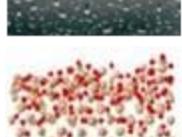
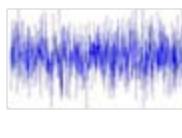


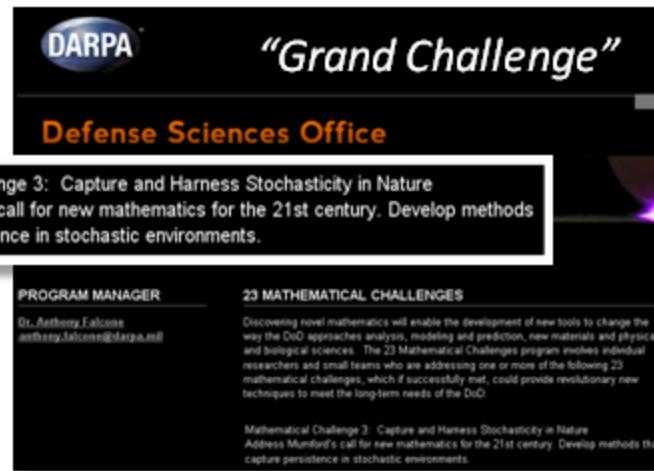
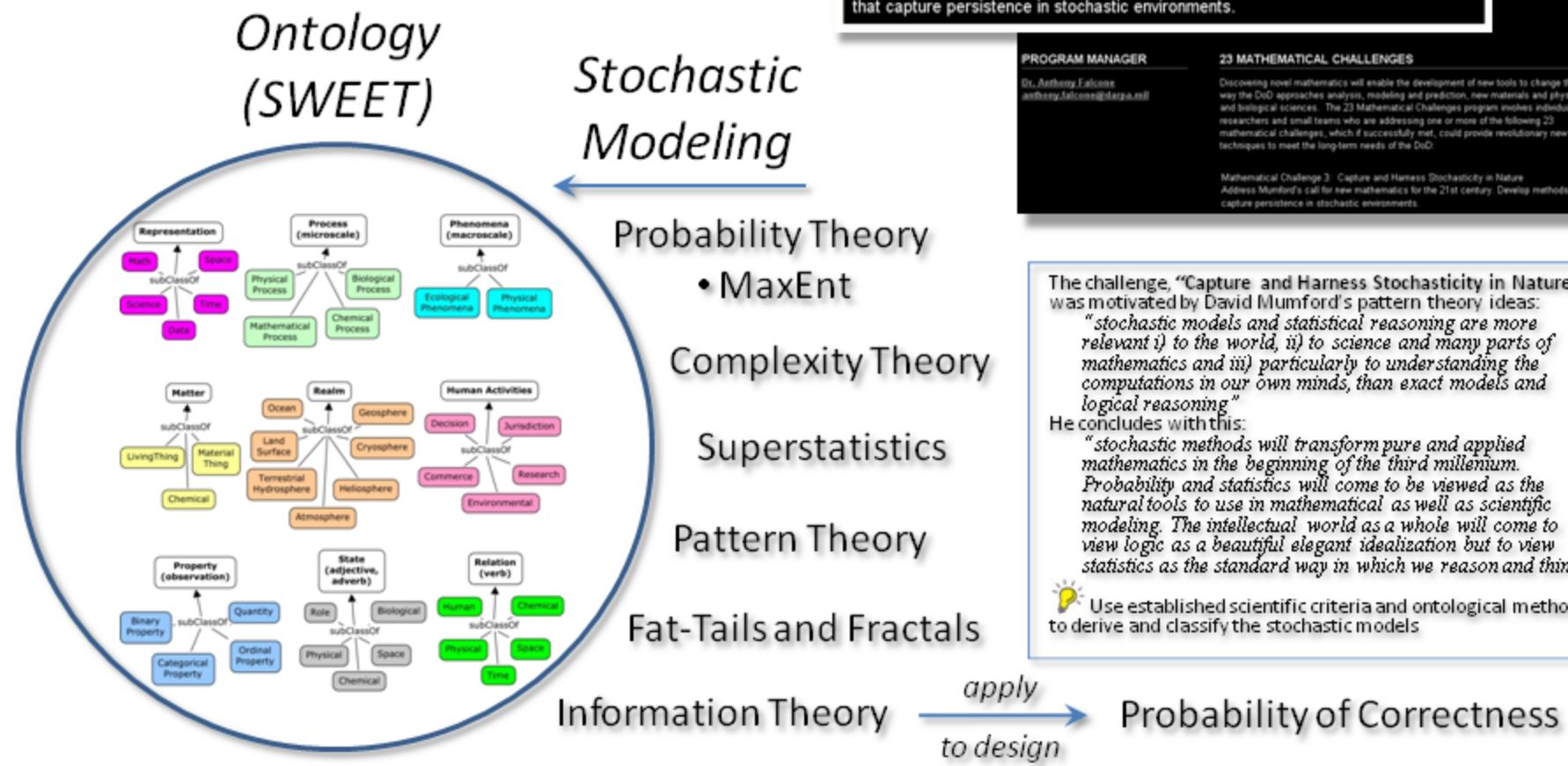
Environmental Context Characterization and Model Approaches

Stochastic Modeling | Ontological Organization and Process Modeling

1.  [Gross Terrain Features](#) - Simple distribution for sampling likelihood of terrain slopes
2.  [Fine Terrain Features](#) - Methods for modeling spatial frequencies for power spectral densities (PSD)
3.  [Wave Energy Statistics](#) - Simple approach for generating PSD's, pulling data from real-time stations
4.  [Wind Energy Statistics](#) - General results for modeling wind energy distribution and autocorrelations for persistence
5.  [Rainfall Statistics](#) - General model for rainfall distributions within storms using composite process
6.  [Corrosion and Oxidation](#) - Model of oxidation and corrosion growth which improves on the Deal-Grove formulation.
7.  [Thermal Dispersion](#) - Simplification of thermal diffusion model to account for disorder and variability in the media.
8.  [Lake Size Statistics](#) - Maximum entropy estimation for lake size distributions for regions
9.  [Particle Size Statistics](#) - General method for modeling size distribution of particulates such as ash and ice crystals
10.  [Clutter Modeling](#) - Maximum entropy models for E-M signals and noise

