

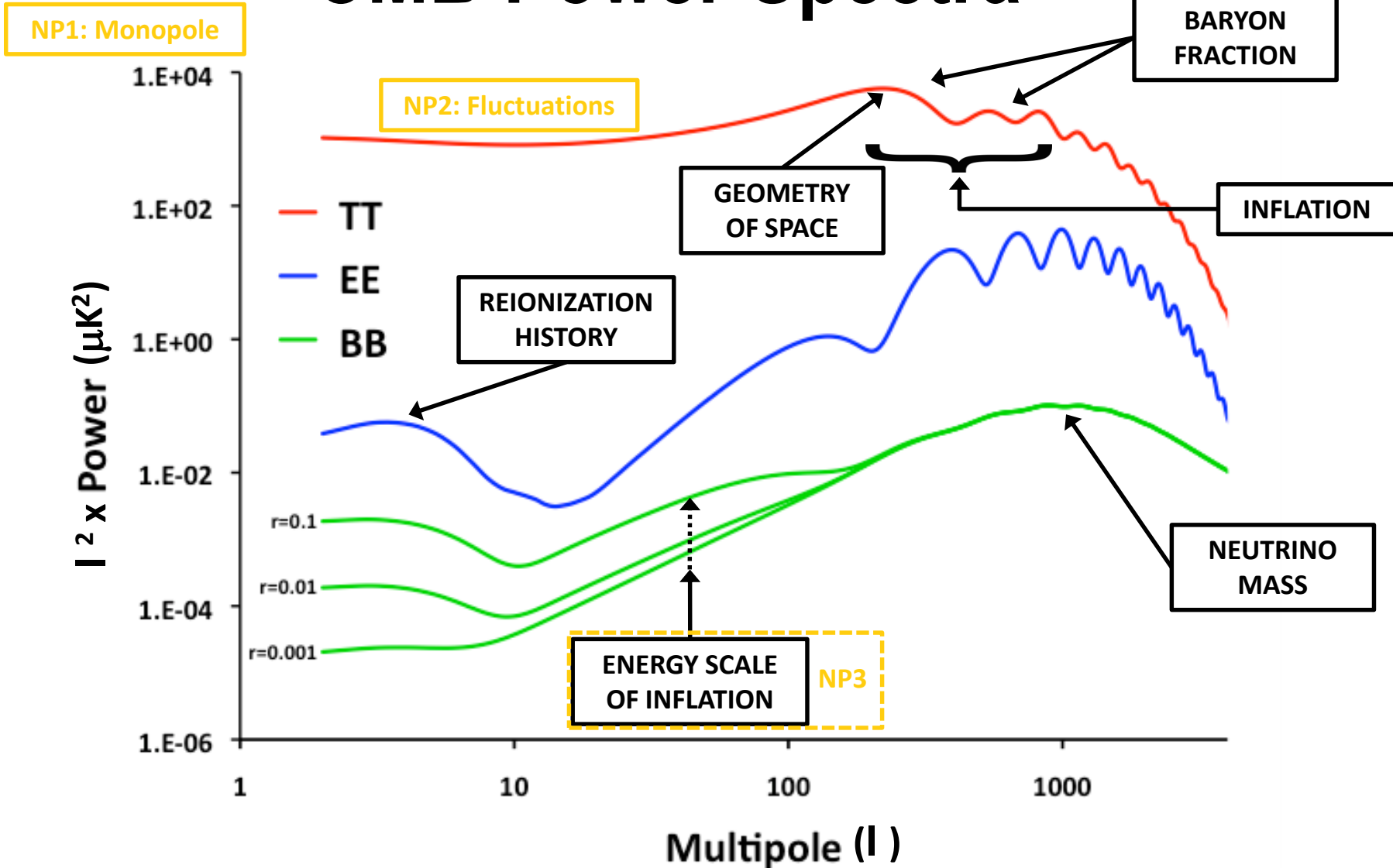
Data Issues

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