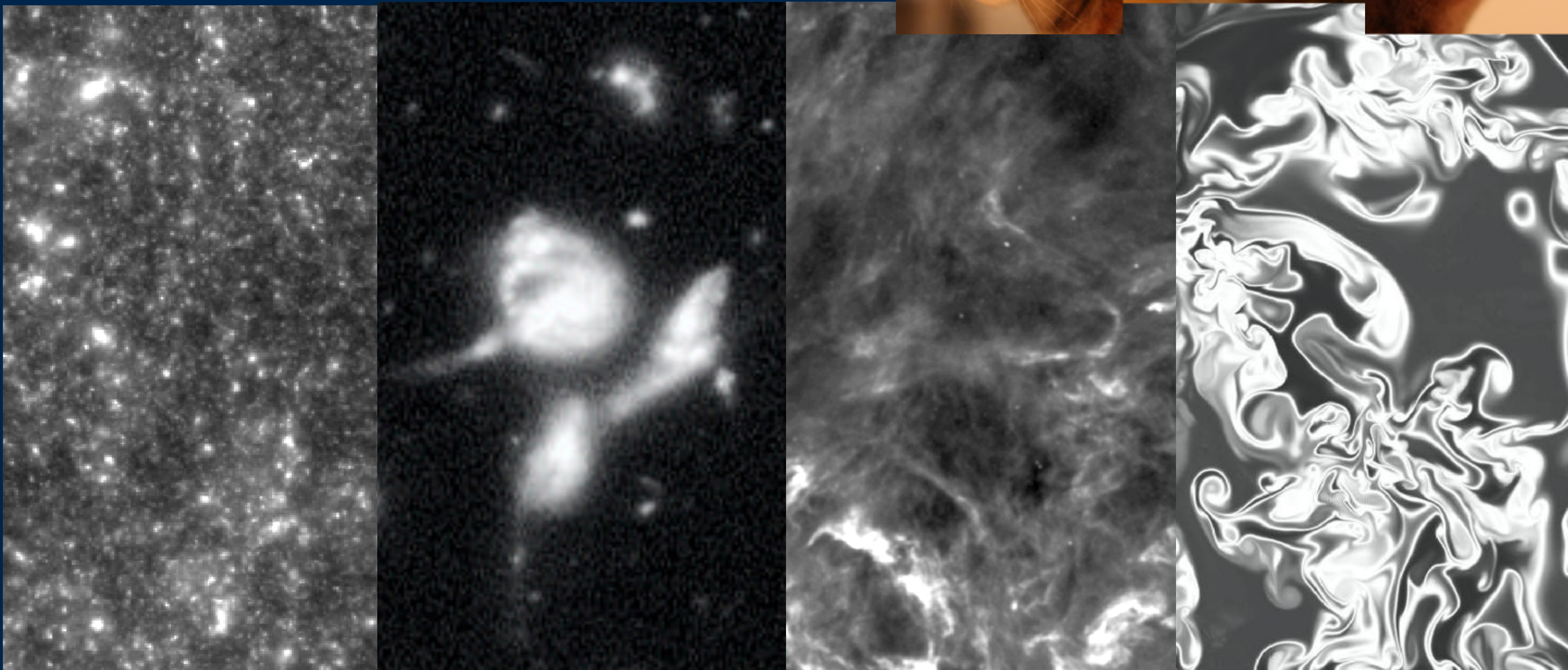
TsingHua University
Mar 16th, 2021

with Brice Ménard, Yuan-Sen Ting, & Joan Bruna

How do we extract information from a **field**?



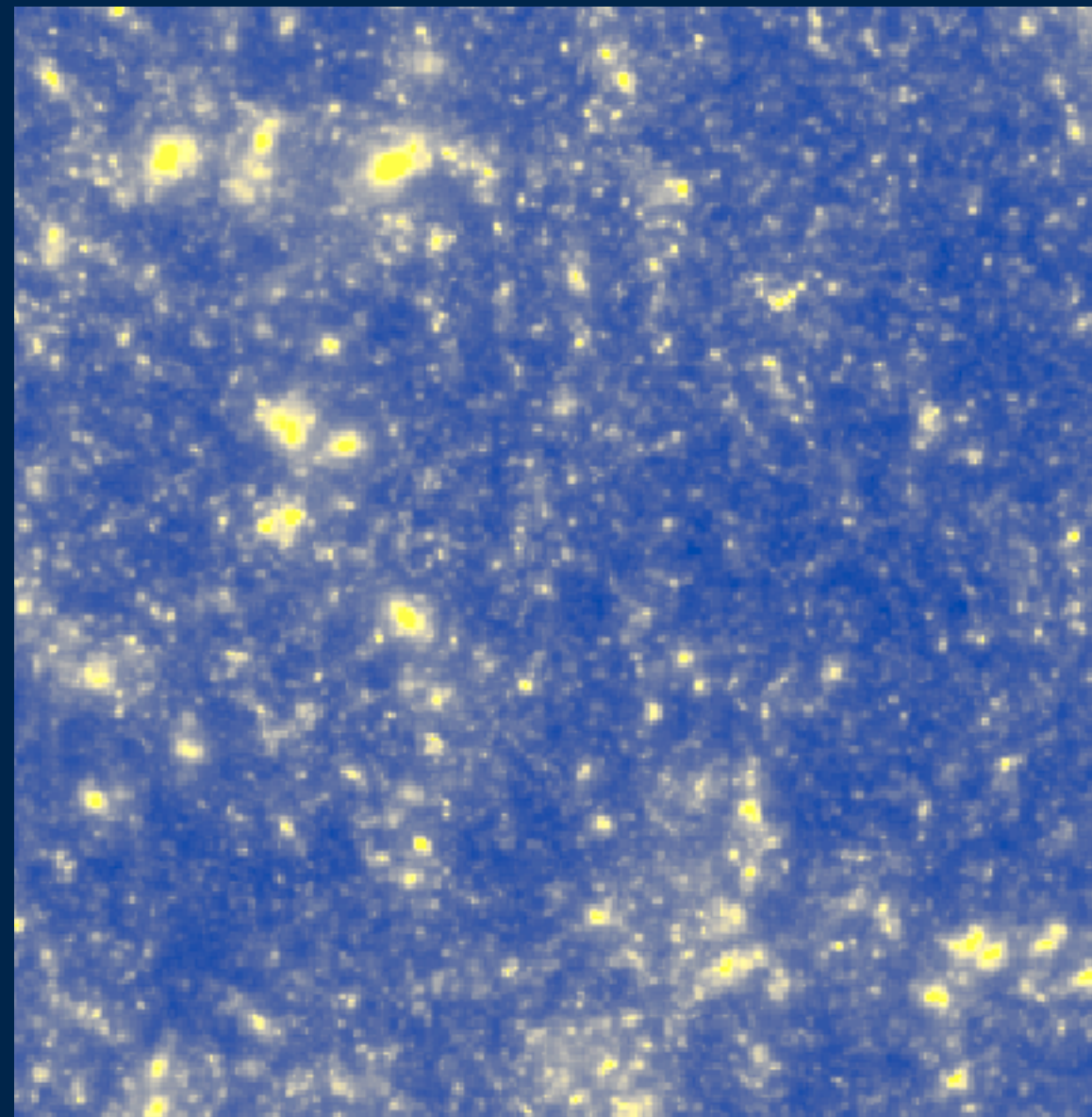
daily-life
information



physical
parameters

1. compression
2. better metric

How do we **characterize** a field?



data



a description
using a vocabulary

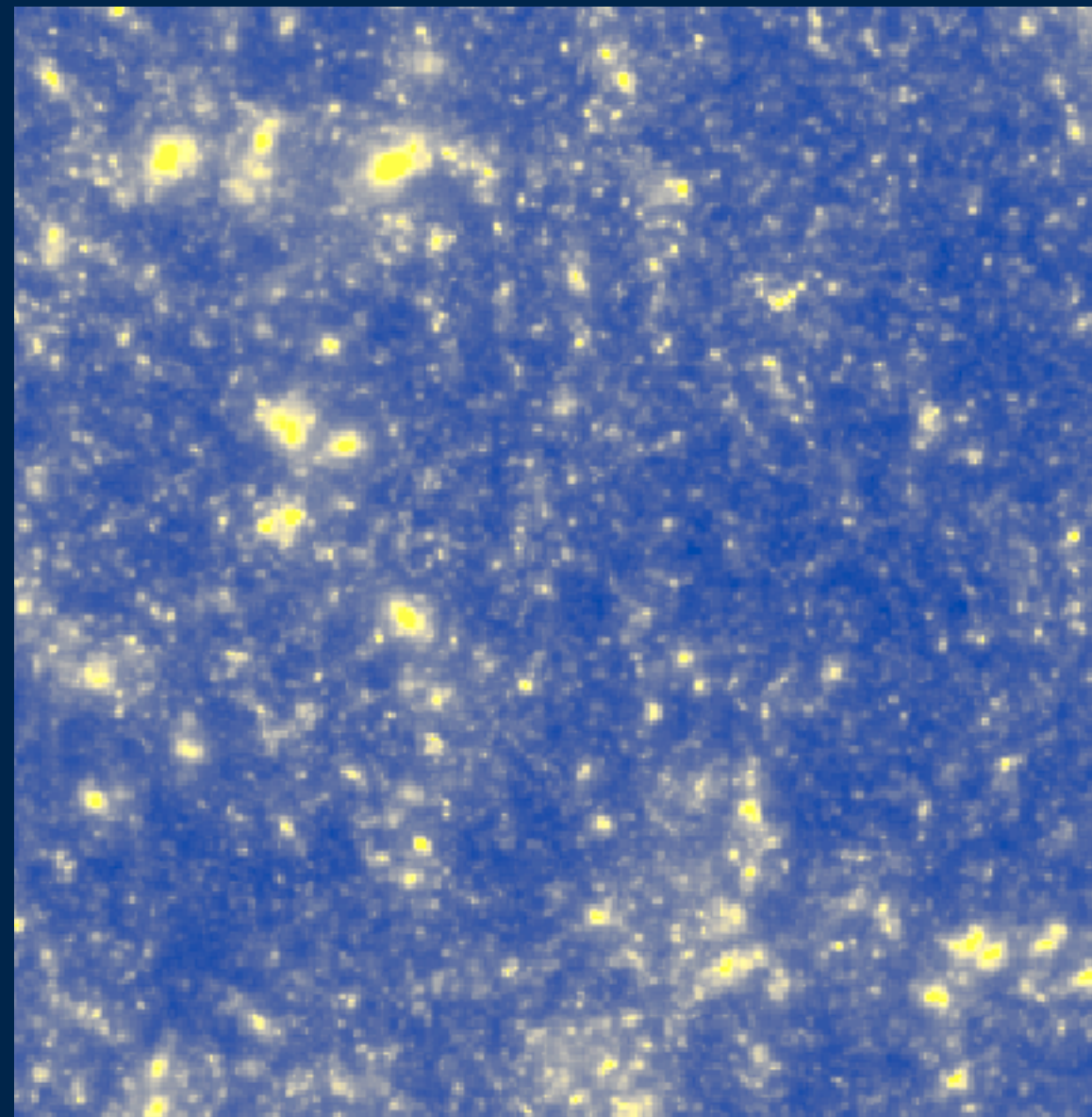
extract patterns



physical
parameters

information

How do we characterize a field?



data



power spectrum

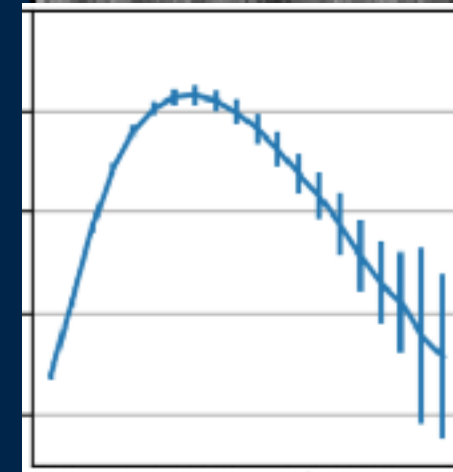
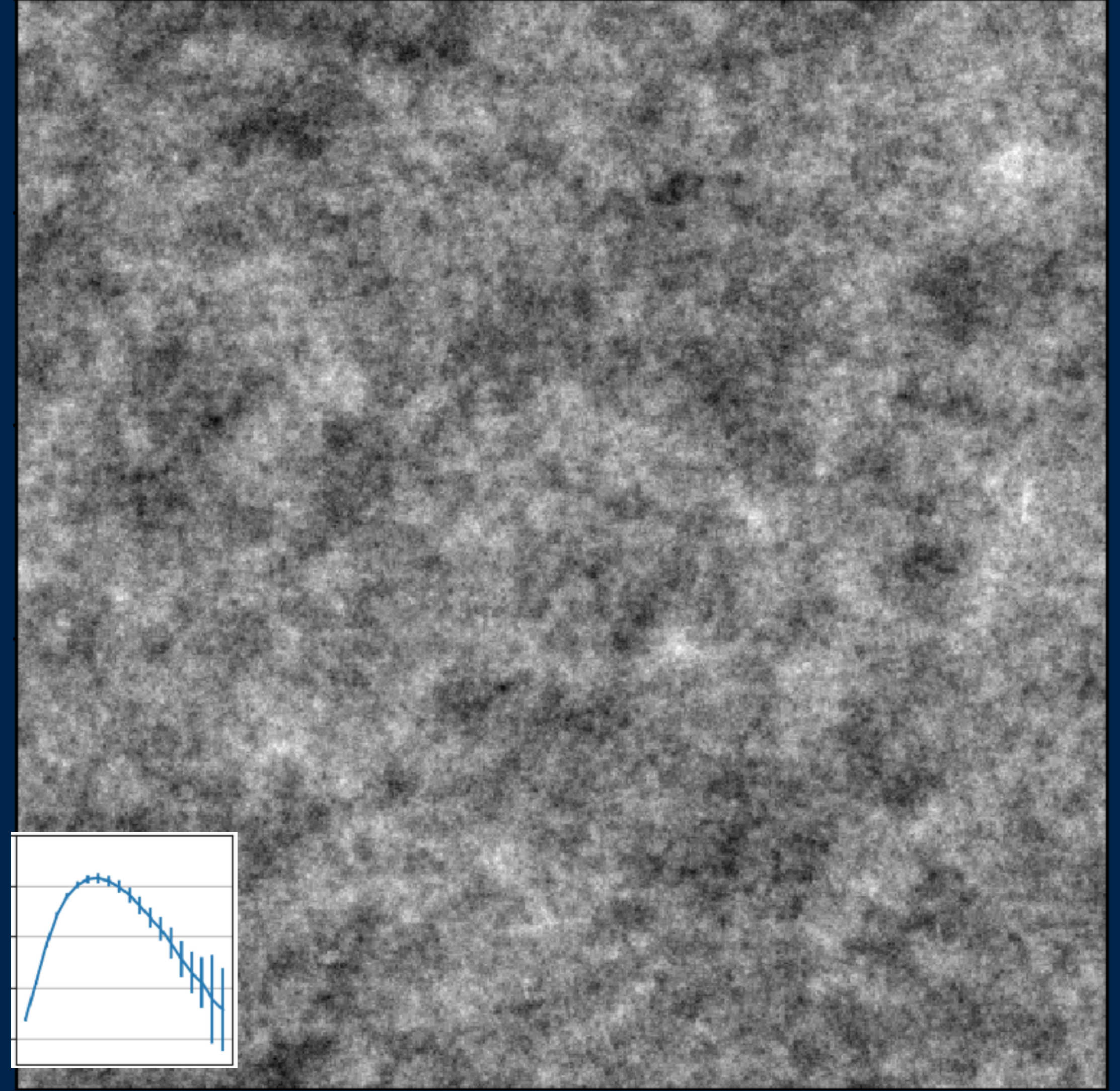
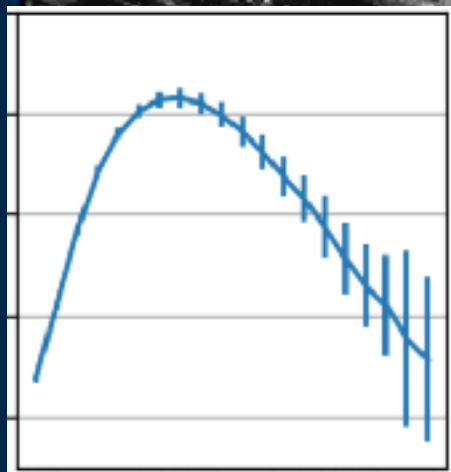
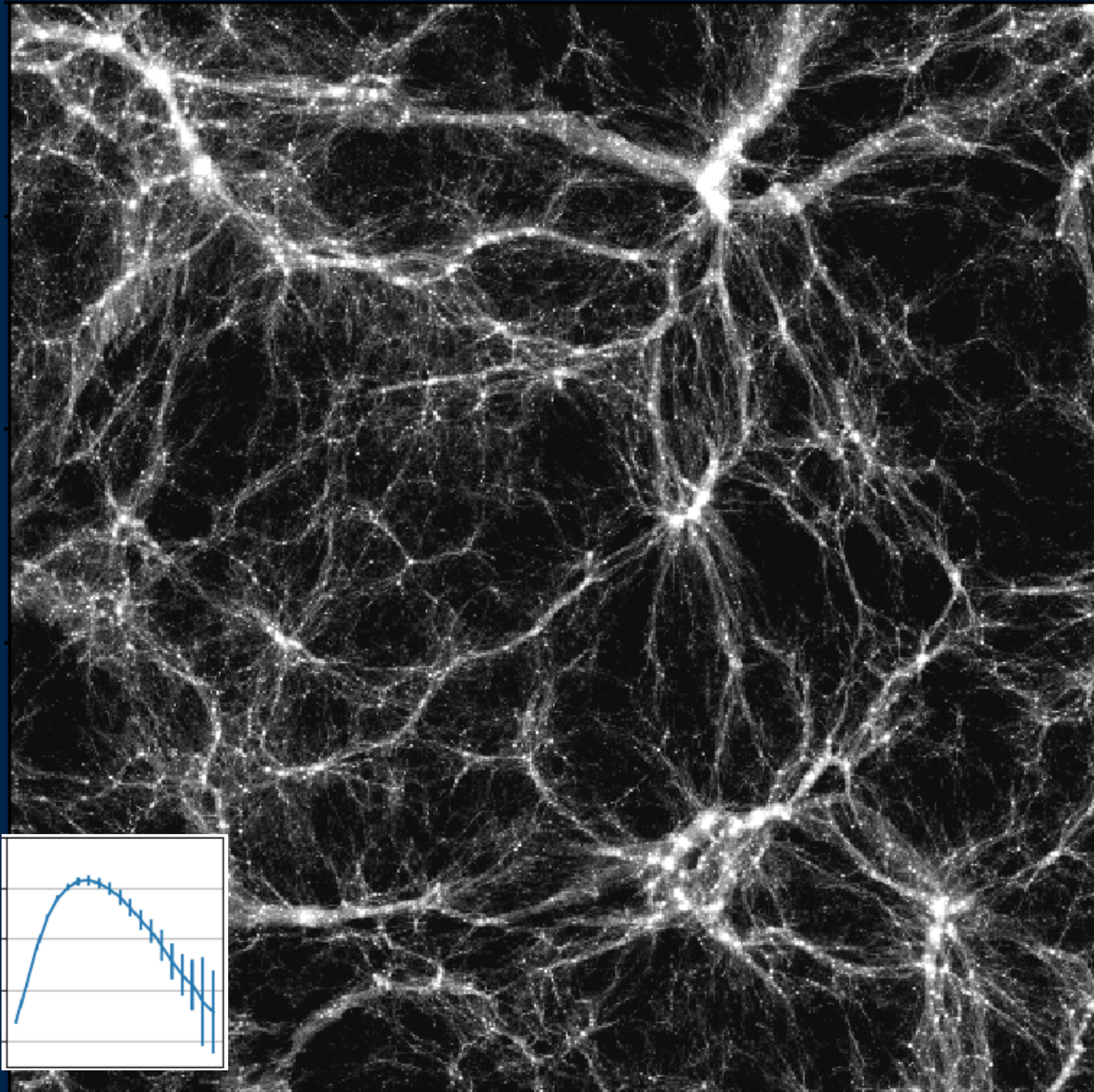
$$P(k) = \langle \delta_k \delta_{-k} \rangle$$