Insanely Great Idea

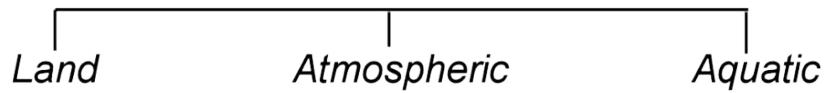
- Applying an *ontological classification* strategy to *stochastic models* of the environment.
 - Simple, efficient while comprehensive
 - A workflow-driven semantic web provides services



The challenge, "Capture and Harness Stochasticity in Nature", was motivated by David Mumford's pattern theory ideas:
"stochastic models and statistical reasoning are more relevant i) to the world, ii) to science and many parts of mathematics and iii) particularly to understanding the computations in our own minds, than exact models and logical reasoning"
He concludes with this:
"stochastic methods will transform pure and applied mathematics in the beginning of the third millennium. Probability and statistics will come to be viewed as the natural tools to use in mathematical as well as scientific modeling. The intellectual world as a whole will come to view logic as a beautiful elegant idealization but to view statistics as the standard way in which we reason and think."
💡 Use established scientific criteria and ontological methods to derive and classify the stochastic models

Ontological Organization and Process Modeling

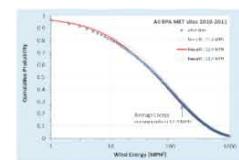
Extracting commonality and providing reusable and scalable extractions



To estimate probabilities of correctness or a PCC to certify a design, the context models must be in the form of continuous or discrete probabilities. For the latter, the discrete probabilities could evaluate to Go/NoGo outcomes over a sampled set of contexts.

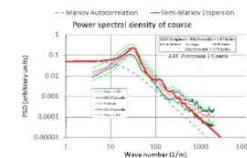
Probability Density Functions PDF

Cumulative Density Functions CDF

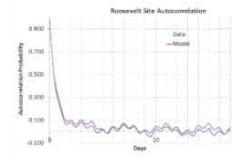


Histograms

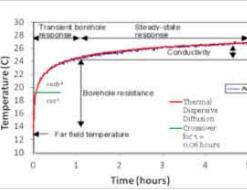
Power Spectral Densities PSD



Autocorrelation Functions ACF



Growth Curves



Data and Lookup Tables

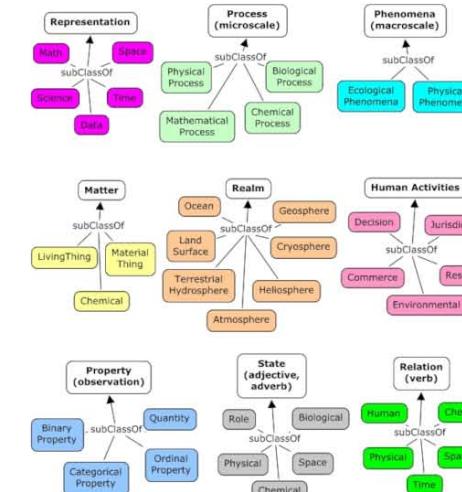
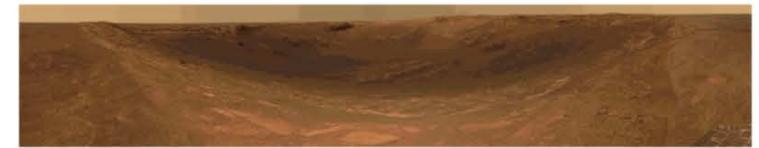
Soil_order	Soil_code	Histology	Awoder	Effective	Waterholding	Hydrologic
y		y	0.437	0.417	4.95	[cm/h]
Sand	S		0.437	0.401	6.13	2.99
Louany Sand	LS		0.463	0.412	11.07	1.02
Sandy Loam	SL		0.463	0.434	9.09	0.34
Loam	L		0.463	0.434	16.58	0.65
Silt Loam	SIL		0.501	0.496	16.58	0.65
Sandy Clay Lo.	SCL		0.398	0.33	21.85	0.15
Clay Loam	CL		0.464	0.393	20.88	0.1
Silty Clay Lo.	SCL		0.471	0.432	27.3	0.1
Sandy Clay	SC		0.43	0.321	23.9	0.06
Silty Clay	SIC		0.479	0.423	29.22	0.05
Clay	C		0.475	0.365	31.63	0.03
Silt	SI		0.466	0.496	16.58	0.65
Finely Drained	A		0.437	0.417	4.95	11.78
Intermediate	B		0.463	0.434	9.09	0.34
Intermediate	C		0.464	0.393	20.88	0.1

Constants and Properties

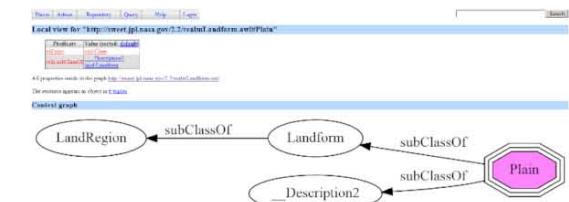
Periodic Table of the Elements	
1 H	2 He
3 Li	4 Be
5 B	6 C
7 N	8 O
9 F	10 Ne
11 Na	12 Mg
13 Al	14 Si
15 P	16 S
17 Cl	18 Ar
19 K	20 Ca
21 Sc	22 Ti
23 V	24 Cr
25 Mn	26 Fe
27 Co	28 Ni
29 Cu	30 Zn
31 Ga	32 Ge
33 Ge	34 As
35 Br	36 Se
37 Kr	38 Xe
39 Rb	40 Sr
41 Y	42 La
43 Lu	44 Ce
45 Pr	46 Nd
47 Pm	48 Sm
49 Eu	50 Gd
51 Tb	52 Dy
53 Ho	54 Er
55 Tm	56 Yb
57 Lu	58 Hf
59 Tl	60 Pb
61 Pb	62 Bi
63 Po	64 At
65 At	66 Rn
67 Fr	68 Ra
69 Ac	70 Th
71 Pa	72 U
73 Np	74 Pu
75 Am	76 Cm
77 Bk	78 Cf
79 Es	80 Fm
81 Md	82 No
83 Lr	84 Rf