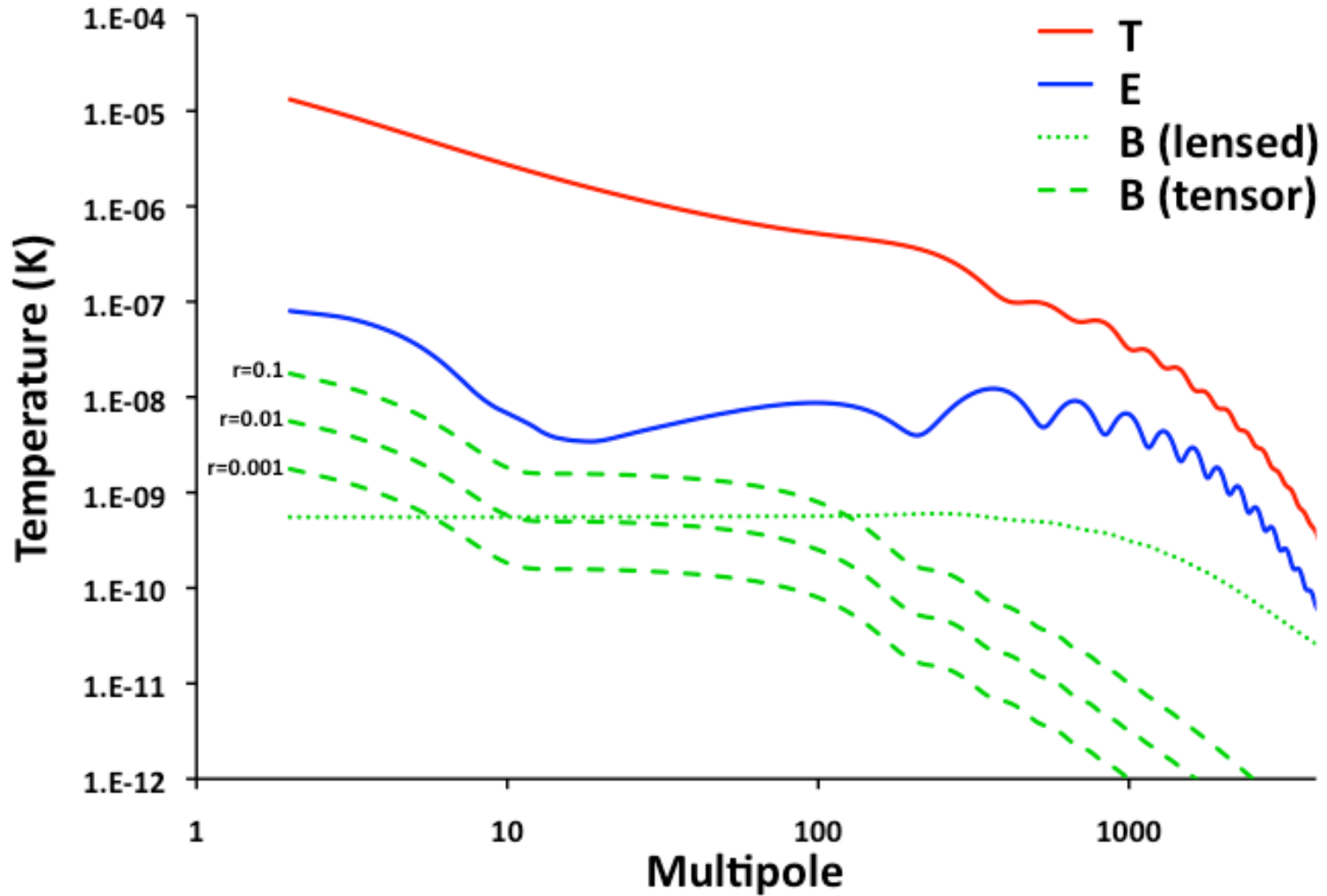
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CMB Power Spectra