extract patterns



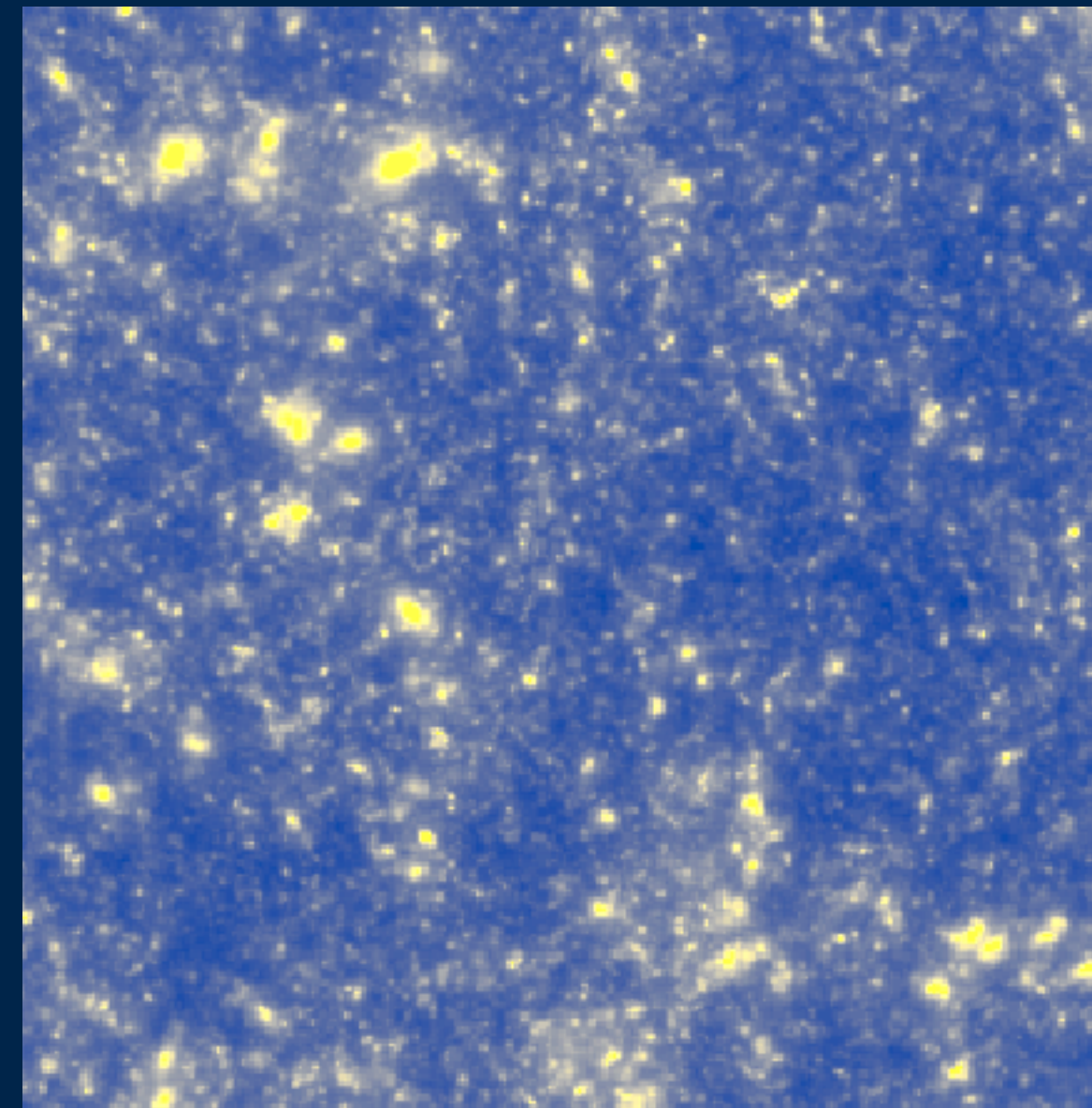
physical
parameters

information



We need to go **beyond** $P(k)$ to capture non-Gaussianity.

How do we characterize a field?



$P(k)$ plus
classic statistics

$$\langle \delta_1 \delta_2 \dots \delta_n \rangle$$

CNN

physical
parameters

How do we characterize a field?

1. lose information (Carron 2011)
2. too many configurations



$P(k)$ plus
classic statistics

$$\langle \delta_1 \delta_2 \dots \delta_n \rangle$$

CNN

physical
parameters

How do we characterize a field?



$P(k)$ plus
classic statistics
 $\langle \delta_1 \delta_2 \dots \delta_n \rangle$

physical
parameters

CNN

convolutional neural network

CNN learns & extracts patterns.



Conv. layer



Conv. layer

...



fully connected



my cat : 99%
your cat : 1%

data

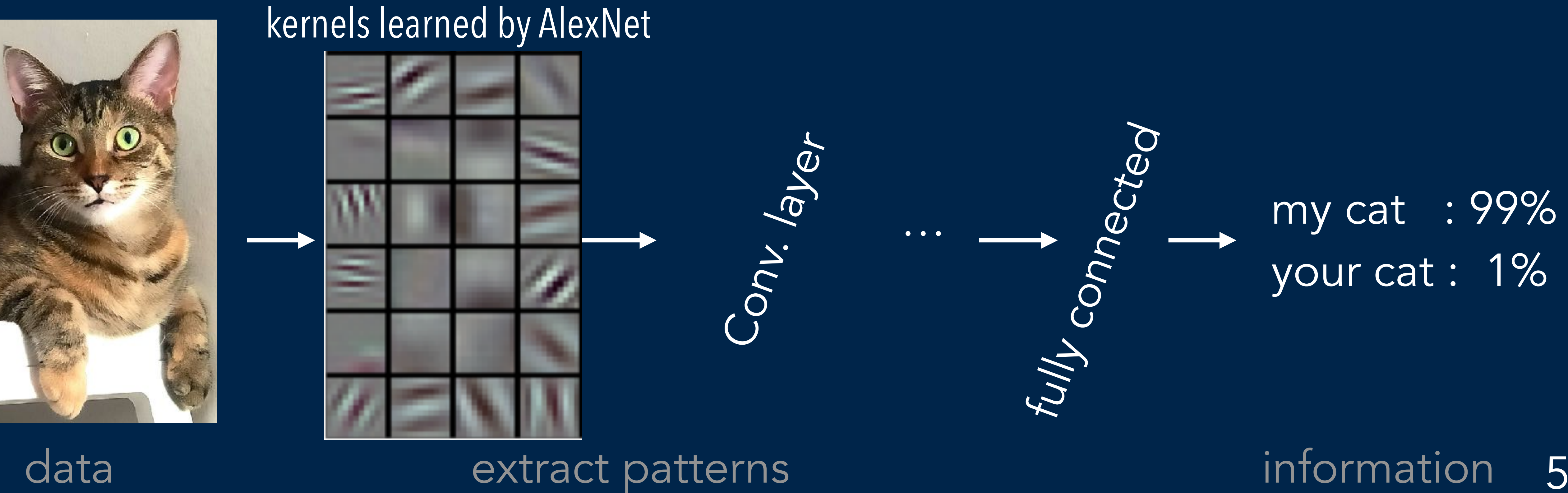
extract patterns

information 5

convolutional neural network

CNN learns & extracts patterns.

- 1. convolution
- 2. nonlinearity
- 3. average
- 4. iteration



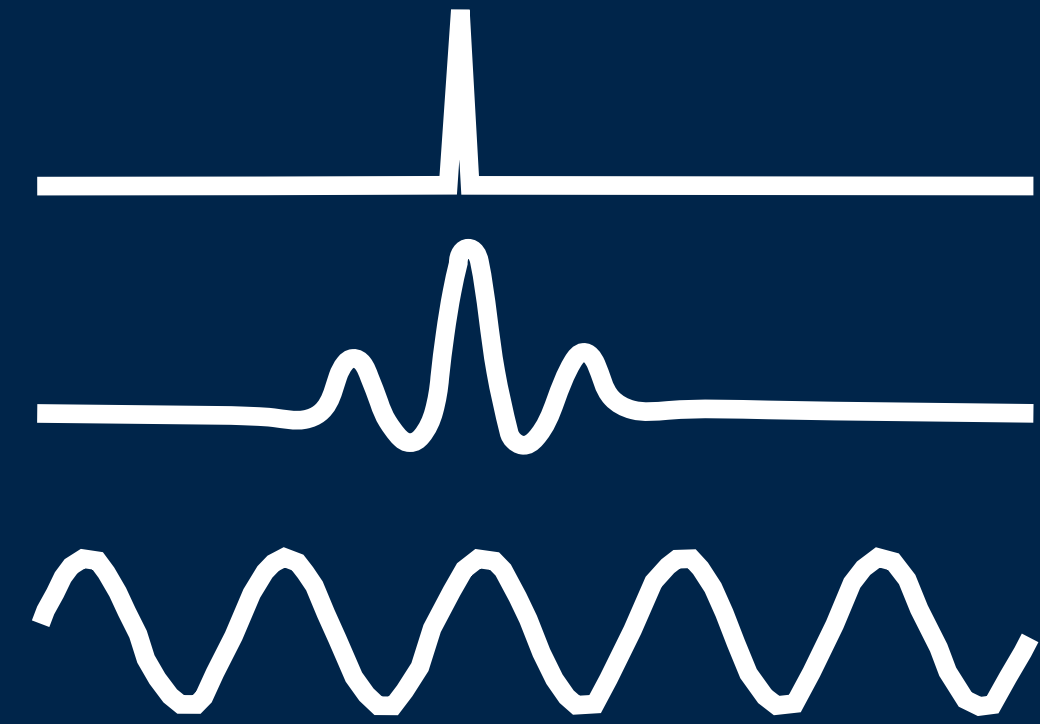
What are wavelets?