Dynamic Models

e.g. compliant contact model



Semantic Web for Earth and Environmental Terminology (SWEET)

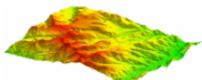




Gross Terrain Features and Topography

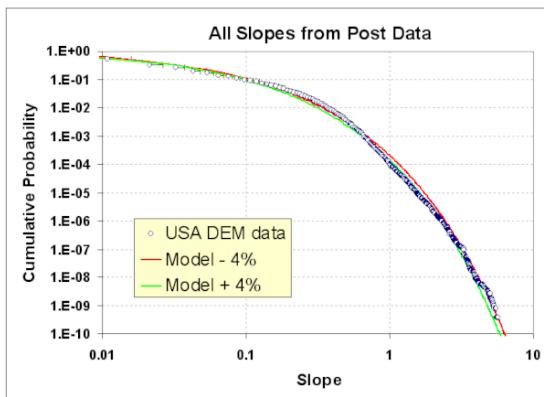
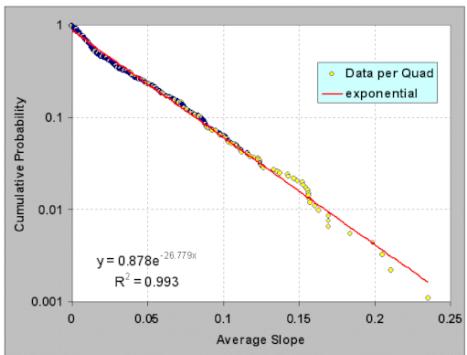
Key : Simple distributions for sampling likelihood of terrain slopes

Generate distributions of terrain slope which match empirical data

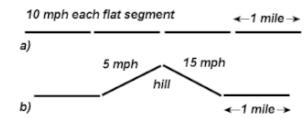
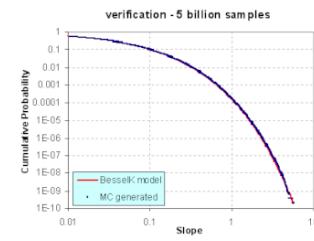


Aggregated regions such as the continental USA (CONUS) map to doubly stochastic distributions
(Modified Bessel function of the 2nd kind)

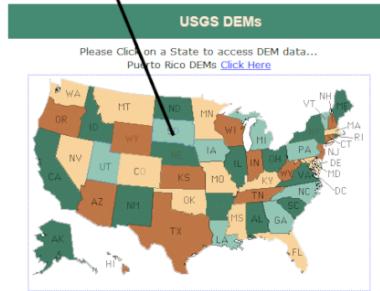
Individual regions map to mean value maximum entropy estimators



Simple two-level sampling generates Bessel distribution



Applications use sampled data for vehicle mobility studies

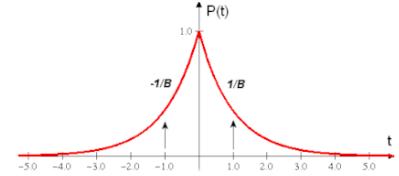


Data from USGS DEM archive.
100 Meter post resolution, converted to average local slope in two dimensions

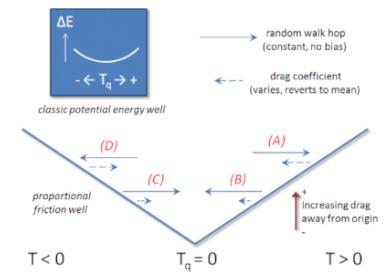
Generate spatial profiles from sampled slopes by random walk generation.



Red noise random walk based on exponential autocorrelation function



Ornstein-Uhlenbeck model provides reversion to the mean statistics



Fine Terrain Features

Key : Methods for modeling spatial frequencies and power spectral densities



Figure 4.0-1. Aerial View of the 2-in Washboard and Radial Washboard Courses

Fine terrain at the vehicle scale is represented by features such as washboard roads, cobblestones, chatter bumps, etc

Terrain is described by its roughness or "bumpiness". Dispersed periodicities in the fine terrain is revealed by power spectral density (PSD) curves. Use the techniques perfected by diffraction spectroscopy to model the damped harmonics in the spatial frequency plane. PSD's provide stimuli to vehicle mobility models.