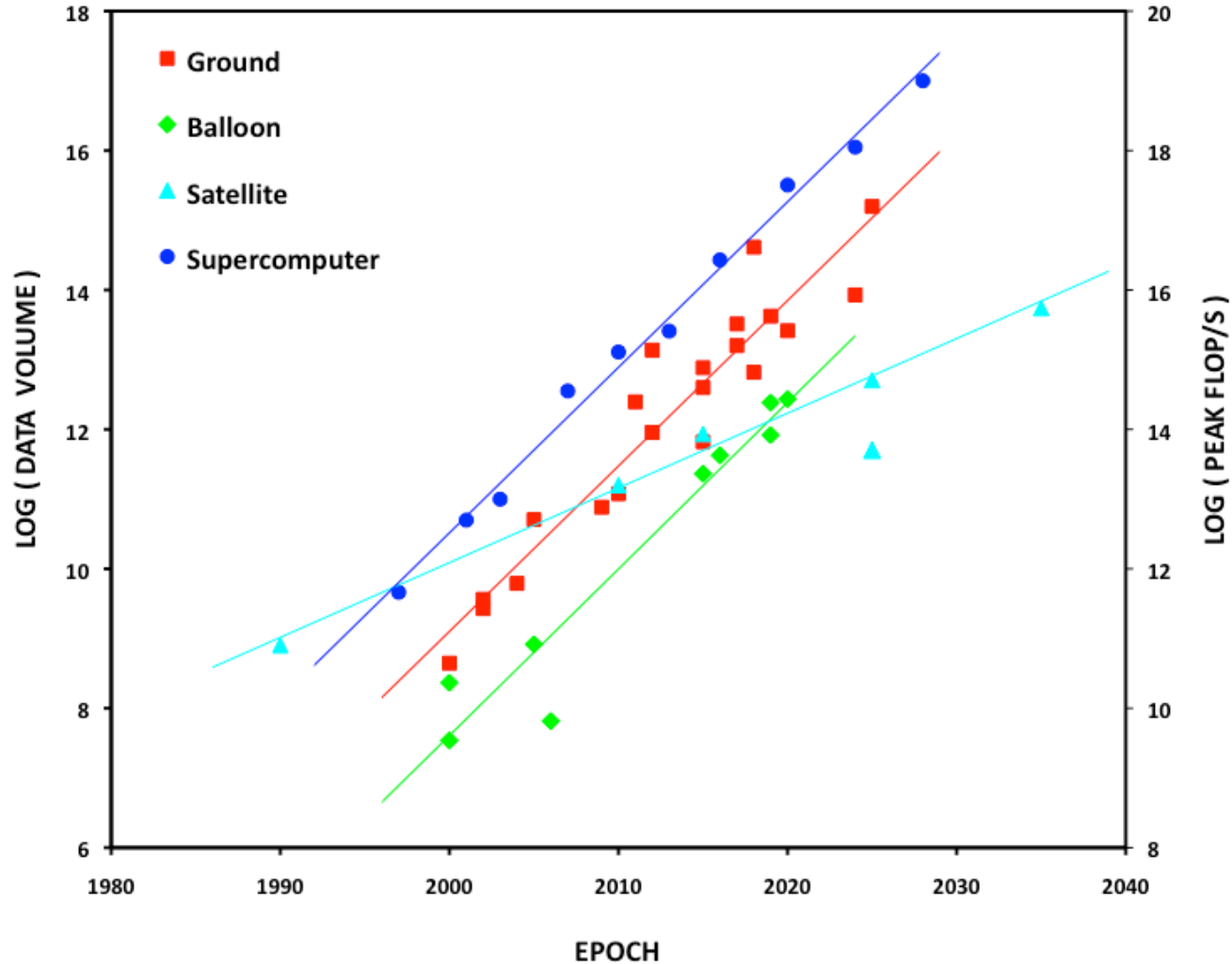
Another View ...



CMB Signals

COMPONENT	AMPLITUDE (K)	ERA
TT : Monopole	1	1968 (Penzias & Wilson)
TT : Dipole	10^{-3}	1971 (Henry/Conklin)
TT : Anisotropy	10^{-5}	1990 (COBE)
TT : 1 st Peak	10^{-6}	2000 (BOOMERanG, MAXIMA)
EE : Reionization	10^{-7}	2005 (DASI)
BB : Lensing	10^{-9}	2015 (SPT, POLARBEAR)
BB : Gravity Waves	$< 10^{-9}$?