band-pass filters

delta function

wavelet

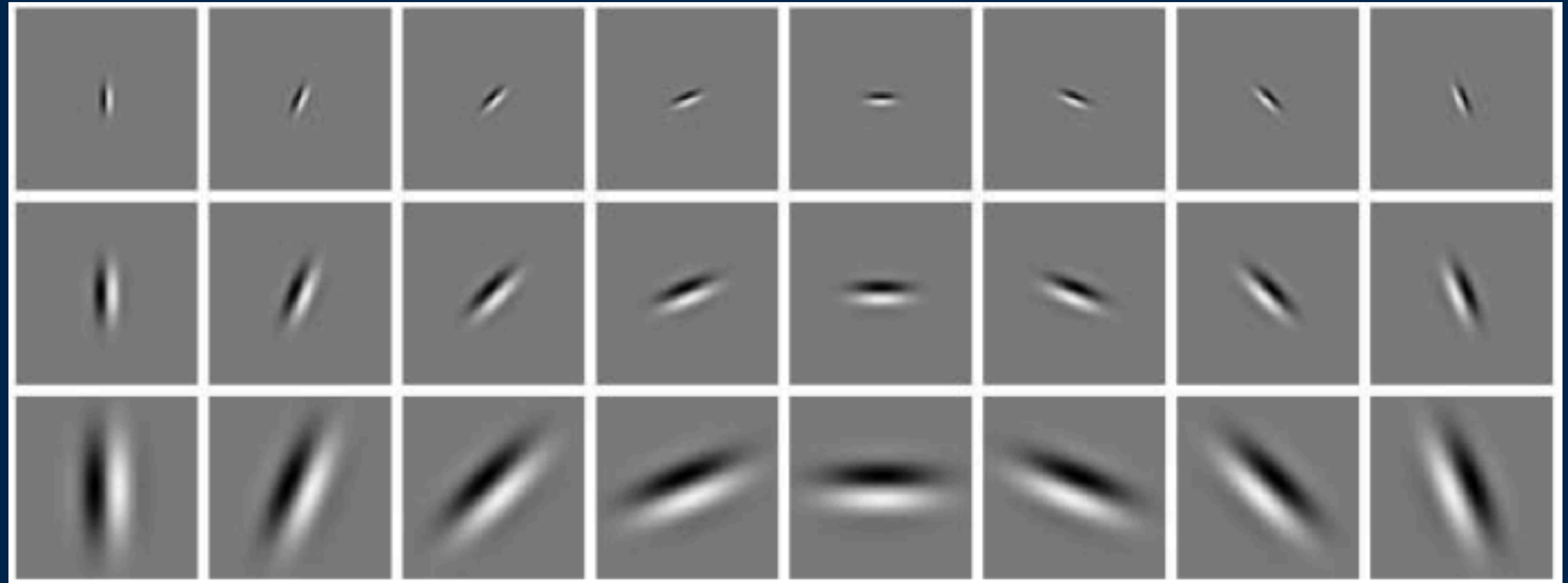
sine function



$\psi^{j,l}$

j : scales

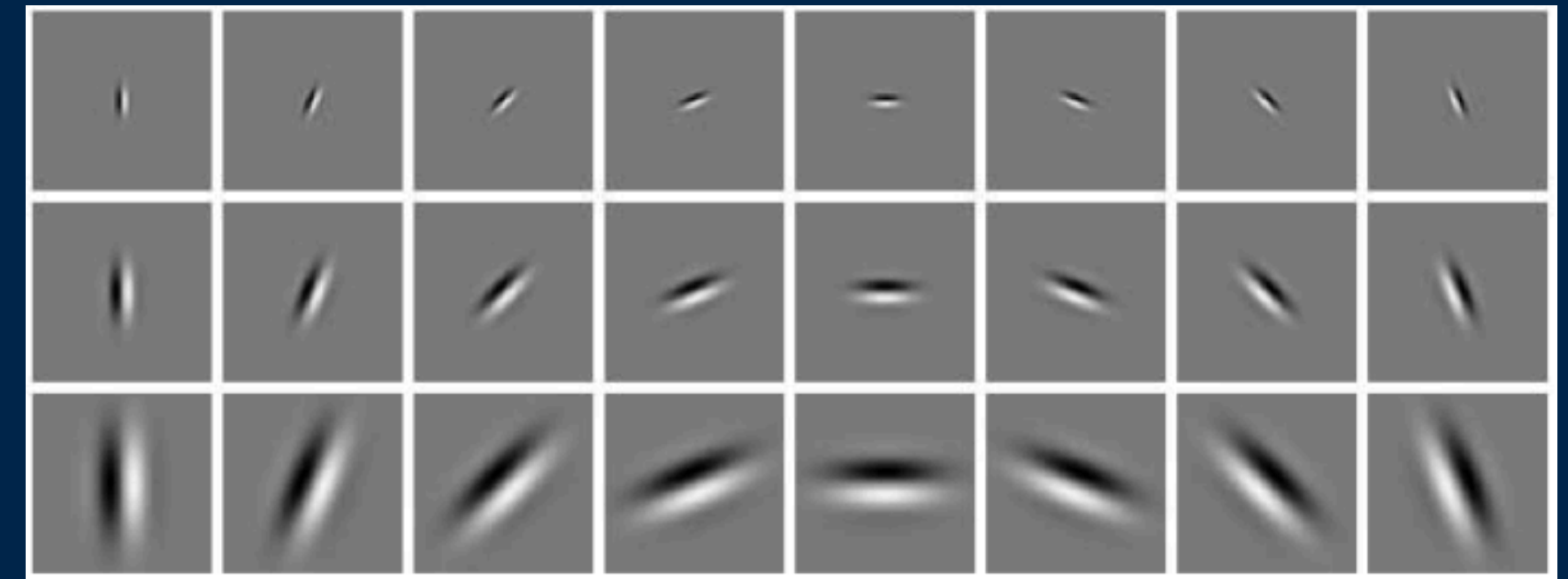
l : directions



receptive fields of mammal vision

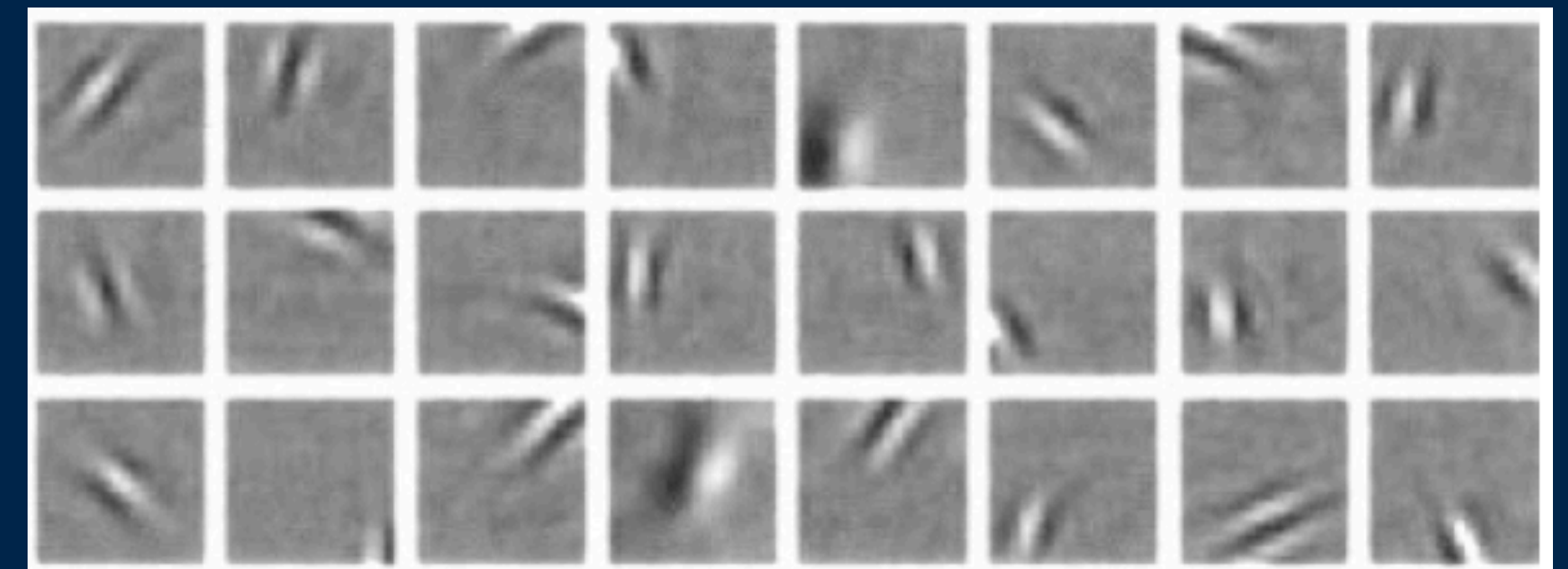
(Hubel & Wiesel 1968)

close to Gabor wavelets



sparse representation of natural images

(Olshausen & Field 1996)



kernels learned in AlexNet

(Krizhevsky, Sutskever, & Hinton 2012)

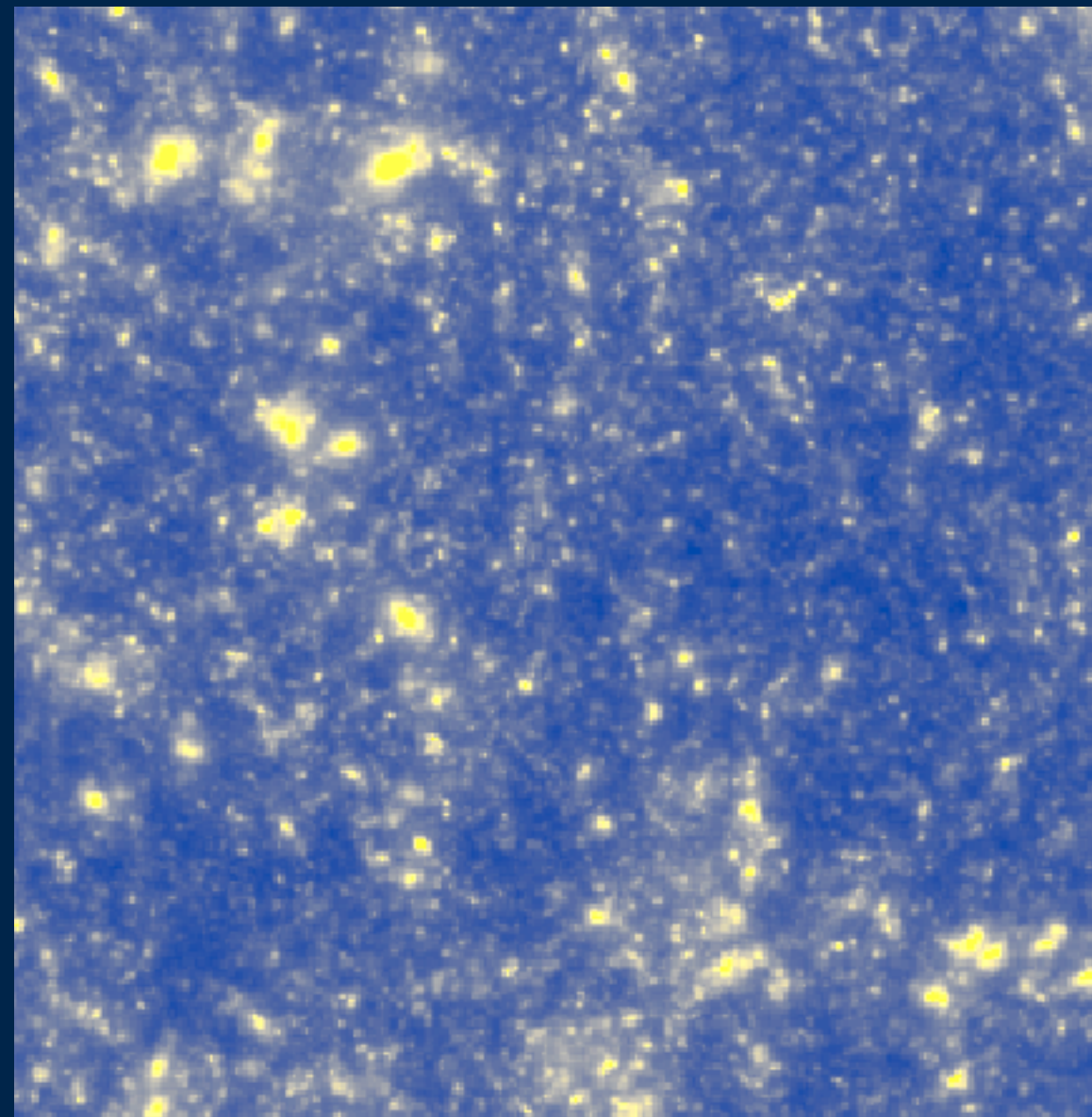


How do we characterize a field?

limited information

but deterministic, controllable

$P(k)$ plus
classic statistics

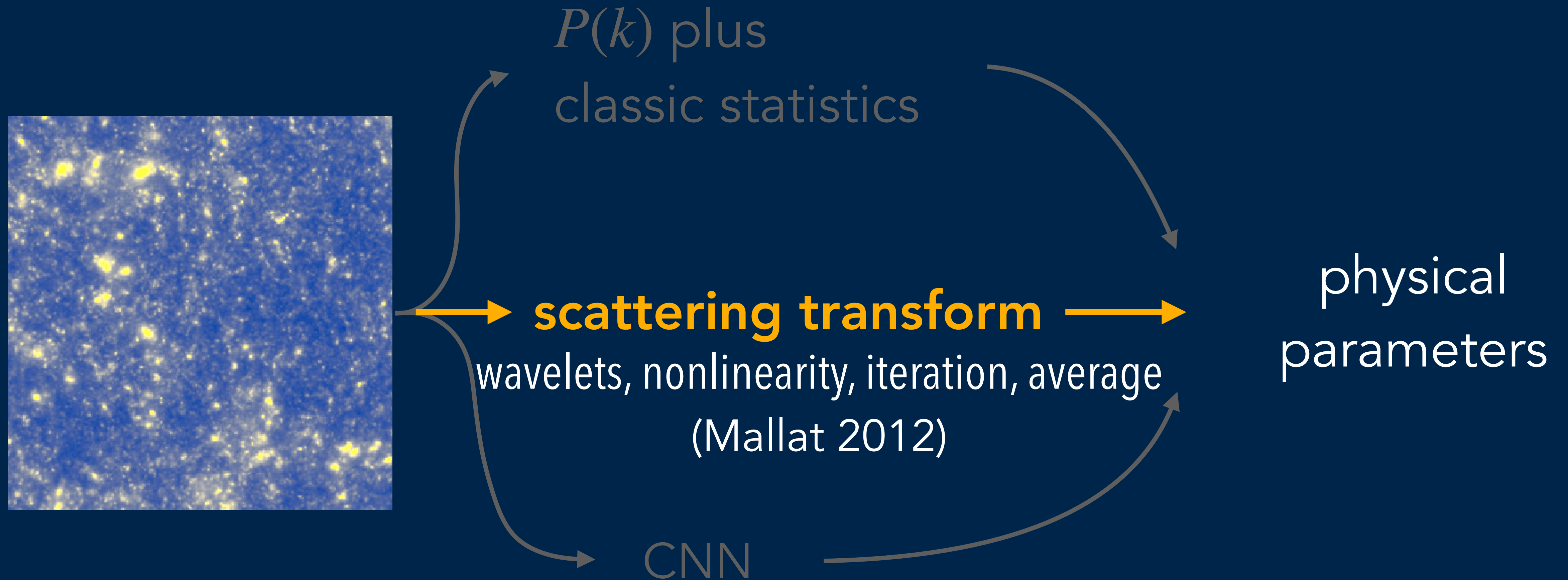


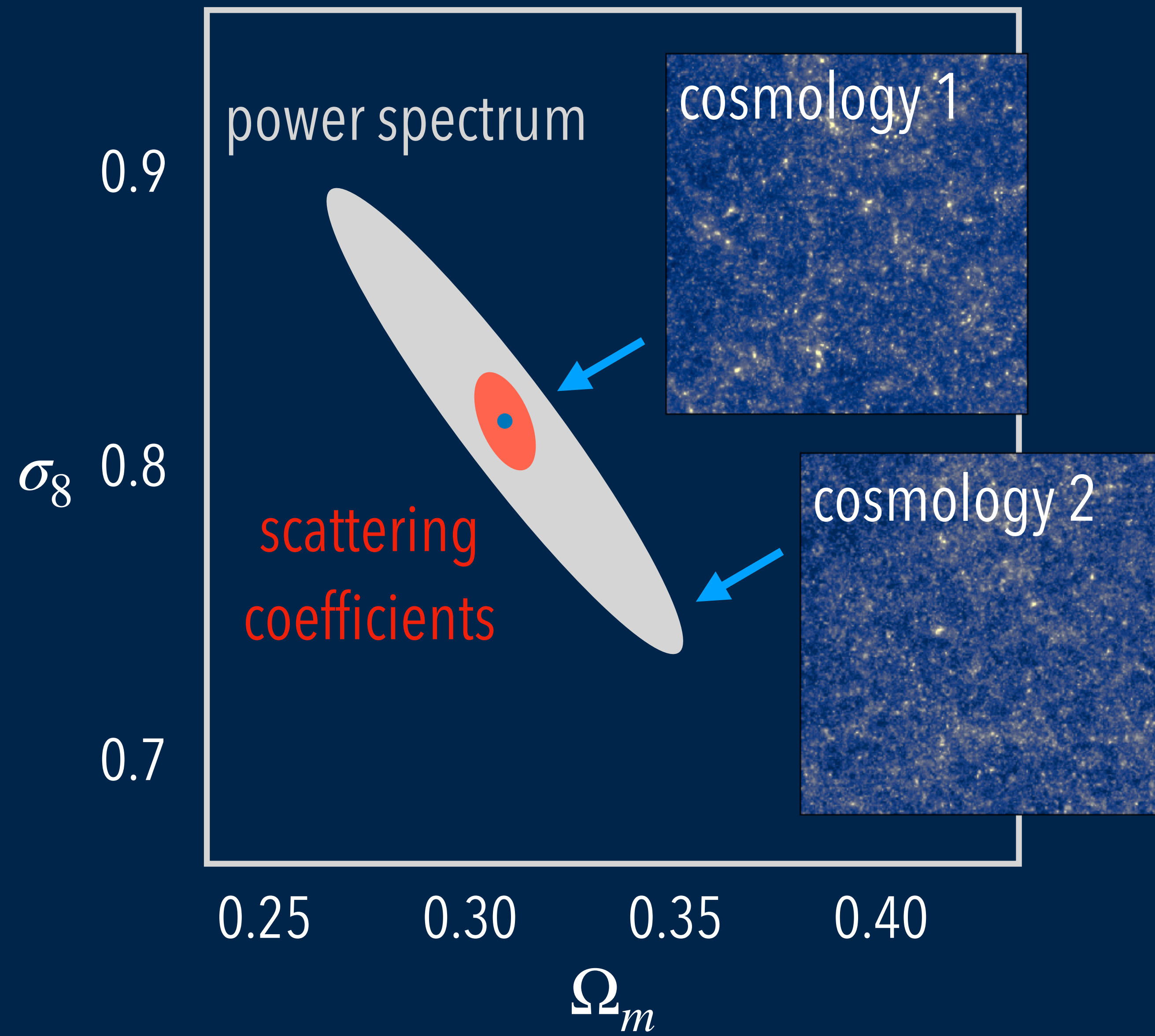
physical
parameters

CNN

wavelet, nonlinearity, iteration
but need learning

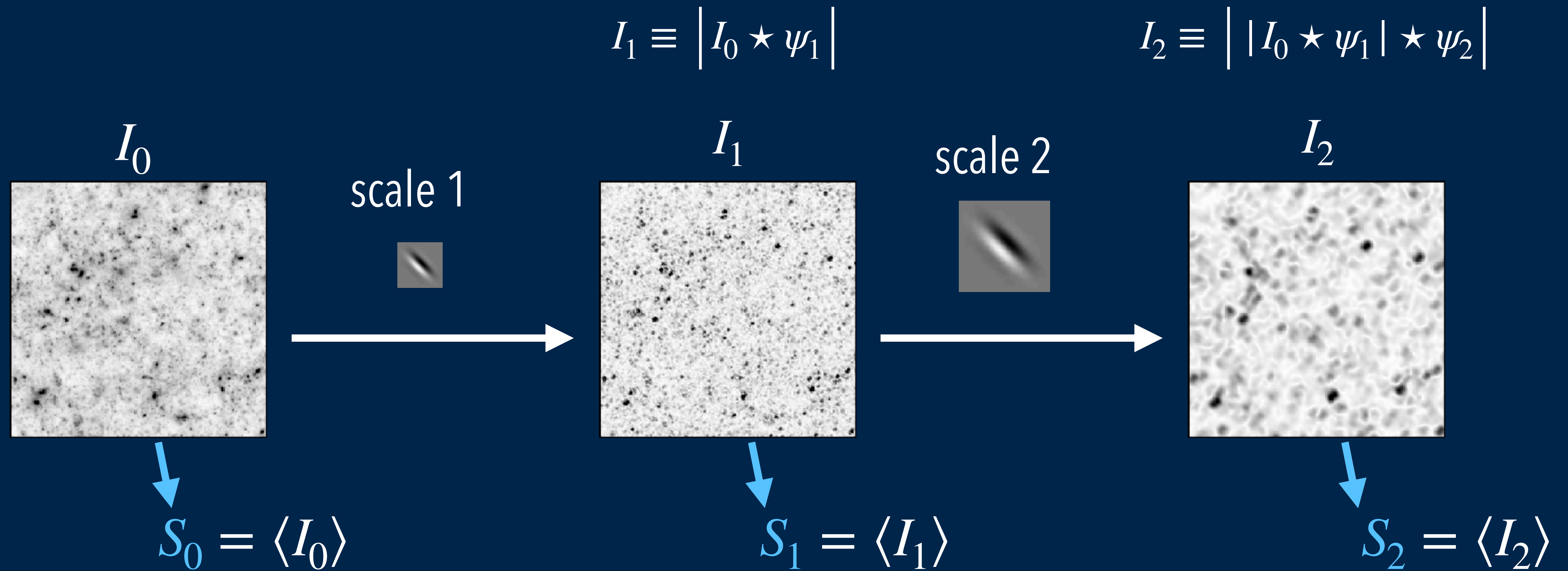
How do we characterize a field?