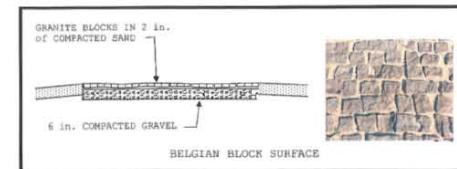


Figure 4.4-1. Belgian Block Surface

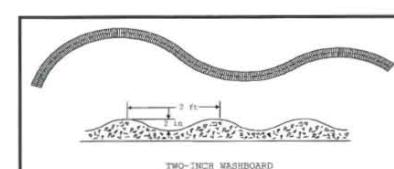


Figure 4.1-1. Two-Inch Washboard

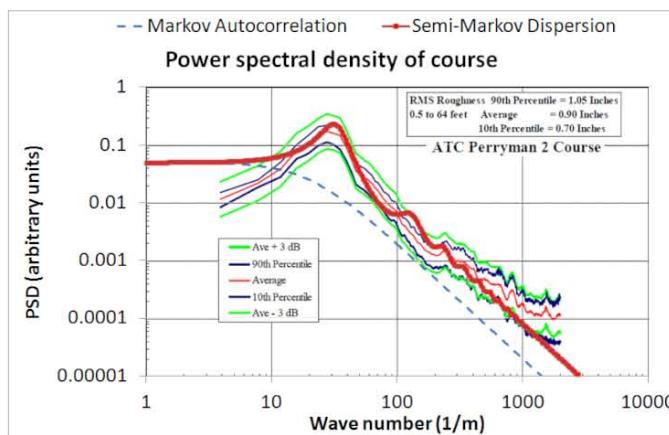


Figure 4.4-2. HMMWV with Trailer on Belgian Block Course



Figure 4.1-2. HMMWV on Two-Inch Washboard

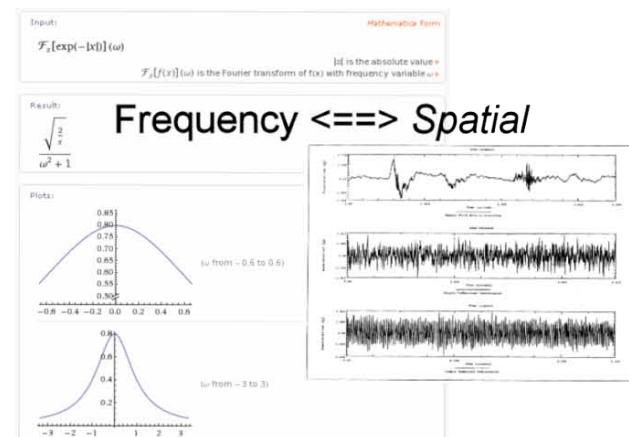
Washboard roads show spatially cyclic features. Dispersion of the waves leads to characteristic damped spectra harmonics as shown below.



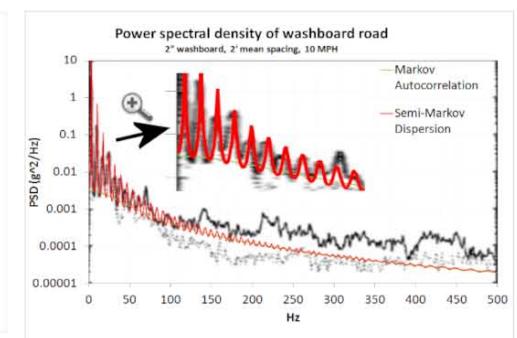
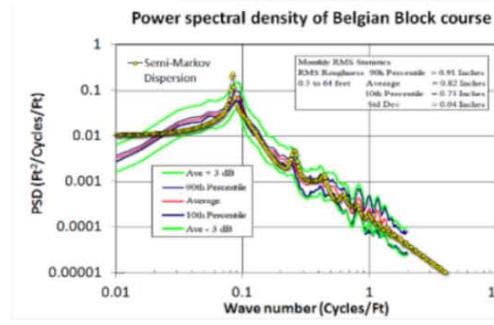
Courses such as Perryman feature range of terrain features with either random or slight spatial periodicities.

Markov or semi-markov models of the terrain roughness are characterized first as autocorrelation functions. A Fourier transform of the autocorrelation generates the corresponding PSD.

Cobblestones such as the Belgian Block course show order in surface roughness.



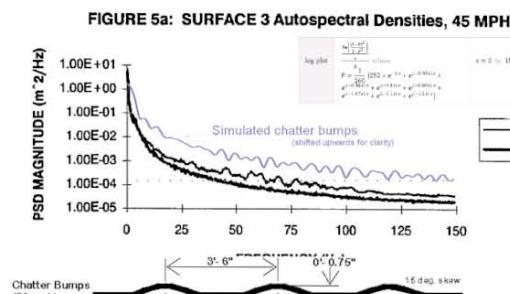
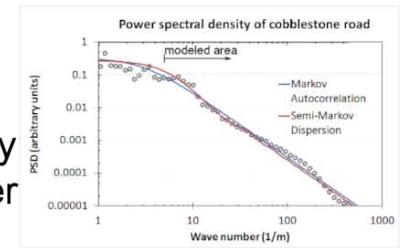
An autocorrelation of a rough surface generates a damped exponential and a Cauchy PSD in the spatial frequency domain. Other periodic fluctuations generate harmonics.



stronger

weaker

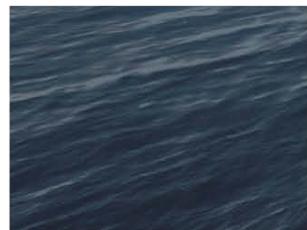
Other cobbles may show weaker order



Simulated chatter bumps in the fine spectra. Only right track contains chatter bumps

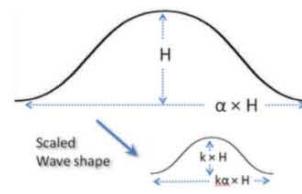
■ Wave Energy Statistics

Key : Simple approach for generating PSD's, extracting data from real-time stations



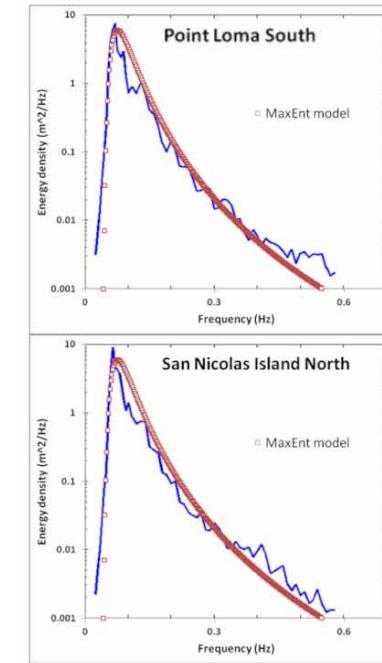
Simulated waves from a commercial rendering package

How are waves characterized and modeled?