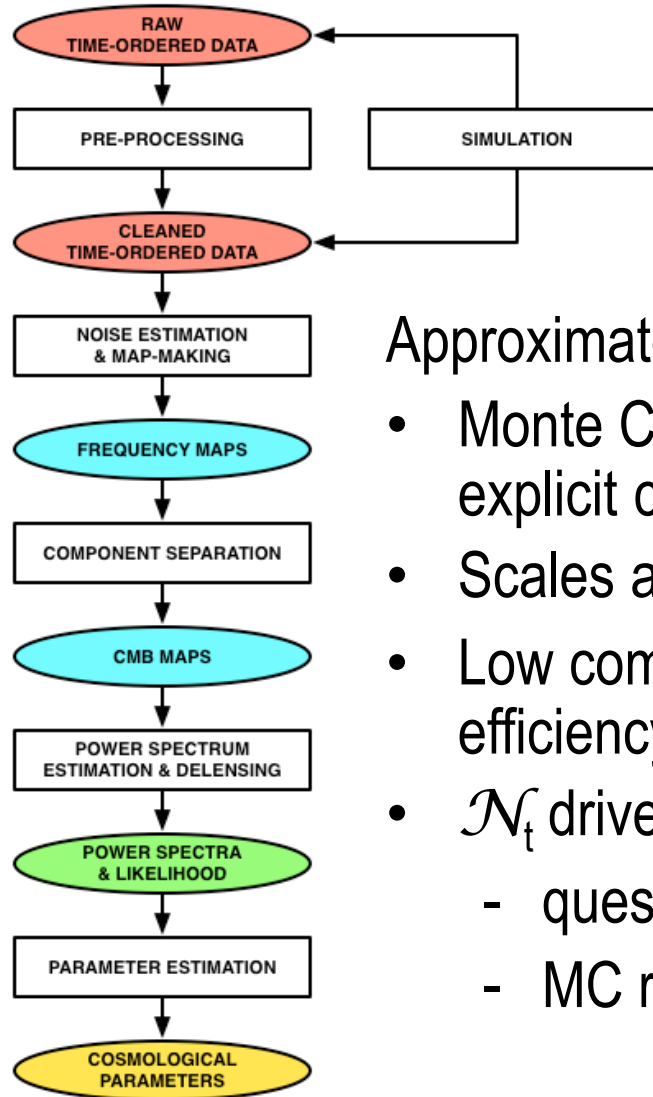
Fainter Signals Mean Bigger Datasets



CMB Data Analysis

Exact analysis:

- Maximize Gaussian likelihoods
- Scales as $O(\mathcal{N}_p^3)$
- High computational efficiency
- \mathcal{N}_p driven up by
 - high resolution
 - large sky coverage
 - many frequencies
 - polarization



Approximate analysis:

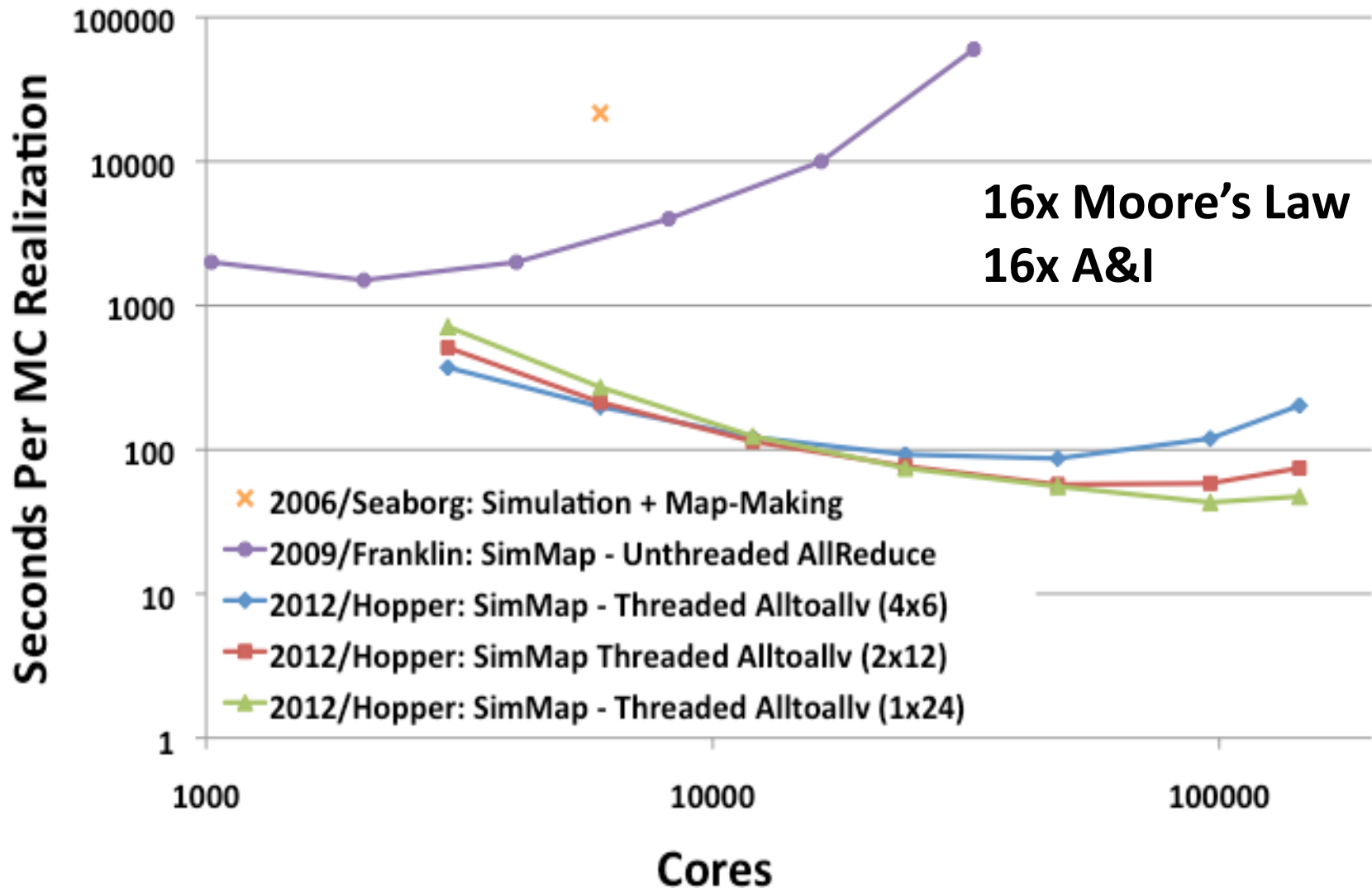
- Monte Carlo replaces explicit covariance
- Scales as $O(\mathcal{N}_t)$
- Low computational efficiency
- \mathcal{N}_t driven up by
 - quest for S/N
 - MC realizations