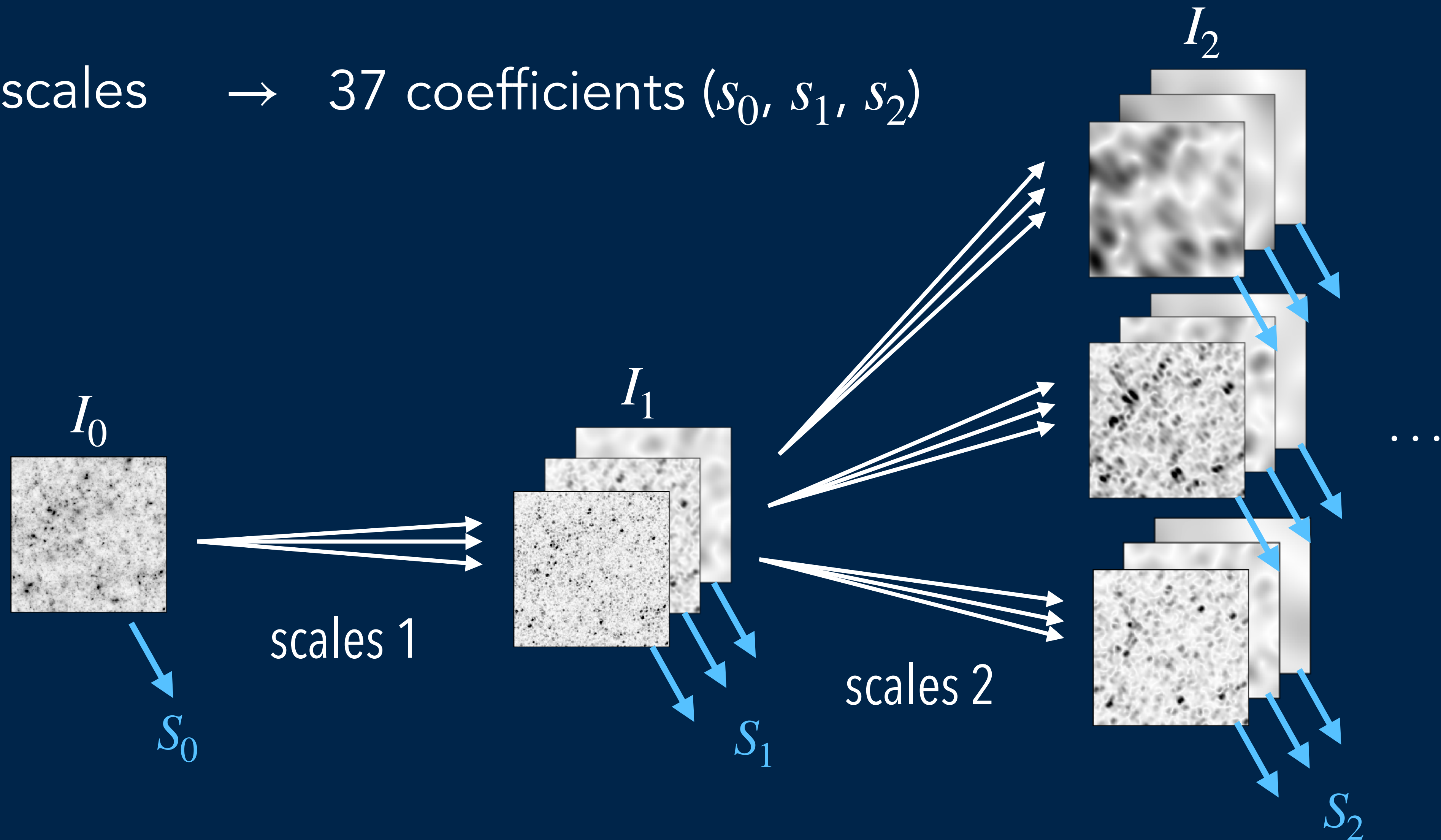
scattering transform

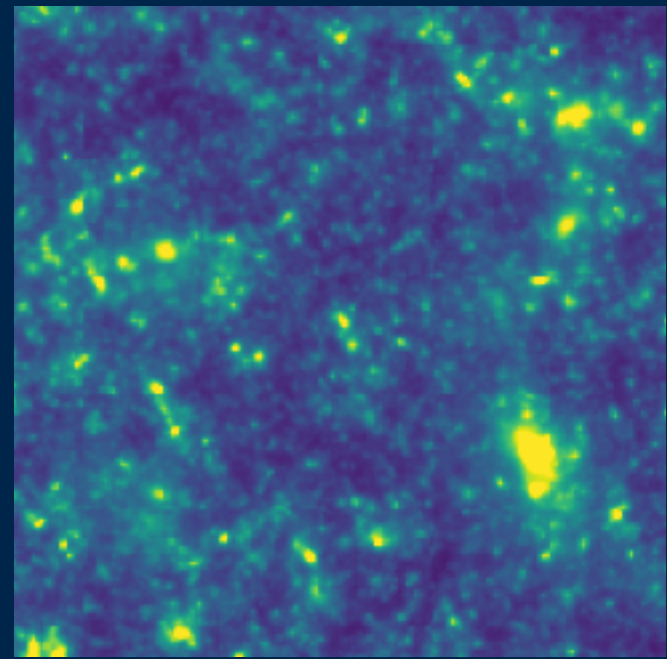
wavelets, modulus, iteration, average (Mallat 2012)



number of coefficients

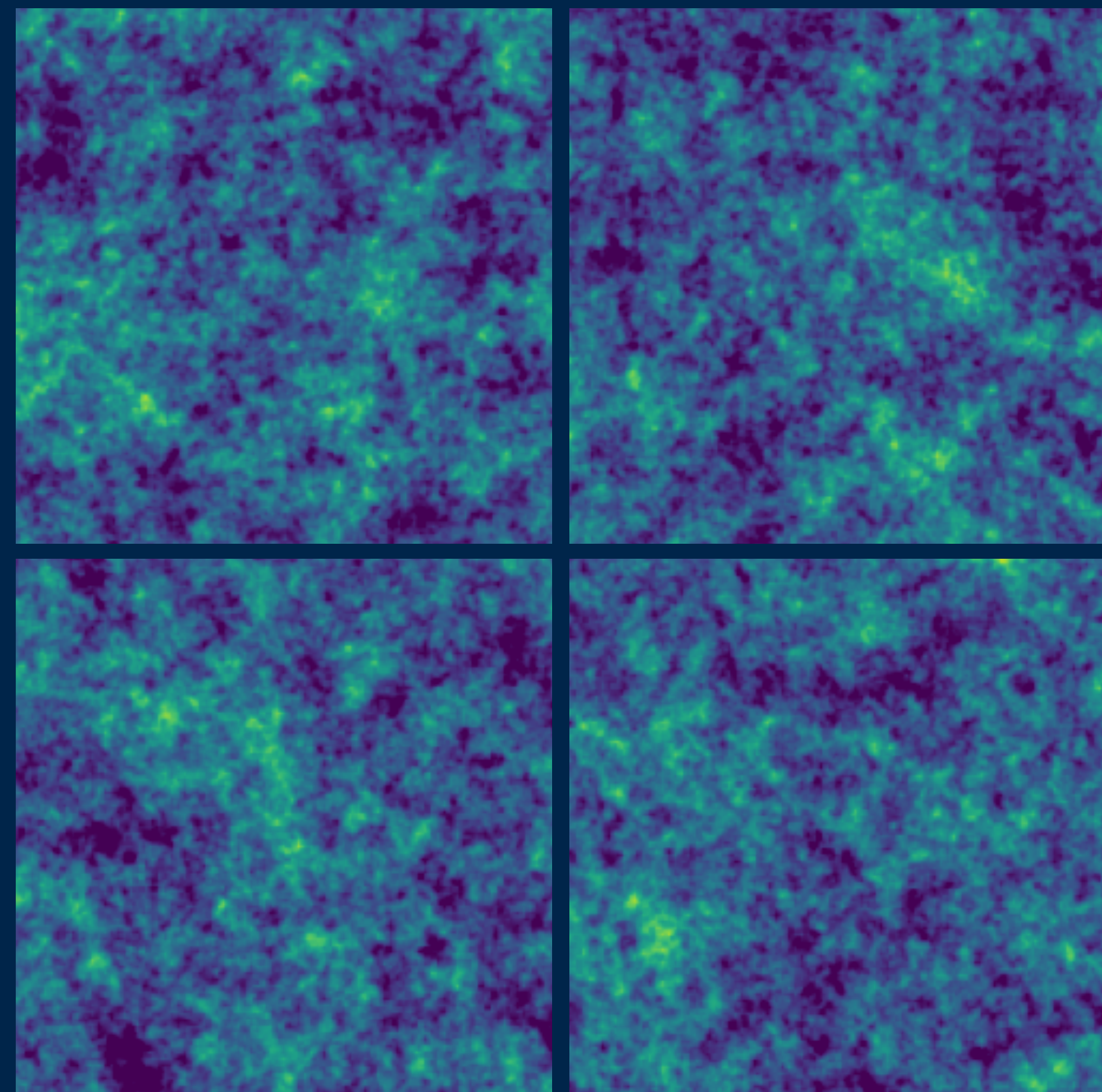
e.g., 8 scales $\rightarrow$ 37 coefficients (s_0, s_1, s_2)