A single wave contains potential energy generated by wind and tidal forces. This energy is released as the wave is dissipated. This leads to the power-law shape shown to the right as a power spectral density.

Typical data is gathered over a time frame to gather sufficient statistics.



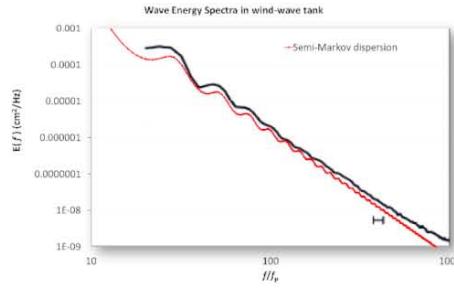
PSD data collected at locations spatially separated but sharing the same weather patterns will show similar profiles.



Data gathered from real-time measuring stations is available from gov sites

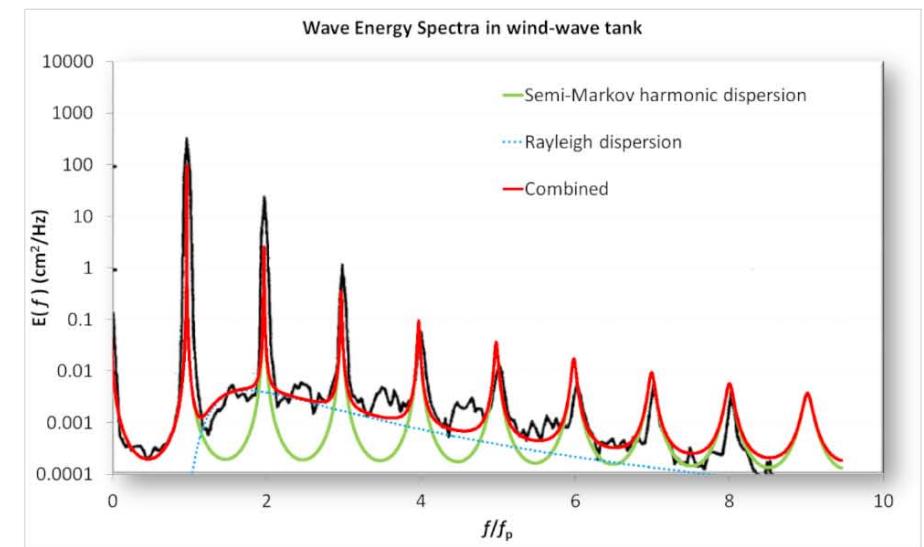


Point Loma South
San Nicolas Island North



If the waves are sufficiently coherent, then we can apply the same spectral techniques as used for fine terrain characterization. Data from wave tank experiments show excellent agreement with dispersive models. The left model shows weaker dispersion than the strong harmonics in the right model.

Spectra showing strong order mixed with background

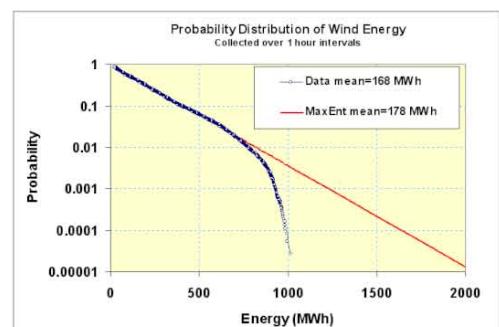
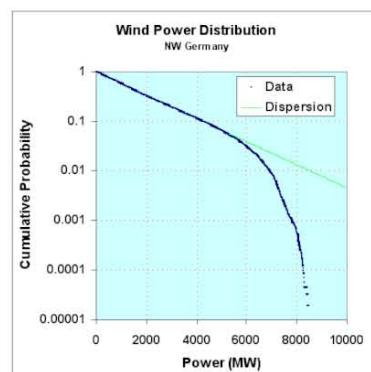


Wind Energy Statistics

Key : Modeling wind energy distributions and autocorrelations for temporal persistence

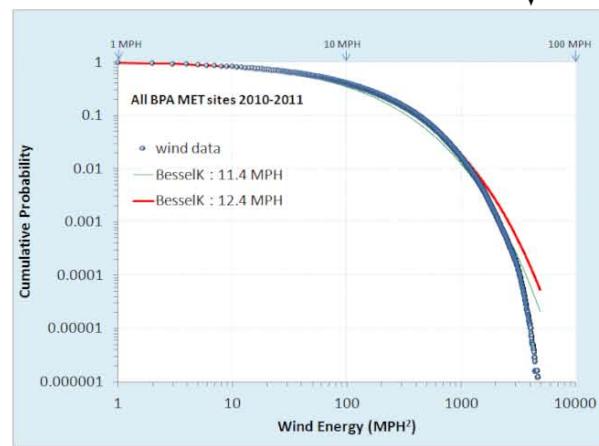
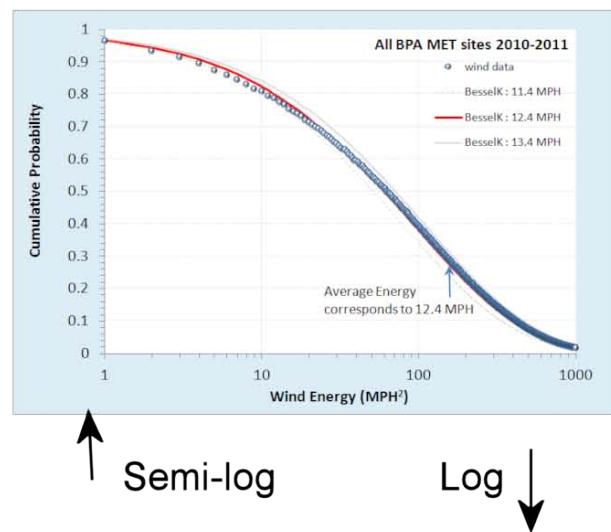
Germany 

Wind statistics available from wind turbine farms



Ontario

An individual station shows a distribution which follows a Max Entropy Rayleigh wind speed profile (subject to a high speed governor cut-off).