Prefactors & Efficiencies

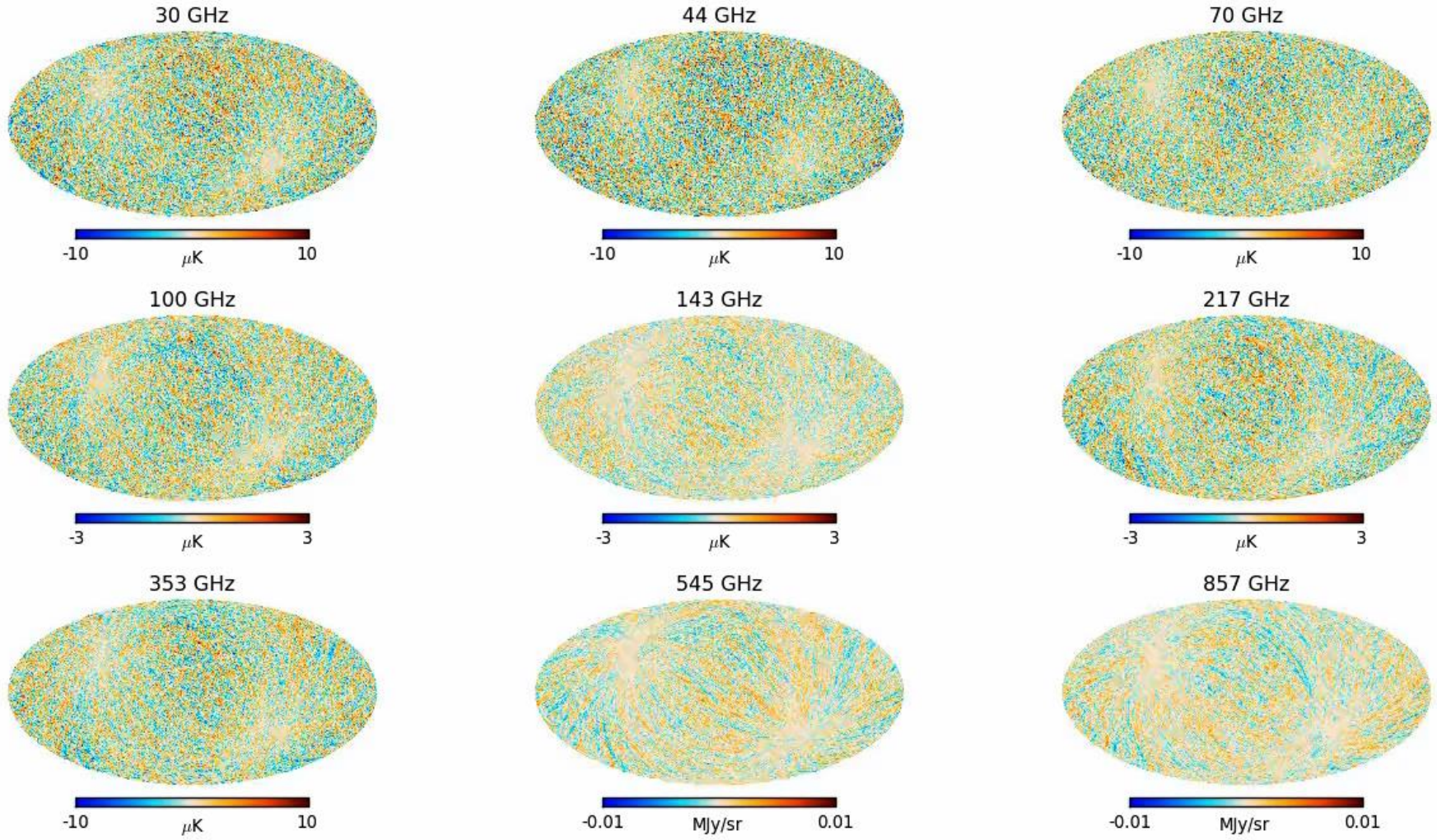
- Exact analysis dominated by covariance matrix operations
 - Floating point operations $\sim 10^{2-3} \mathcal{N}_p^3$
 - Efficiency $> 80\%$
 - Intractable* for $\mathcal{N}_p > 10^5$ (2000), 10^6 (2015), 10^7 (2030)
 - Only viable for small patch/low-resolution data
- Approximate analysis dominated by Monte Carlo map-making
 - Floating point operations $\sim 10^4 \times 10^2 \times 10^1 \mathcal{N}_t$
 - Efficiency $< 1\%$
 - Intractable* for $\mathcal{N}_t > 10^9$ (2000), 10^{12} (2015), 10^{15} (2030)
 - Viable for current/future missions *if we can maintain efficiency*

