The scattering transform: a CNN without training

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COSMO21 May 22, 2024



complex data





a vocabulary power spectrum P(k)?











simple information

data exploration no model

> classification discrete model

parameter inference continuous model



Non-magnetic turbulence

http://turbulence.pha.jhu.edu

Magnetic turbulence





information

Limitations: Iose information too many coefficients



How do we characterize a field?



power spectrum

scattering transform



CNN

cosmic density map

2-order statistics





slices of 3D MHD simulation



scattering statistics





Synthesized with scattering statistics



Ising model Turing pattern











sea temperature

solar UV image

cosmic matter



generated with scattering statistics (translation invariant)



How do we characterize a field?



scattering transform

from power spectrum to scattering transform

 $P(k) \propto \langle I \star e^{ikx} \rangle^2$





local kernel $\psi(x)$

 $S_1(k) = \langle \left| I \star \psi \right| \rangle$

 $\langle \cdot^2 \rangle = P(k)$



modulus





convolutional network

scale $j \approx 1/k$





wavelets and sparsity

vision cells in brain (Hubel & Wiesel 1968)

sparse representation of natural images (Olshausen & Field 1996)

> kernels learned in AlexNet (Krizhevsky, Sutskever, & Hinton 2012)

Dirac wavelet Fourier



close to Gabor wavelets



interpretation

modification of 2, 3, 4-point: log bin + stable non-linearity

structure sparsity $s_{21} \equiv \overline{S_2} / S_1$

Cheng & Menard, 2021b

structure shape $s_{22} \equiv S_2^{\parallel} / S_2^{\perp}$

 $l_1 \parallel l_2$:

Cheng & Menard, 2021b

arranged by 2nd-order scattering coefficients

UMAP 2

Cheng & Menard, 2021b

cosmological simulations

Matilla Zorrilla et al. 2016, Gupta et al. 2018 (Columbia lensing group)

inferring cosmological parameters

Cheng et al. 2020

inferring cosmological parameters

P(l) ~20 coefficients

scattering coefficients 37 coefficients

millions

informative compact			
robust 10 ³ Volumer 10 ²	S ₁ , S ₂ CNN (Rib	p(l)	peak cou
10 ¹	DES HSC	LSST,Euclid,WFIRST	noise
		noise level	

HSC year 1

HECTOMAP

WIDE12H

Cheng, Marques, Grandón, Thiele, Shirasaki, Ménard, Liu, 2024

VVDS

mock mass map with no noise

improvement from ST

high S8 value

_	ACT lensing+BAO SPT lensing+BAO <i>Planck</i> TT+TE+EE+lowE		Madhavacheril+23 Bianchini+20 Aghanim+20a	
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9 ' '	1.0		1.1	
n/0.3				

BOSS galaxies (Georgias Valogiannis)

spherical generative model (Matt Price)

one-variable illustration: moments vs scattering

folding: better than High-order moments

folding the core

Cheng & Menard, 2021b

extension to cross-correlations

Simulation

Synthesized with scattering correlations

extension to cross-correlations

input image synthesis from scattering transform alone $Corr(I, I \star \psi)$

lensing field

cosmic web

25%

synthesized

histogram

50%

75%

lensing

overall amplitude of scattering correlations

spread

sparse

x1 (original) x0 xЗ Cheng, Morel, Allys, Ménard, Mallat, 2024, PNAS Nexus

angular variation of scattering correlations

curvy

pointy

x1 (original)

cosmic lensing

cosmic web

modeled fields with scattering statistics (200 steps)

Cheng, Morel, Allys, Ménard, Mallat, 2024, PNAS Nexus

2D fluid

magnet fluid

anisotropic fluid

How do we characterize a field?

information

number of statistics relation to physics