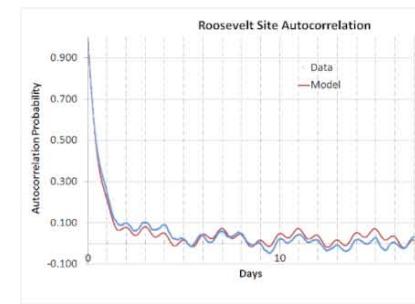
Collection of all wind speeds from the BPA sites reveals a doubly stochastic distribution.

The characteristic function is a Modified Bessel.



Historical Meteorological Readings: 5-Minute Averages, Monthly Mt. Hebo Meteorological Data
All files are in Excel CSV format.

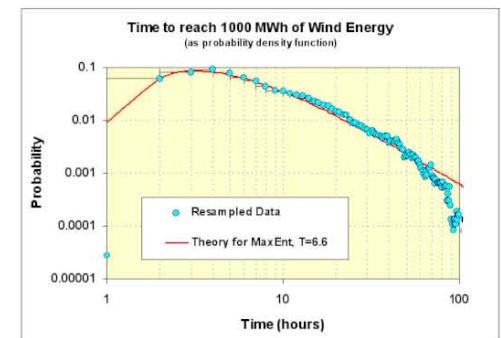
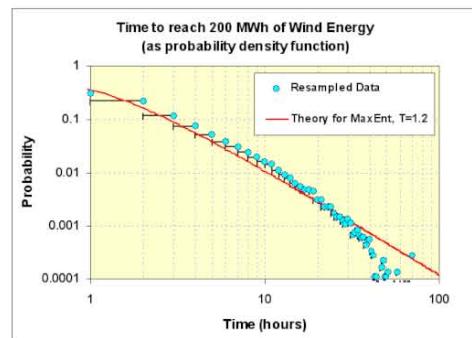
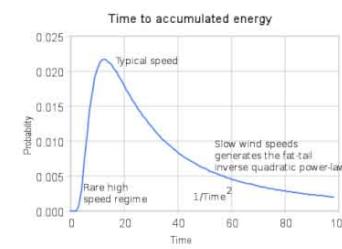
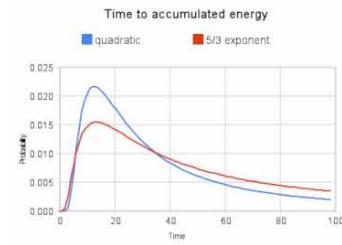
Jan 2010	Feb 2010	Mar 2010	Apr 2010	May 2010	Jun 2010	Jul 2010	Aug 2010	Sep 2010	Oct 2010	Nov 2010	Dec 2010
Jan 2011	Feb 2011	Mar 2011	Apr 2011	May 2011	Jun 2011	Jul 2011	Aug 2011	Sep 2011	Oct 2011	Nov 2011	Dec 2011
Jan 2012	Feb 2012	Mar 2012	Apr 2012	May 2012	Jun 2012	Jul 2012	Aug 2012	Sep 2012	Oct 2012	Nov 2012	Dec 2012



Bonneville Power Authority in the Pacific Northwest provides sets of data from several MET stations

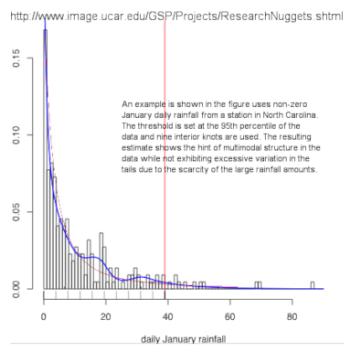
The autocorrelation function of wind speed over a period of time from a single MET station shows a fast damping and then a repeat cycle corresponding to a single day and a weaker signal over several days

Various tests such as may be required for convection dissipation or vehicle drag are simulated by integrating the wind speed or wind energy distributions.

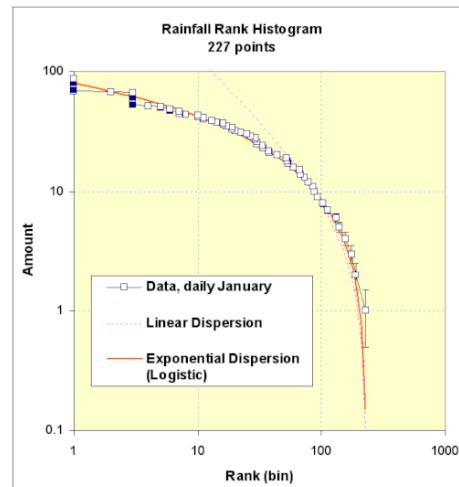


Rainfall Statistics

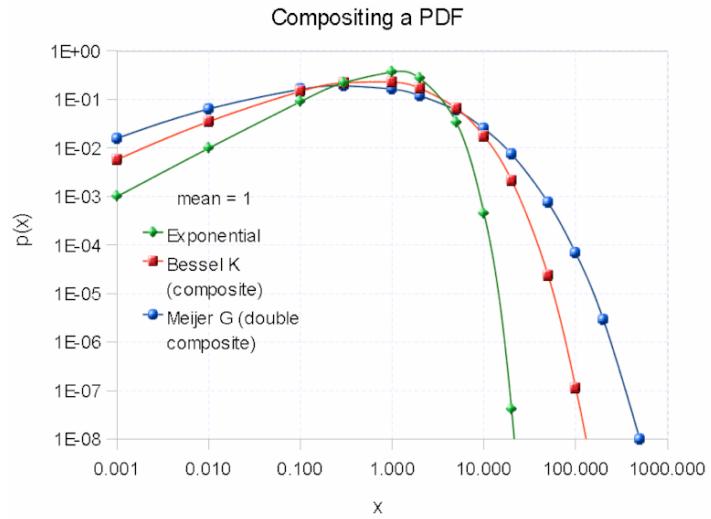
Key : General model for rainfall distributions using a composite process