* More than 1% NERSC total annual cycles

Planck Monte Carlo Capability



10^4 MC: Planck Noise



~ 1 million CPU-days

High Performance Computing

- The most computationally challenging CMB analyses require:
 - Very many cycles
 - High bandwidth/low latency communication
 - Fast parallel I/O
- These features *define* high performance computing (HPC)
 - supercomputers, not grid/cloud/@home/etc.
- How these features are delivered evolves with architecture
 - clock-speed => node-count => accelerators => many-core => ?
 - communication and (especially) I/O grow slower than Moore's Law
- Analysis implementations have to evolve accordingly.

Coming HPC Challenges

- HPC systems are becoming progressively
 - less homogeneous
 - less balanced
 - more power-constrained
- ... likely requiring our implementations to incorporate
 - run-time self-tuning
 - energy-awareness
 - on-the-fly map analysis
 - unknown unknowns
- What do we do when Moore's Law stops?

Mission-Class Computing

- Long-term CMB missions require:
 - Reliable allocation of persistently state-of-the-art HPC
 - Moore's Law growth during the mission lifetime
 - Over-scoped (shared) systems for bursty usage
 - Access for large numbers of users around the world
 - Cycles and storage both to produce and exploit reduced data
- eg. Planck at NERSC
 - 1% of NERSC cycles per year for 15 years
 - 5 generations of top-10 supercomputer
 - 200+ users from 10+ countries (& 100+ suborbital)

HPC Centers

		Resource		Access	
		Capability (Top 10)	Roadmap (3+ Generations)	Users (100+ Worldwide)	Allocation (10+ Years)
USA	DOE/NERSC	#34/(top10)			
	DOE/LCFs	#2/#5			
	NSF	-			
	NASA	#11*			
Europe	CINECA/BSC	#32/#77			
	CSC	#51			
Japan	Riken	#4			
China	NSC	#1 (20MW)			

Conclusions

The scientific return of next-generation CMB missions will depend on our ability both to access and to exploit the capabilities of the next several generations of challenging HPC architectures.

Planck has been the key vehicle for these (and many other) data analysis efforts for the last decade or more; with its ending we risk losing this critical expertise.

Analysis-induced systematics are a key – but widely neglected – issue for precision CMB measurements.