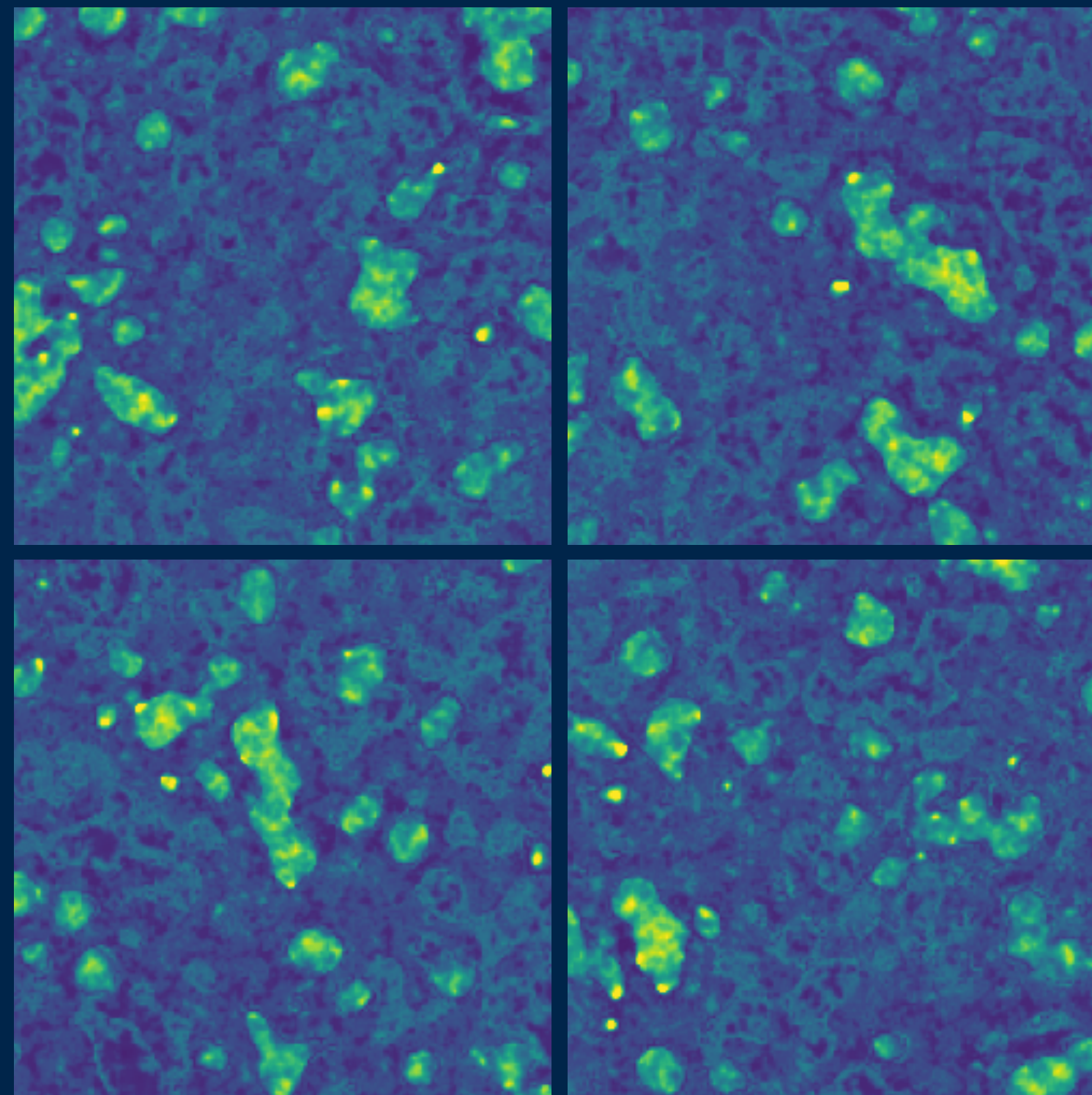


lensing map

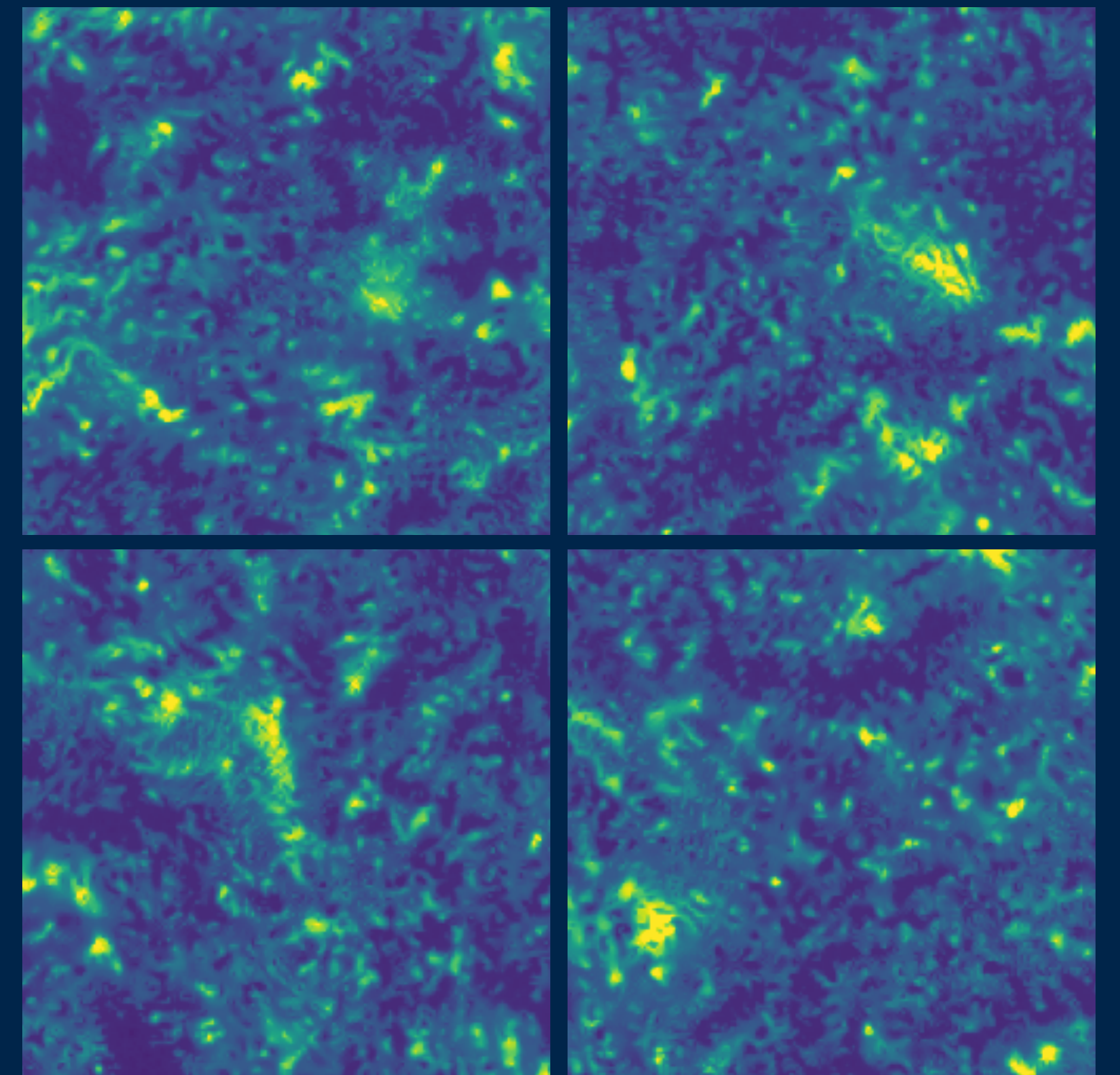
with $P(l)$



with $B(l_1, l_2, l_3)$ and $P(l)$

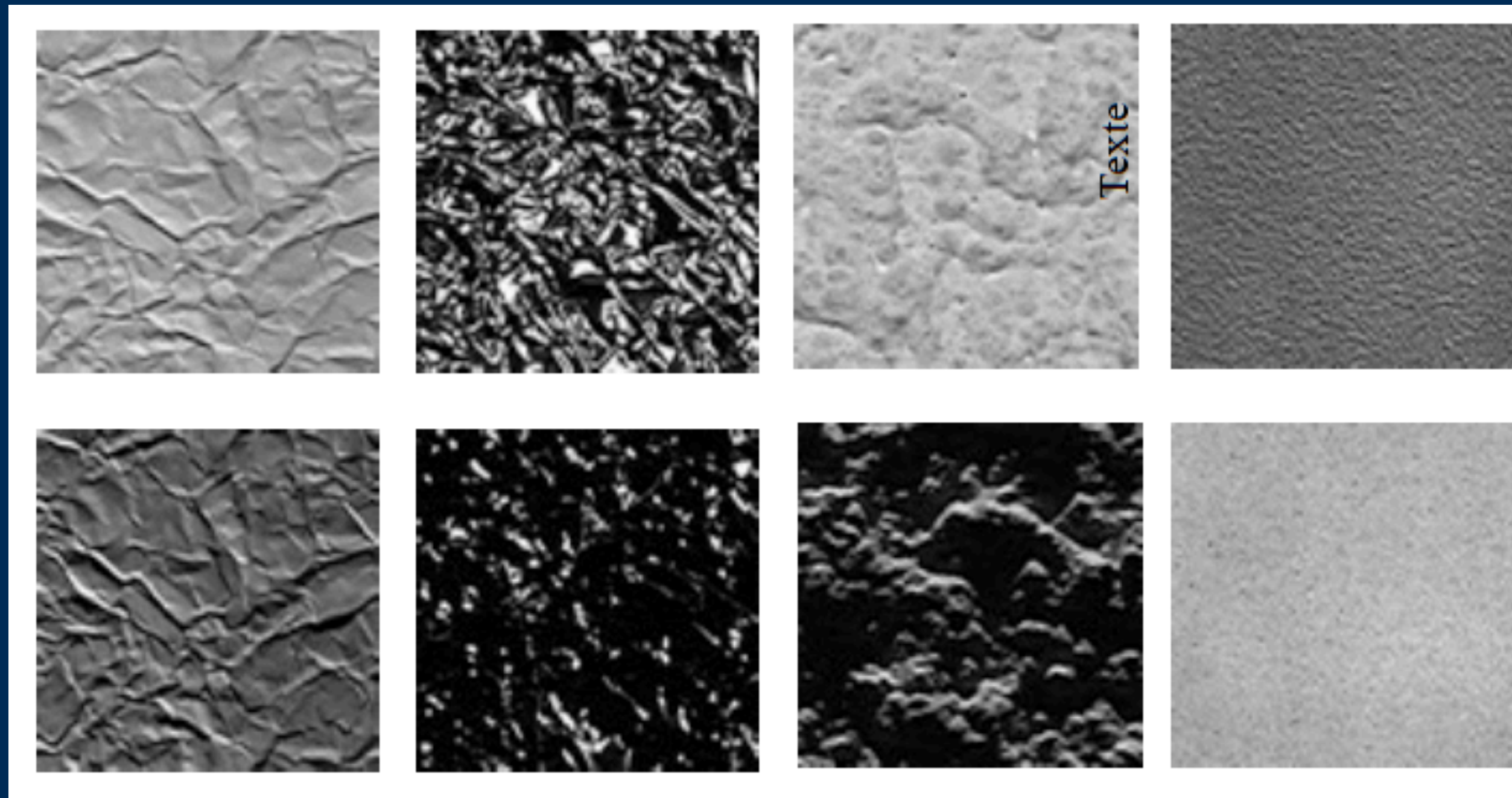


with S_1, S_2



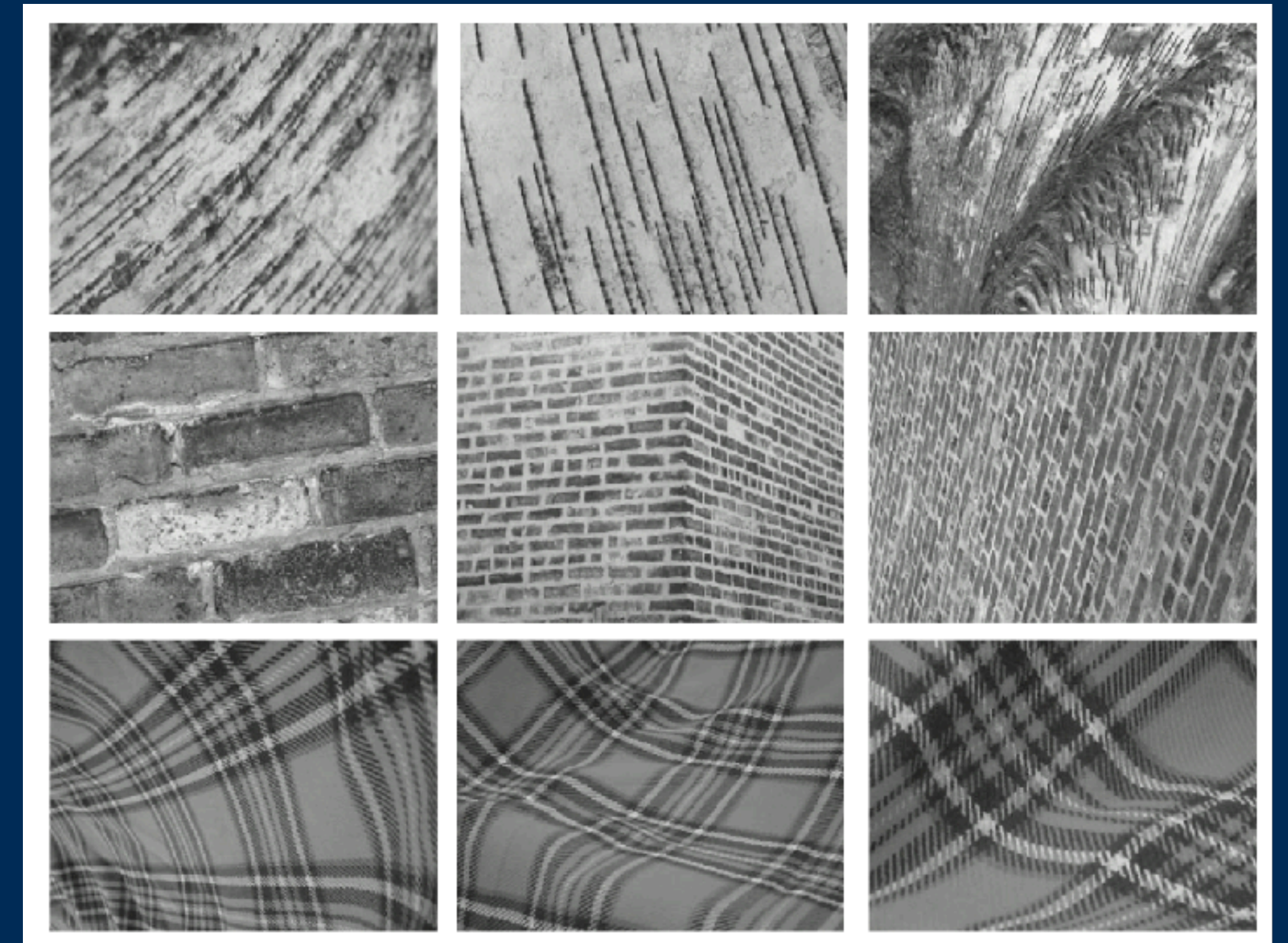
scattering transform in computer vision

CUREt database



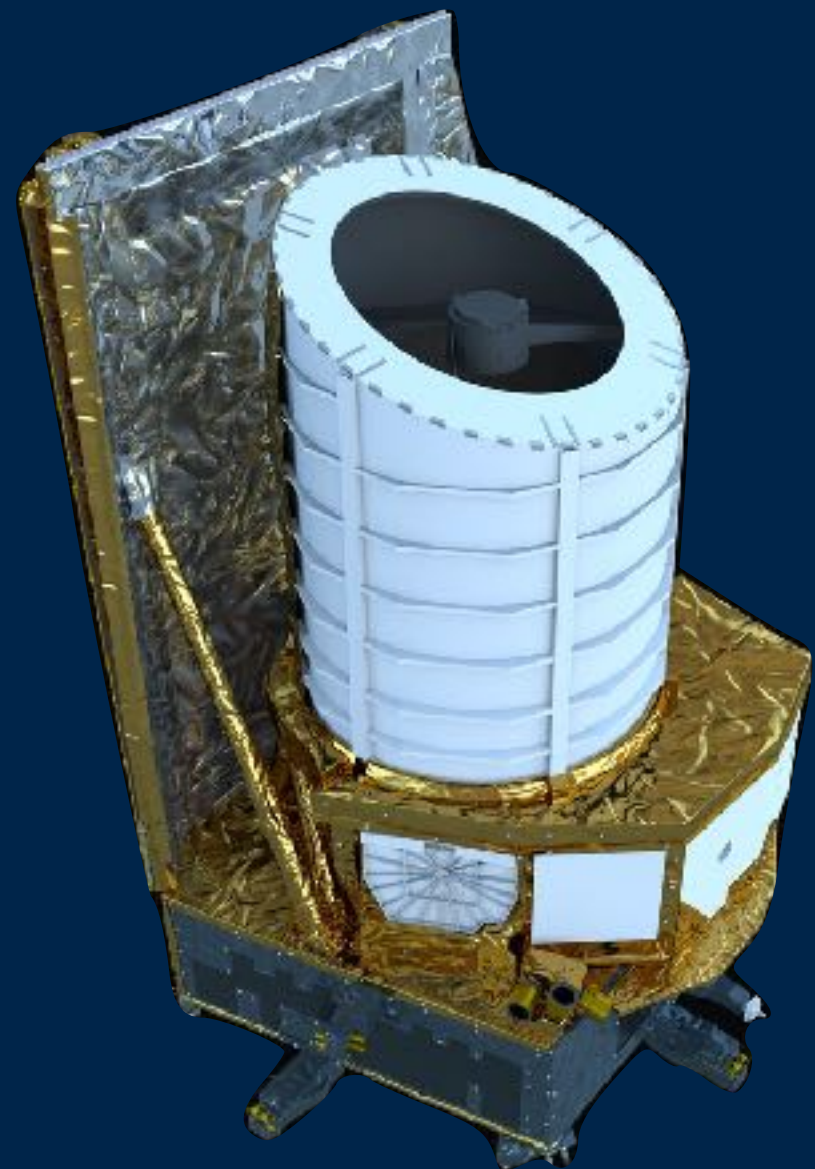
Bruna & Mallat 2013

UIUC database



Sifre & Mallat 2013

weak lensing cosmology