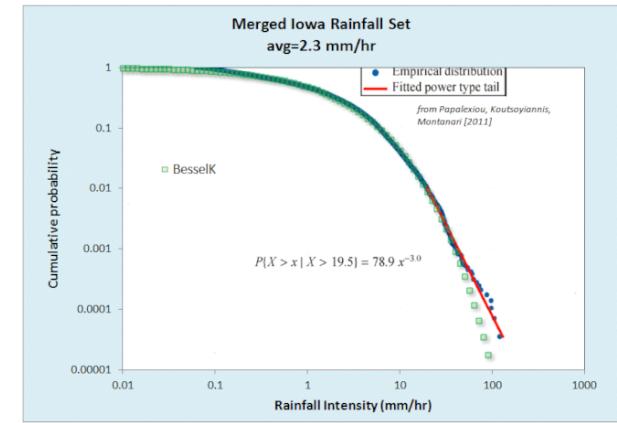
Rainfall shows a wide dynamic range and is considered a multi-scale phenomena. Heavy storms generate high amounts of rainfall within short periods.



A ranked histogram of the rainfall statistics describes the rare events. The data set to the left has a limited amount of entries while the data set below captures high resolution rainfall rates over the course of several events.

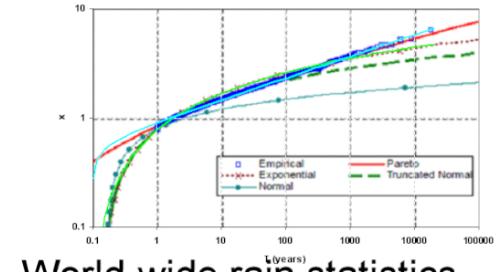
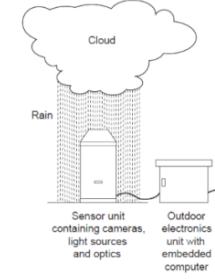
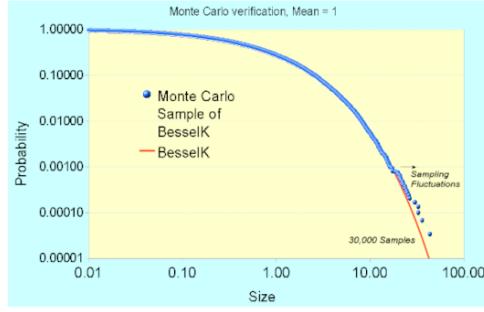


The composite process of gathering statistics from several events leads to a fattening of an originally narrow distribution. The modified Bessel K results from a two-level composite.



The result fitted to the high-res Iowa data shows excellent agreement

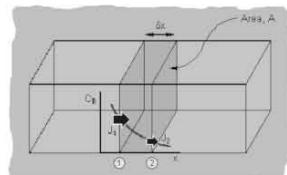
Generating simulation sampling statistic from the Bessel K requires drawing two samples from a maximum entropy exponential distribution and multiplying the results.



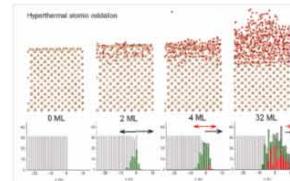
World-wide rain statistics

Corrosion and Oxidation

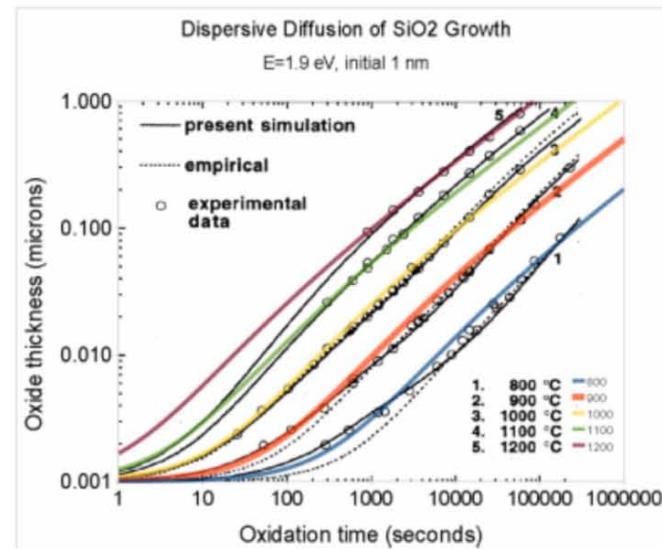
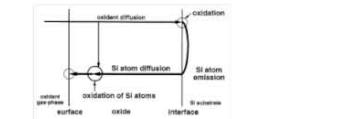
Key : Model of oxidation and corrosion growth which improves on the Deal-Grove formulation