raw
data



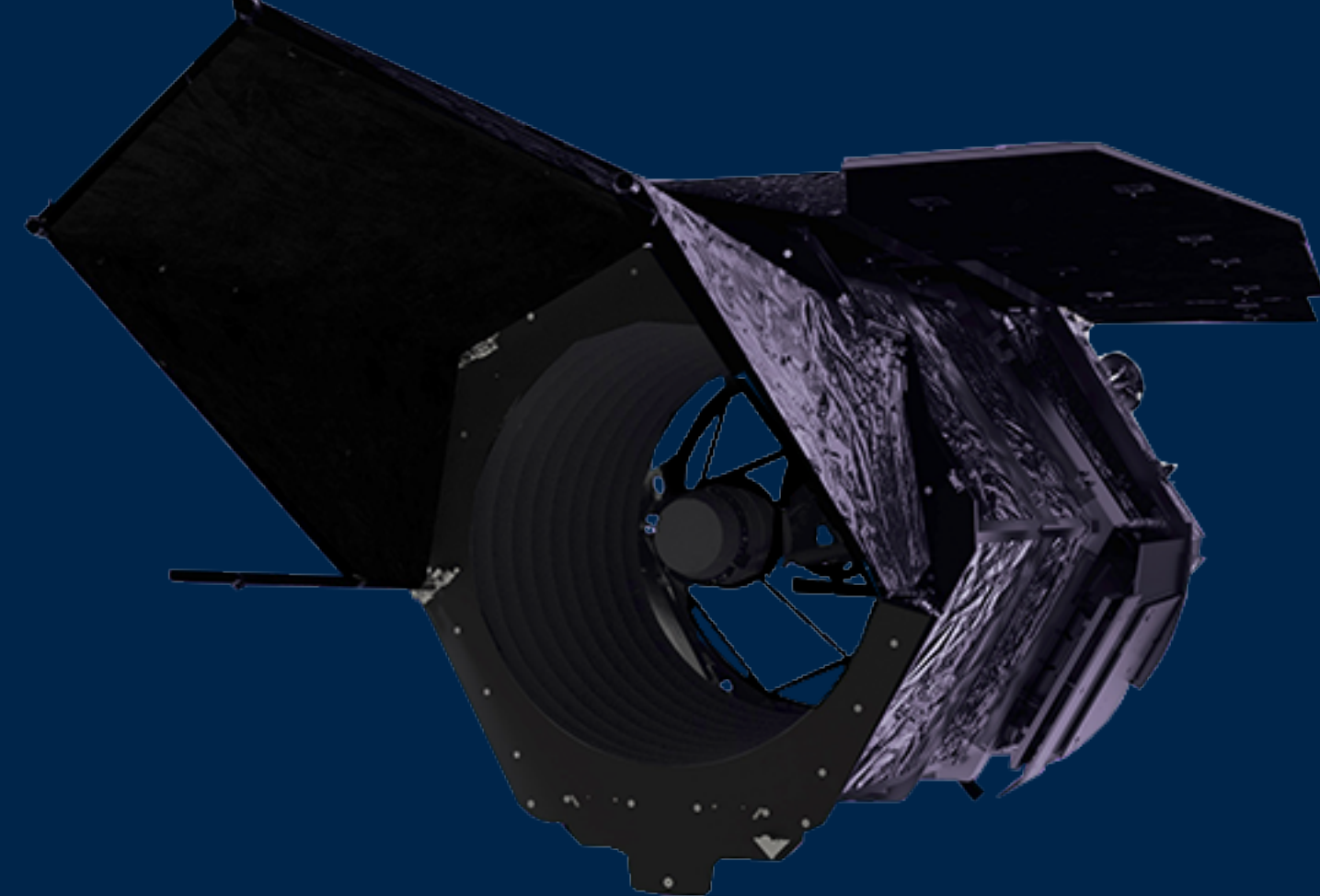
galaxy
catalog



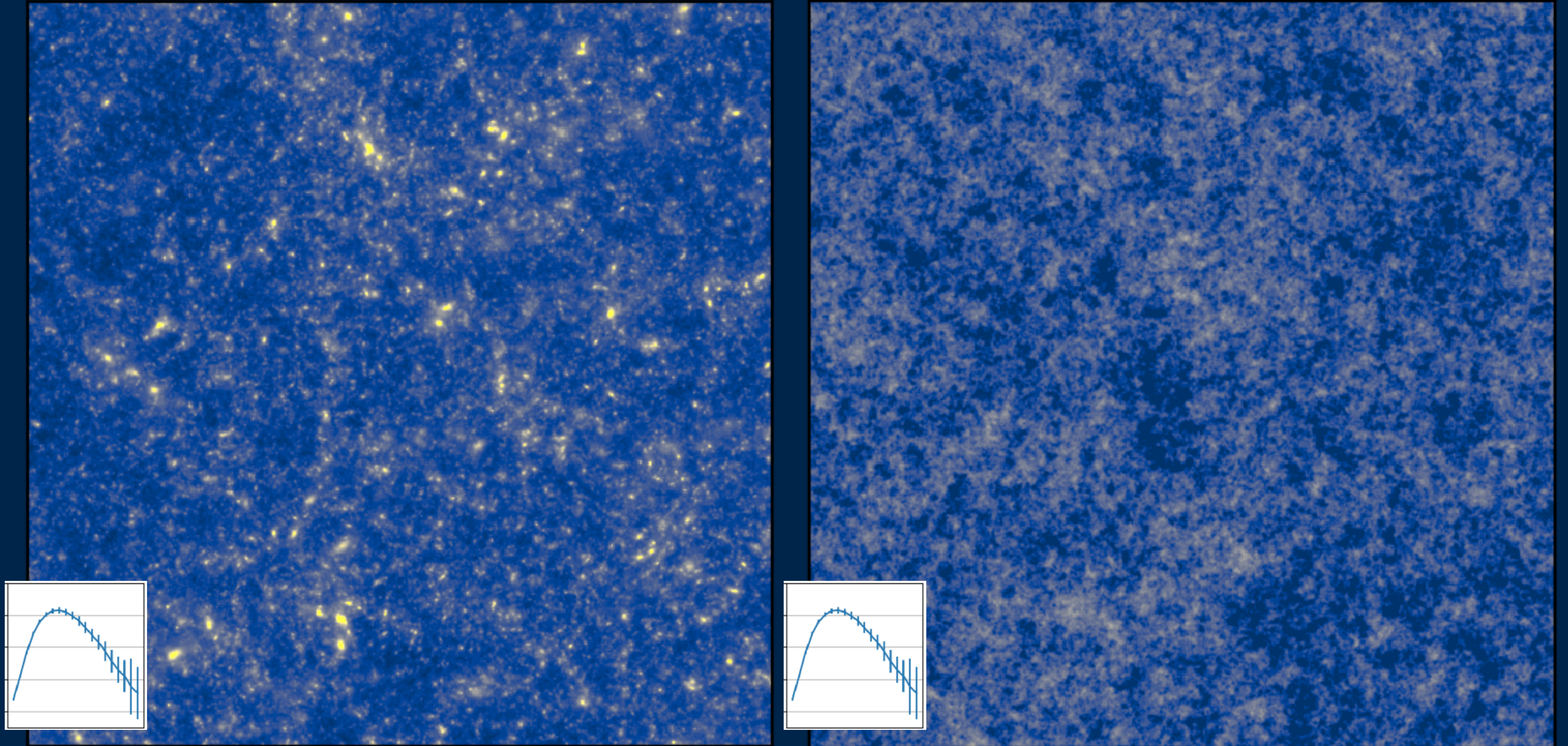
mass
map



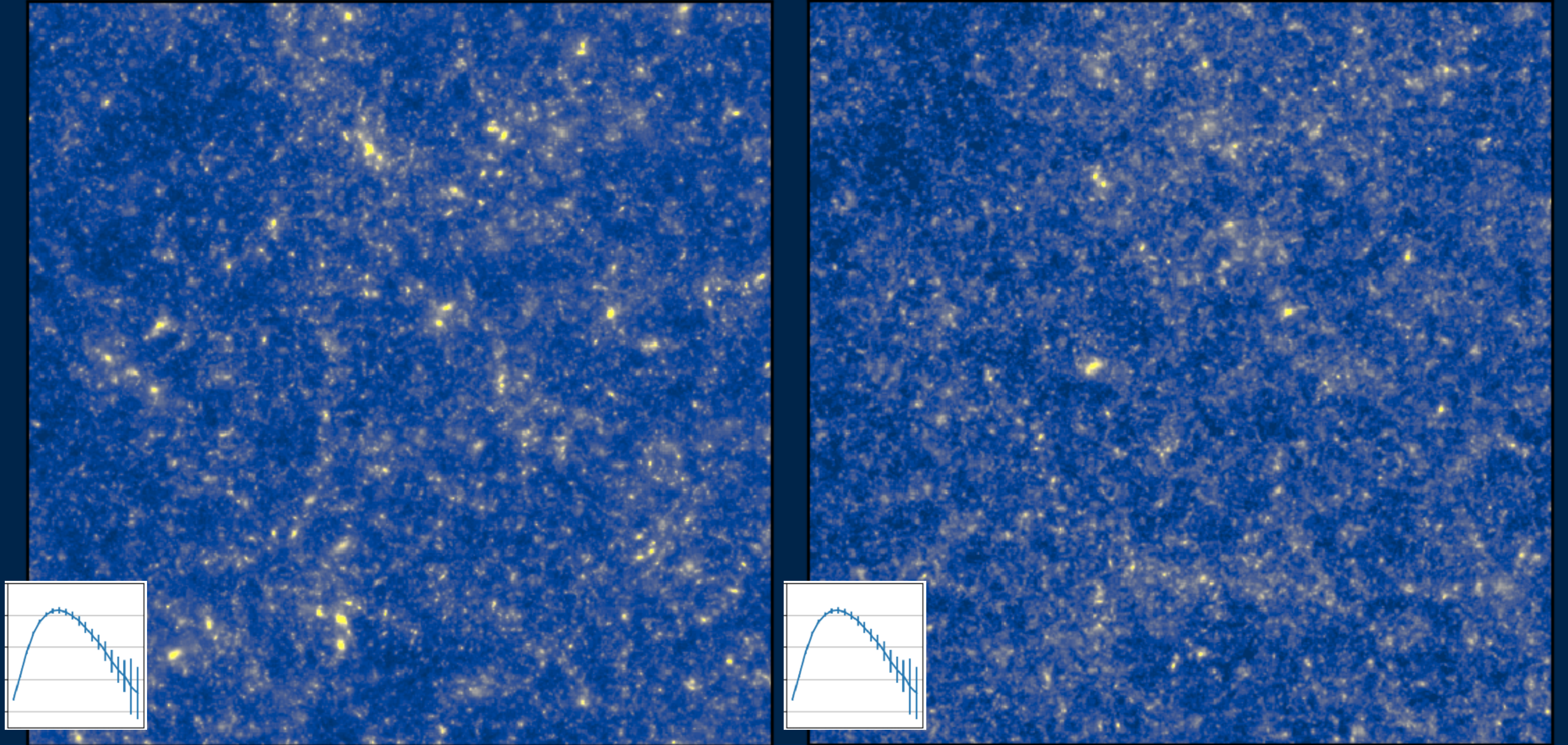
$\sigma_8, \Omega_m, w, M_\nu$

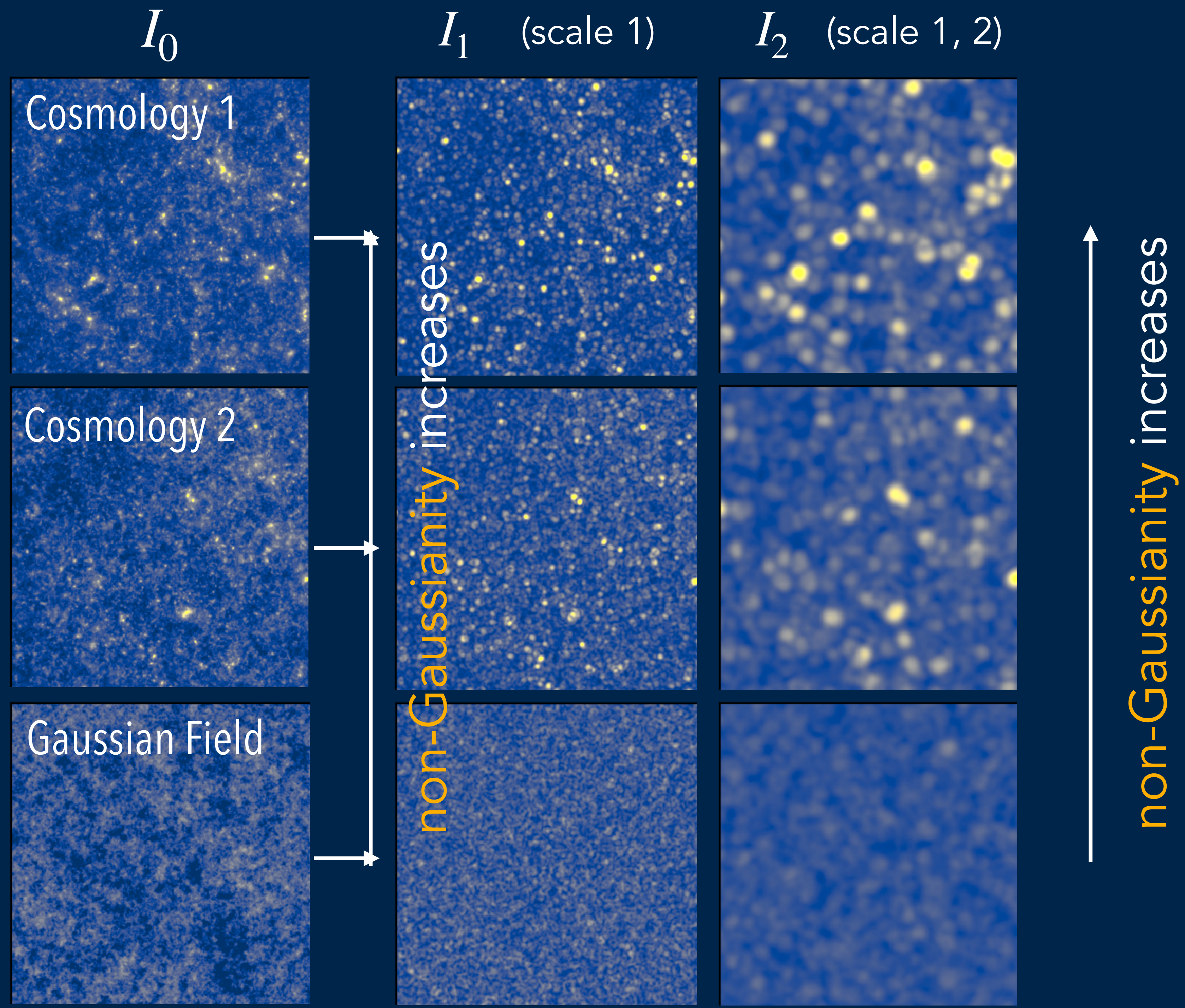


simulated mass maps (from Columbia lensing group)

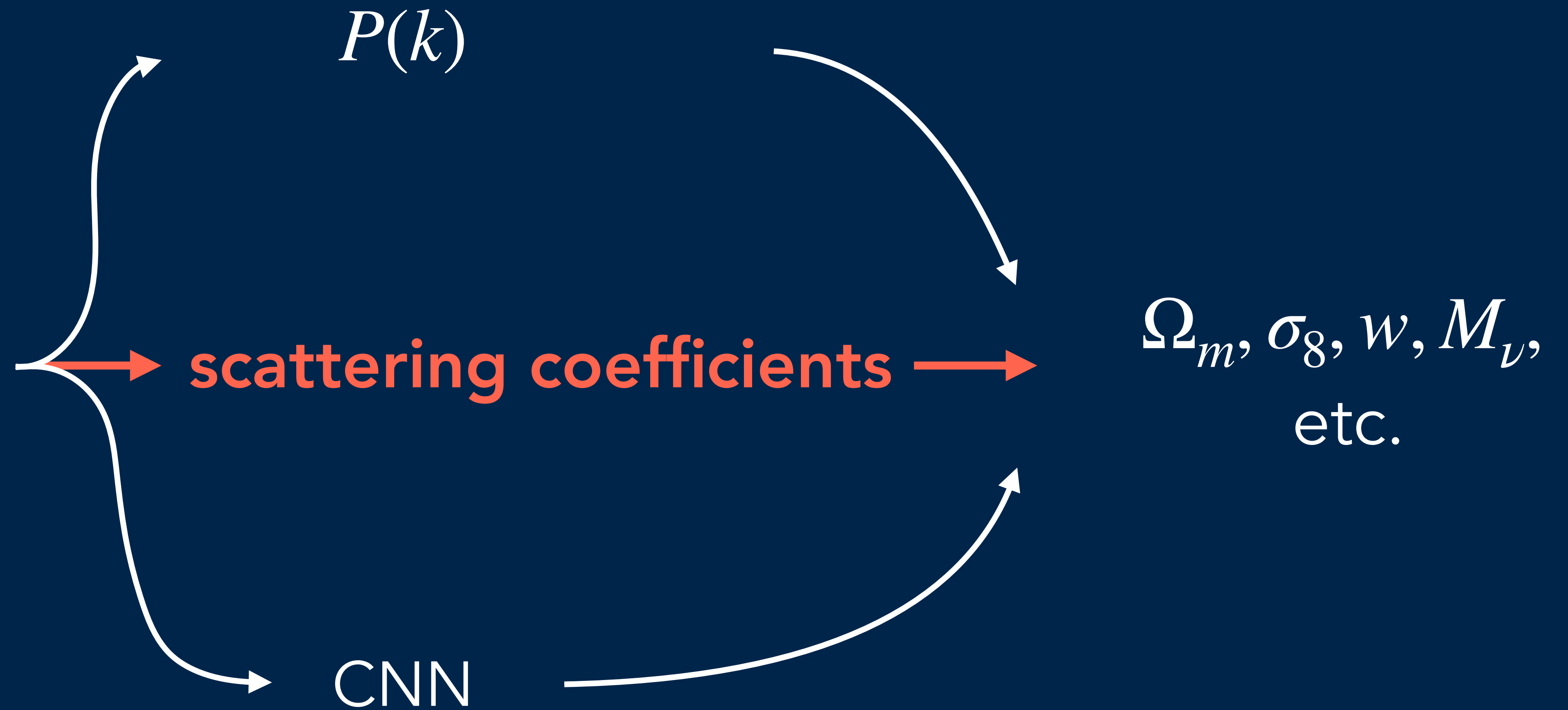
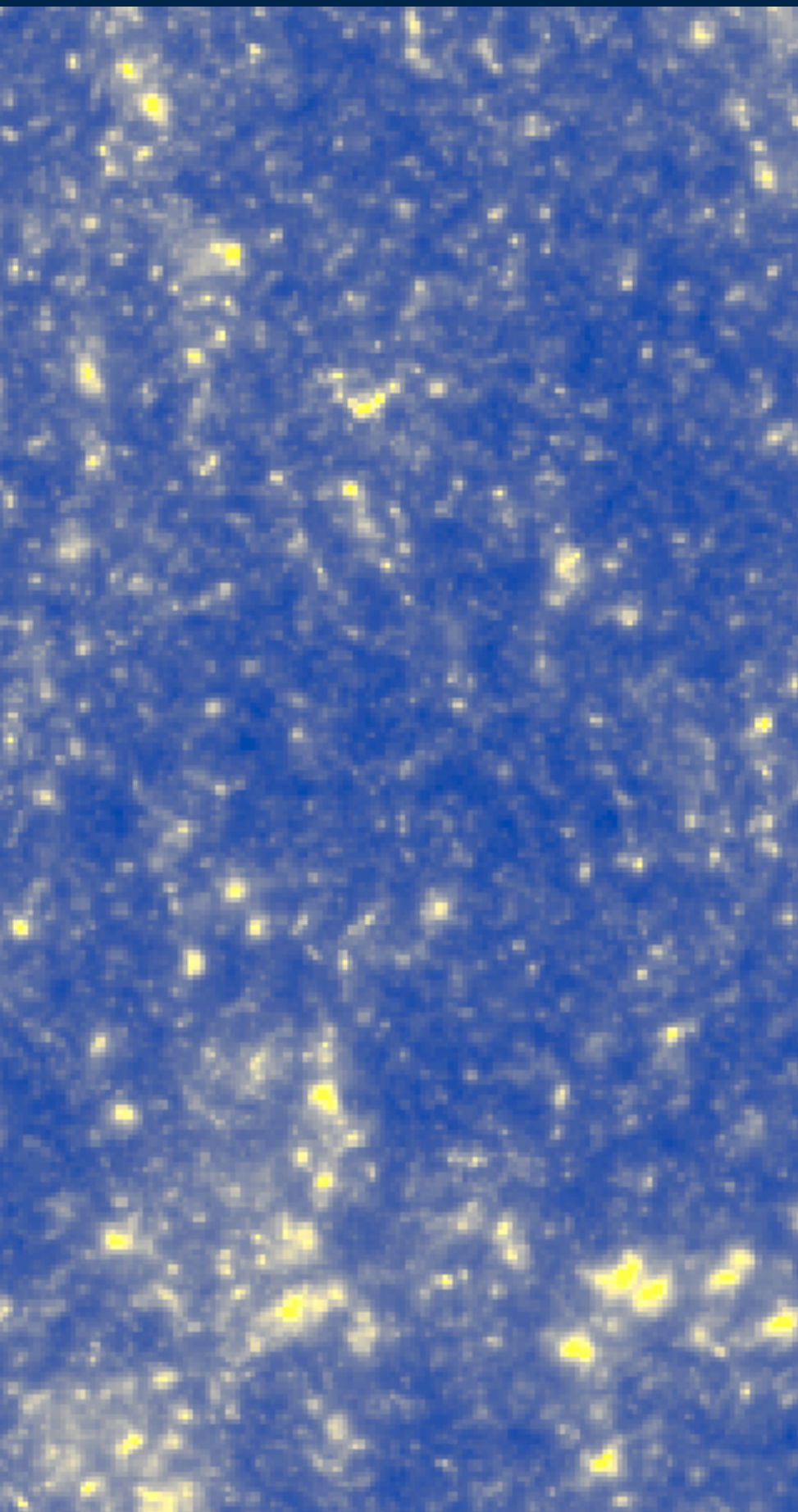


simulated mass maps (from Columbia lensing group)



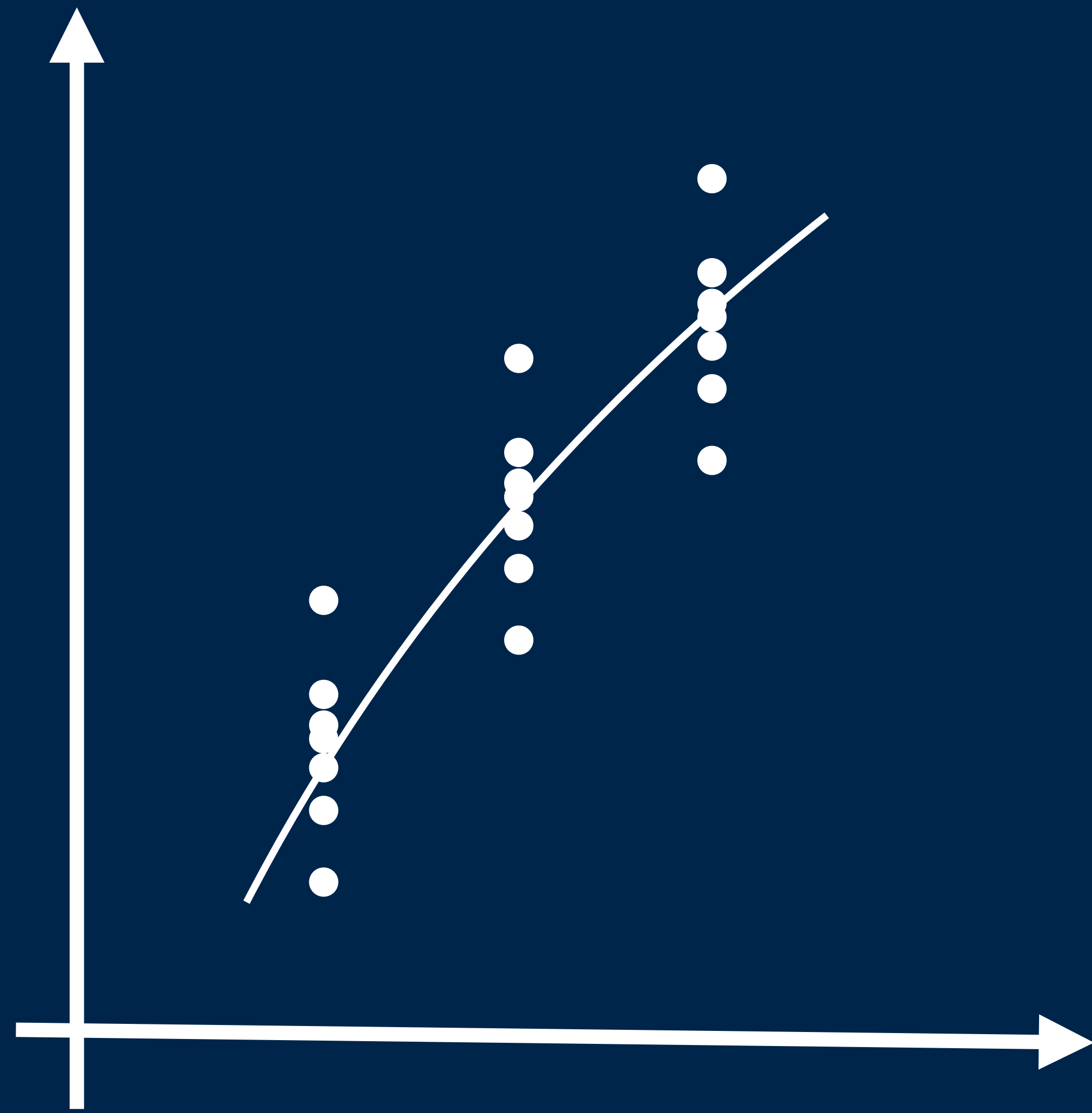


weak lensing cosmology



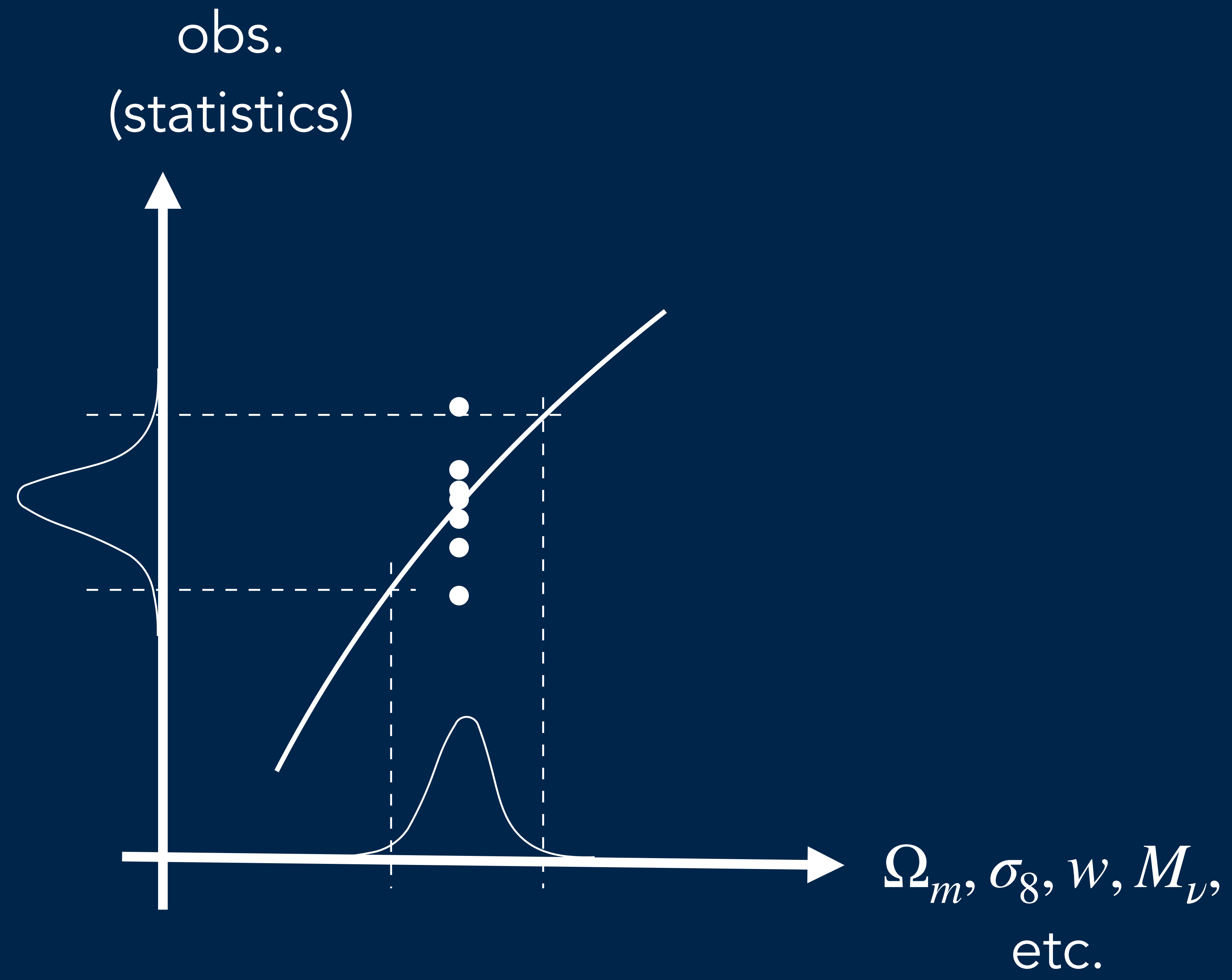
Fisher forecast

obs.
(statistics)

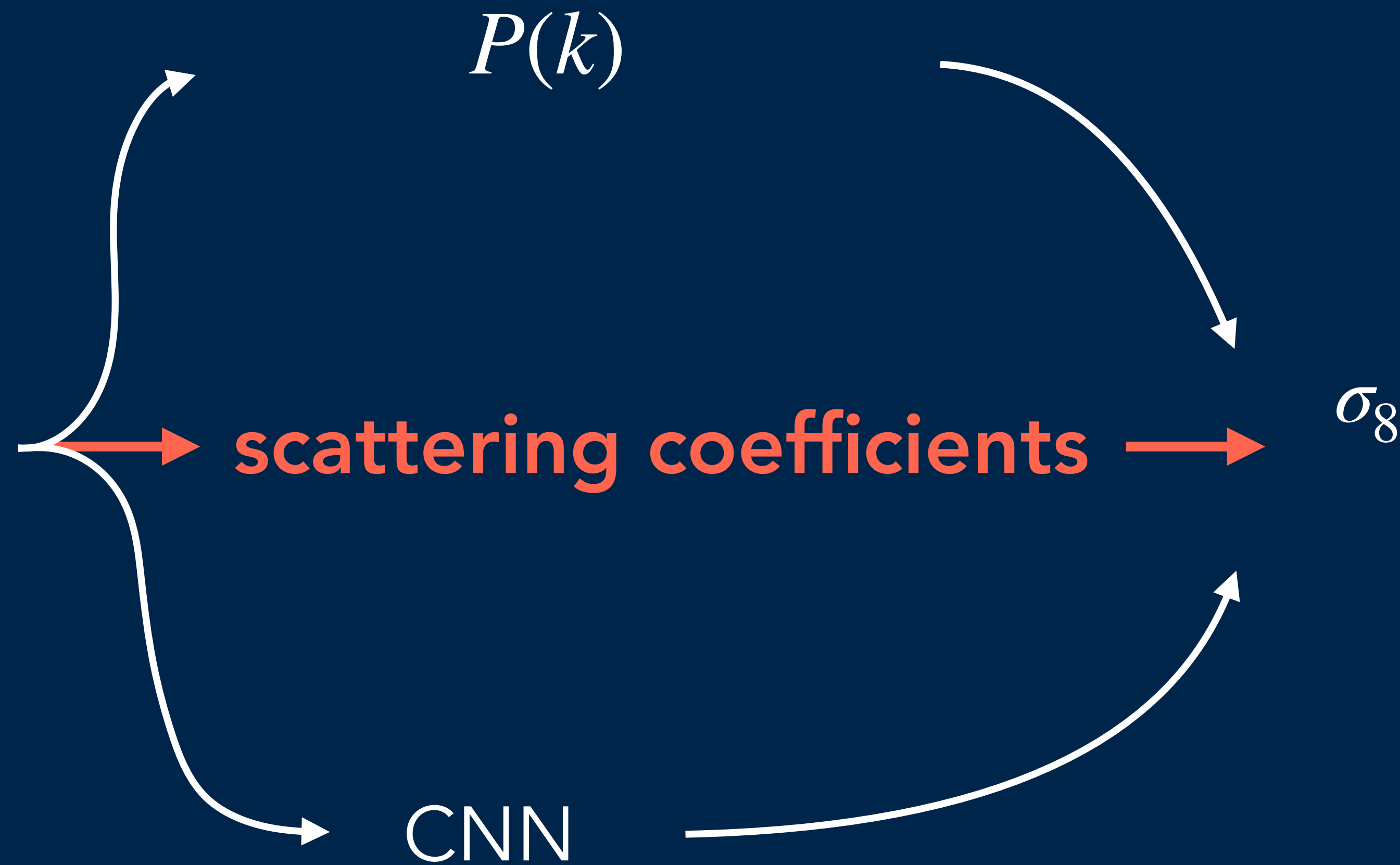


$\Omega_m, \sigma_8, w, M_\nu,$
etc.

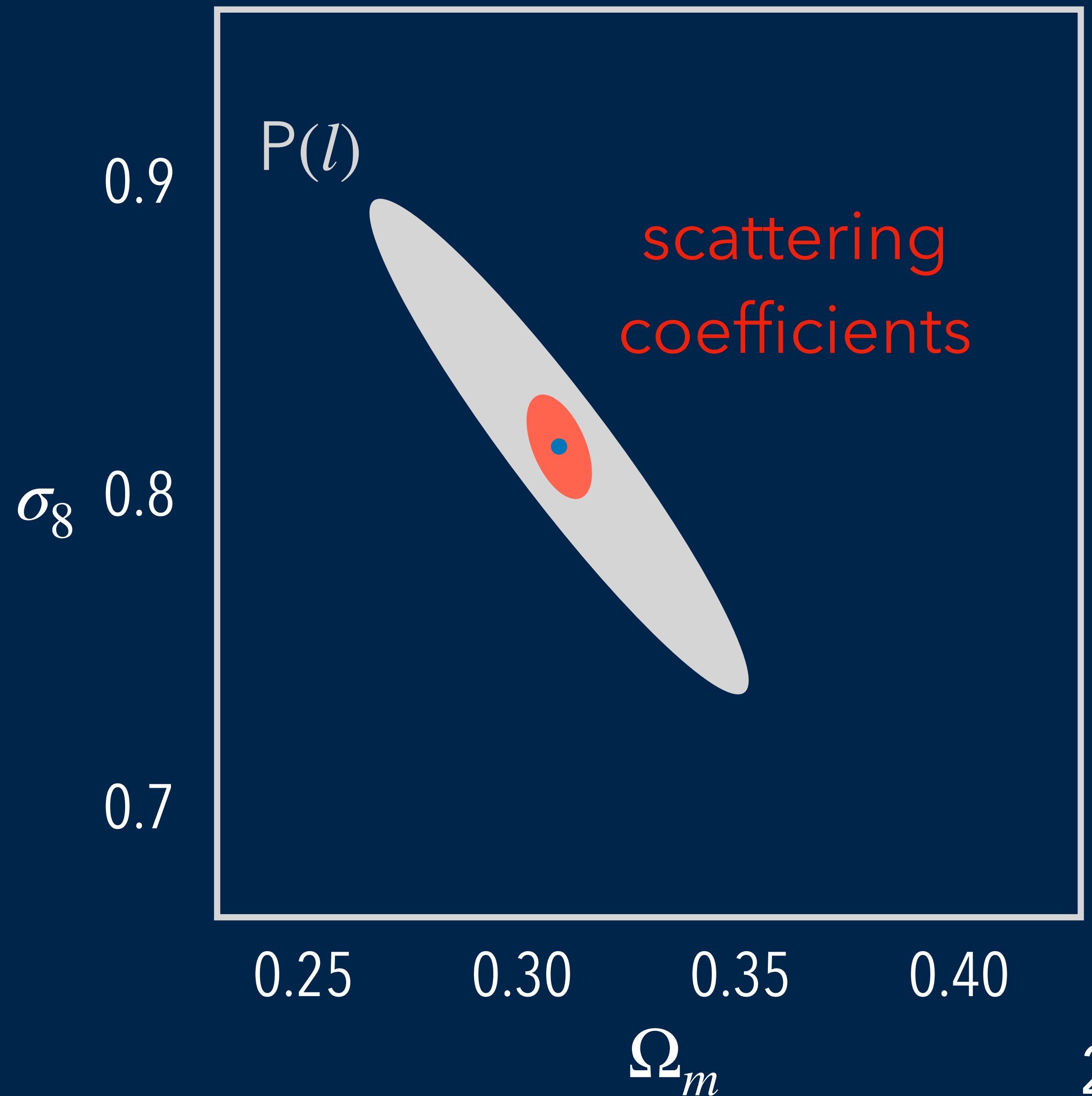
Fisher forecast



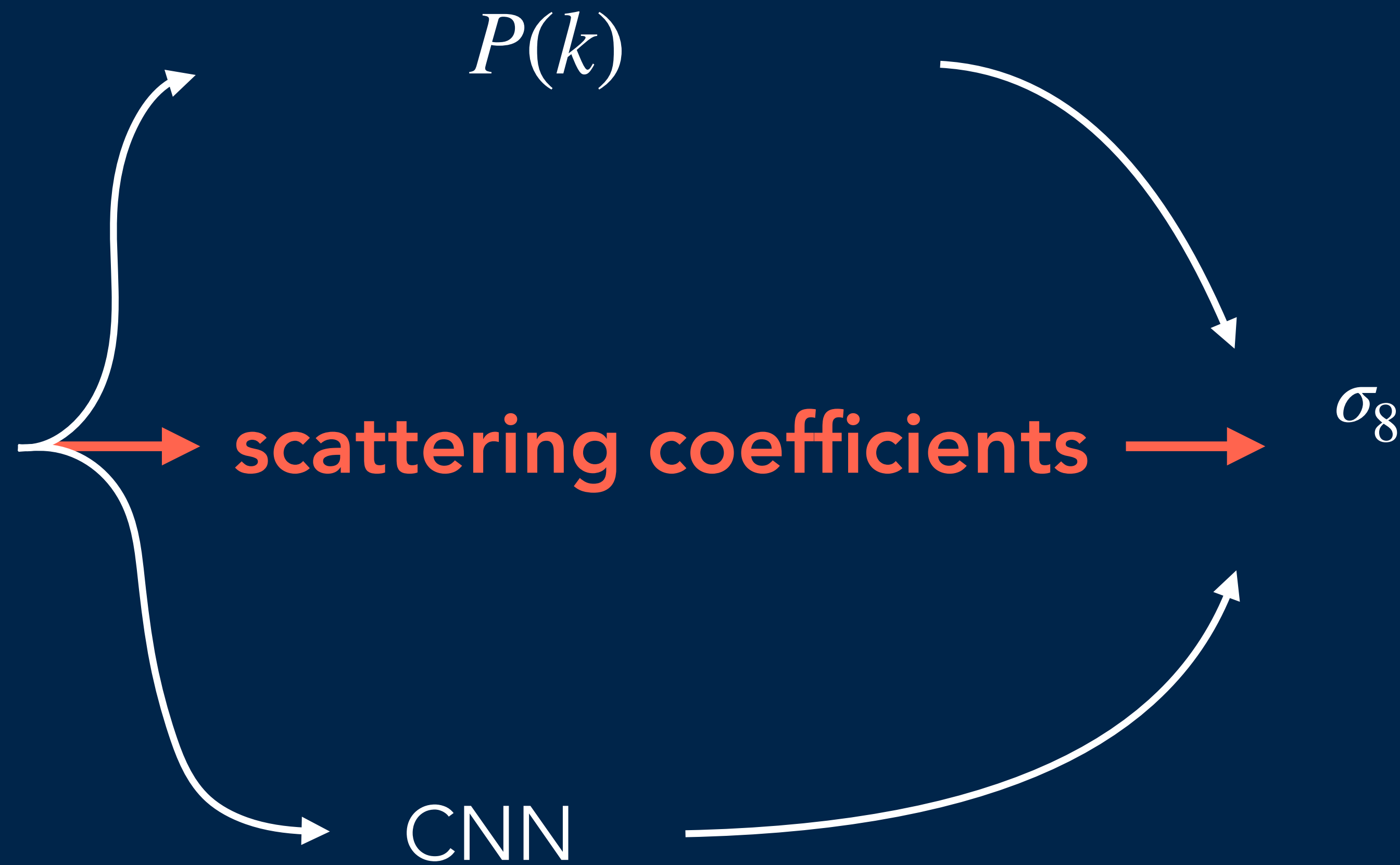
weak lensing cosmology



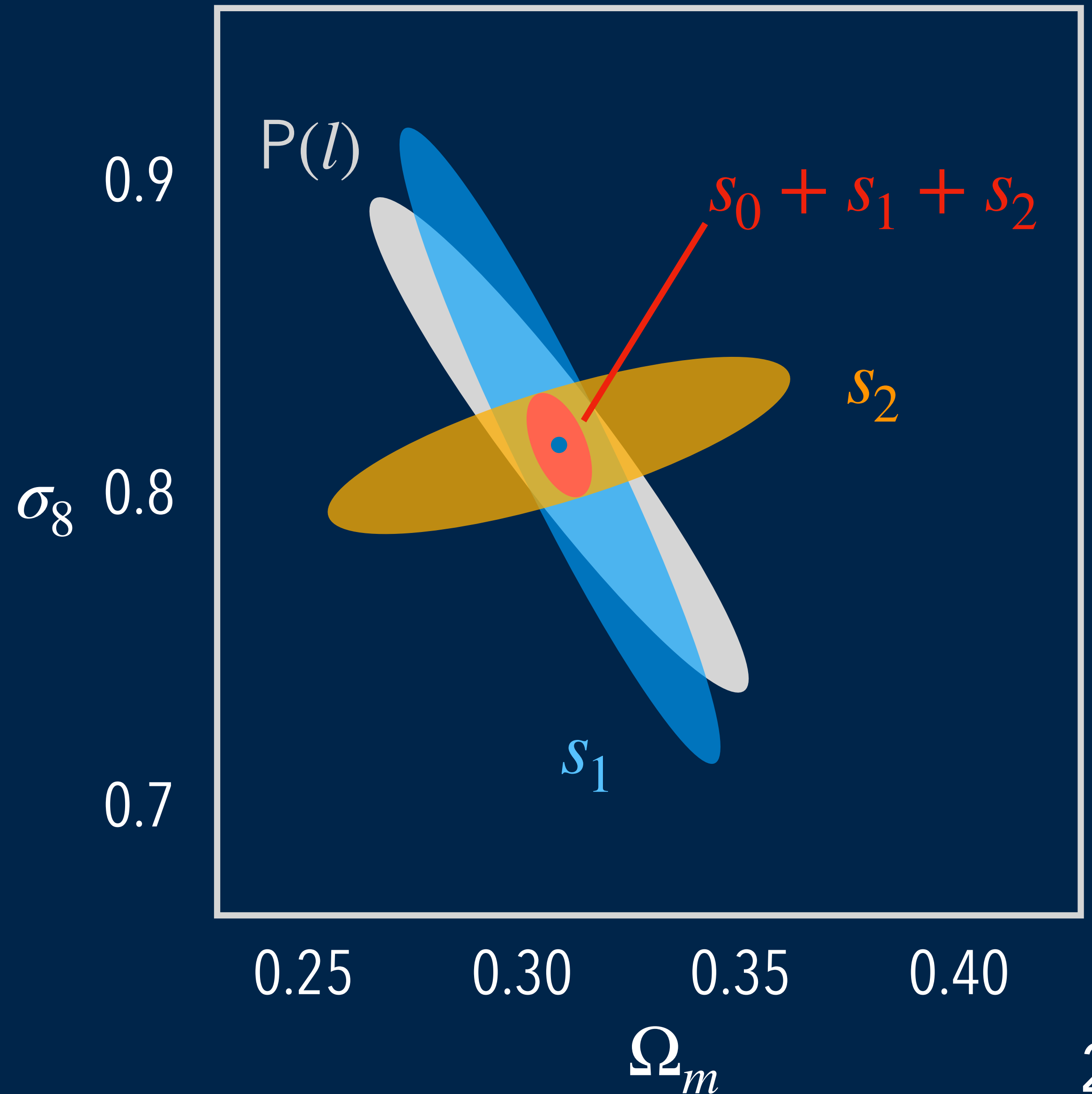
3.5x3.5 deg² noiseless map
scale range: 1 arcmin to 3.5 deg



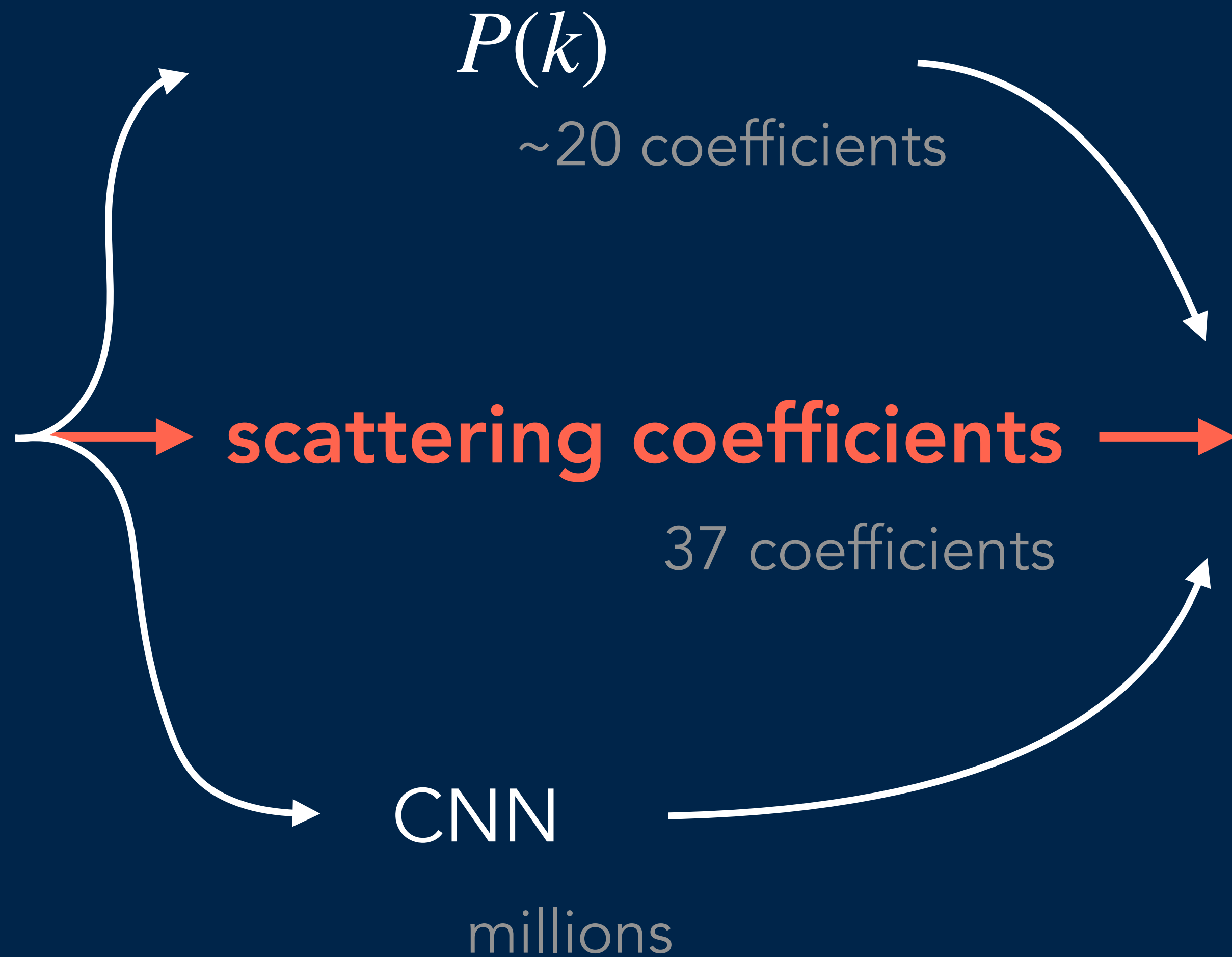
weak lensing cosmology



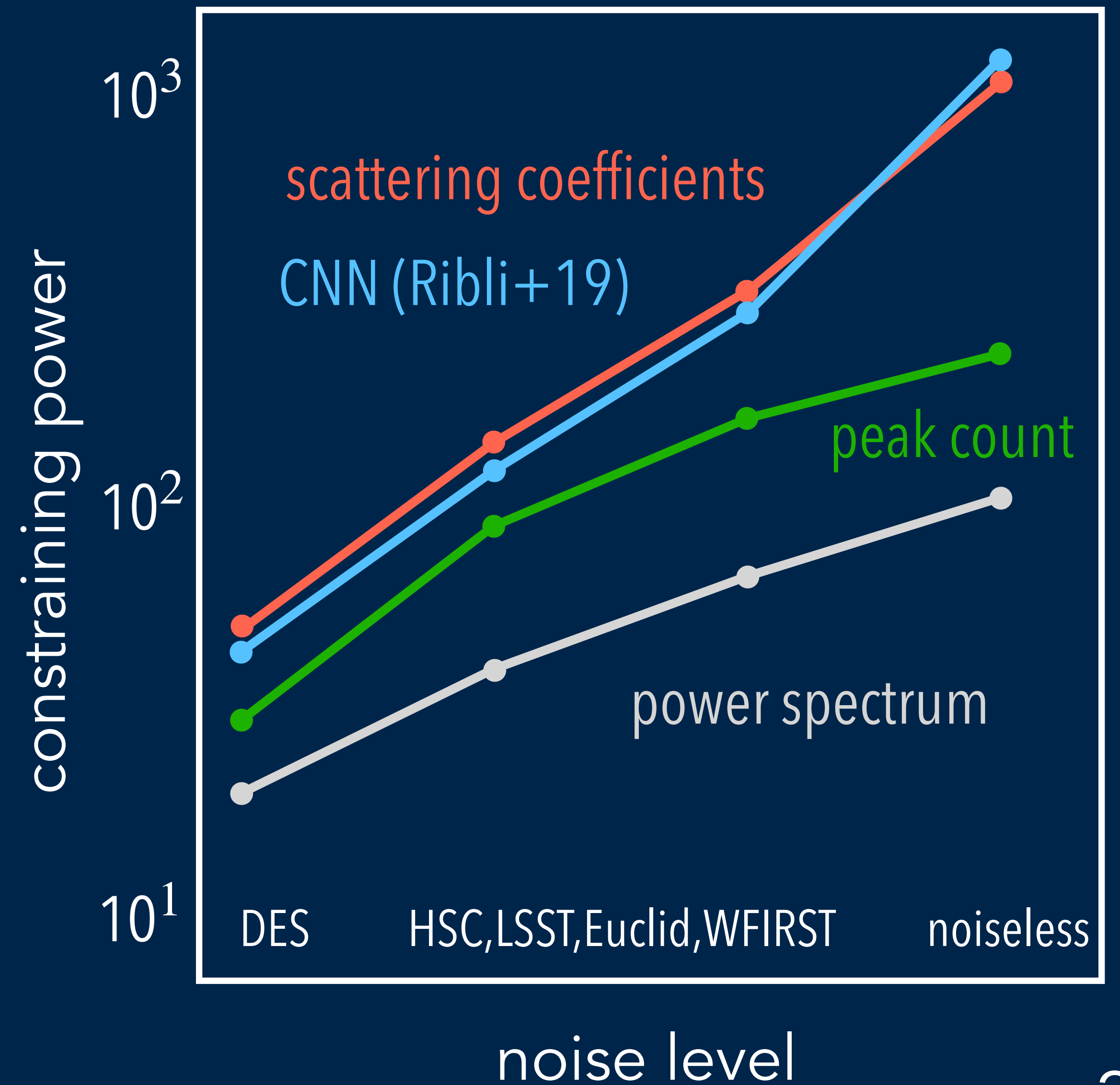
3.5x3.5 deg² noiseless map
scale range: 1 arcmin to 3.5 deg



weak lensing cosmology



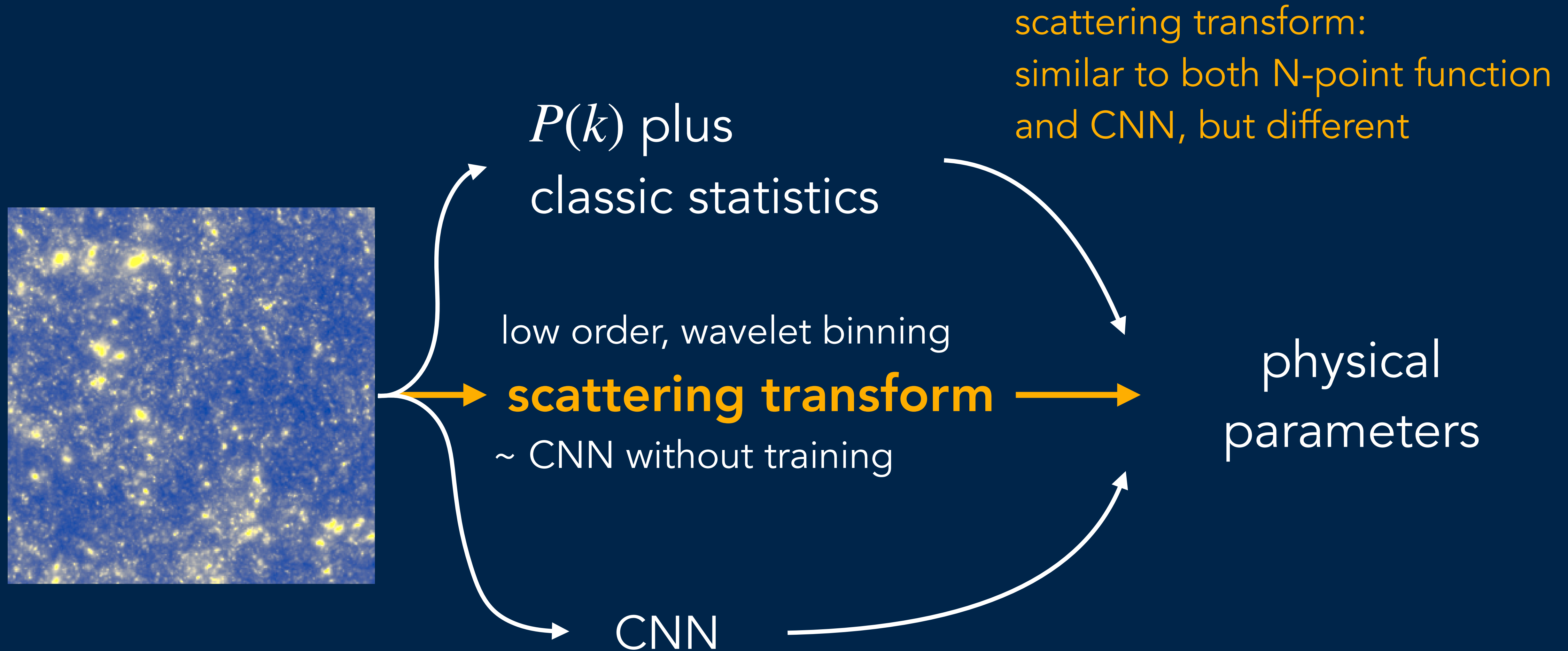
scale range: 1 arcmin to 3.5 deg



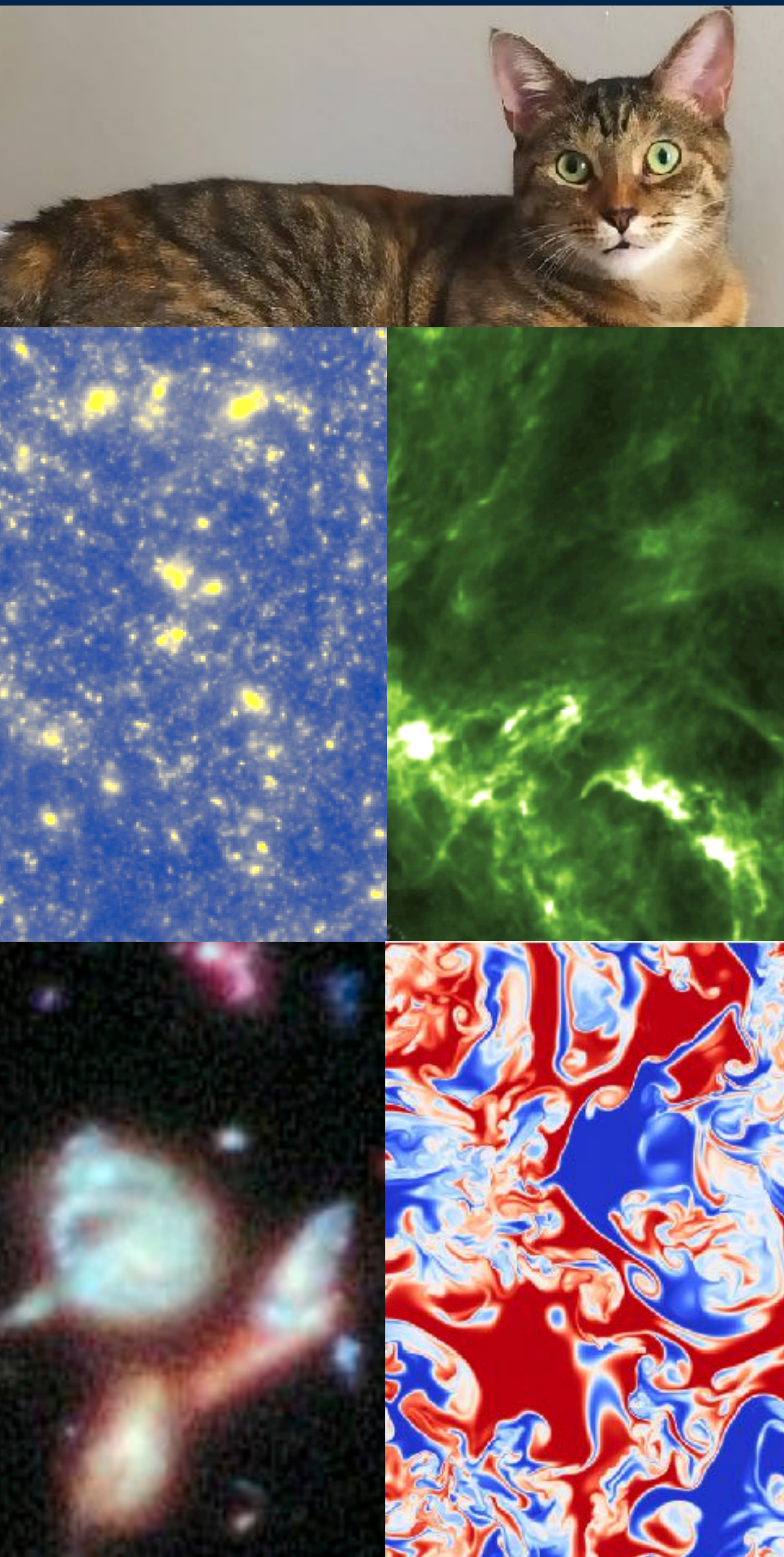
dark energy & neutrino mass
on real data ...



How do we extract information from a field?



How do we extract information from a field?



$P(k)$ plus
classic statistics

scattering transform

CNN

physical
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