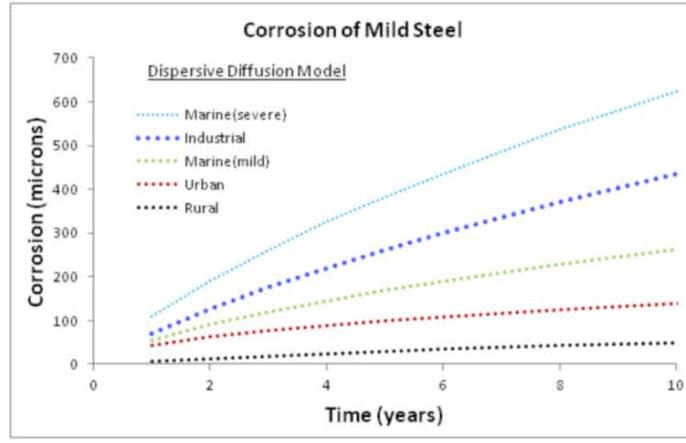
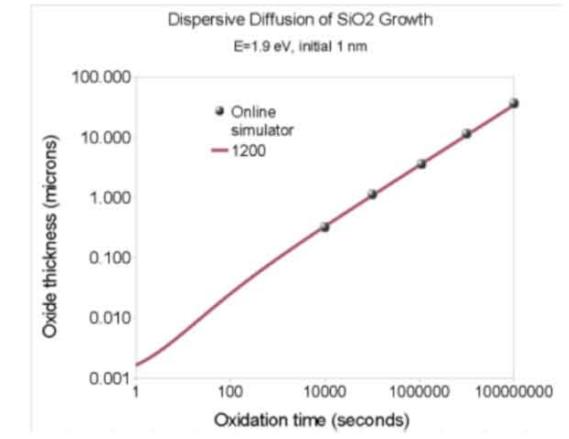
The processes of corrosion and oxidation are driven significantly by diffusional behavior. A gradient in the concentration will drive a diffusional process.



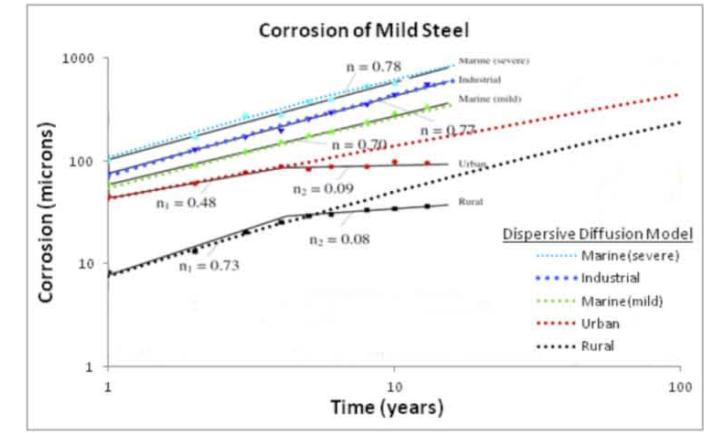
A growing oxide interface will likely show disordered diffusional rates corresponding to the various elemental species involved.



Solving the diffusional Fokker-Planck master equation while assuming a maximum amount of variance in the diffusivities gives excellent agreement with otherwise anomalously predicted growth rates at short time durations. The limit at long times agrees with Fick's law and the conventional Deal-Grove oxidation formulation.



Applying this strategy to general modeling of corrosive environments shows some promising results. The power-law exponent of 0.7 indicates a transition from early linear disordered growth to the square root for long durations.



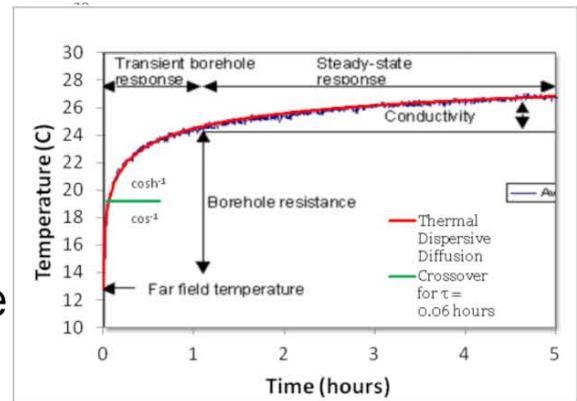
■ Thermal Dispersion

Key : Simplification of thermal diffusion model to account for disorder and variability in the media

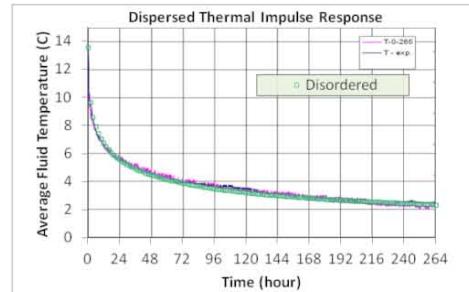
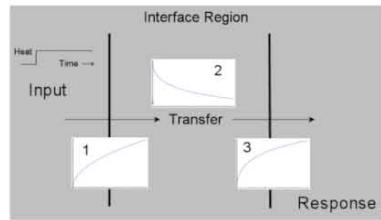


Thermal diffusion typically assumes a homogenous media, yet few environmental contexts will show absolute uniformity

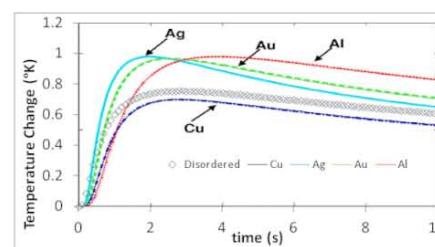
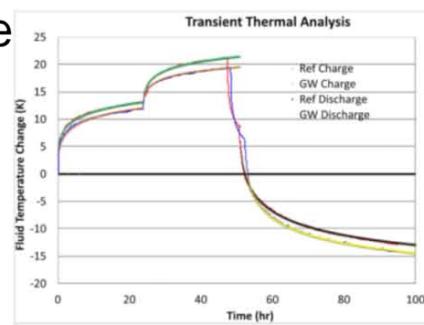
Thermal conduction into a non-uniform earthen heat sink can be modeled well with a dispersive formulation



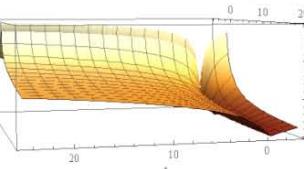
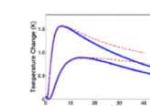
The general recipe assumes a compartment model of heat transfer.



A thermal impulse transfers heat energy across an interface which is partly transferred via non-uniform pathways resulting in a dispersed response.



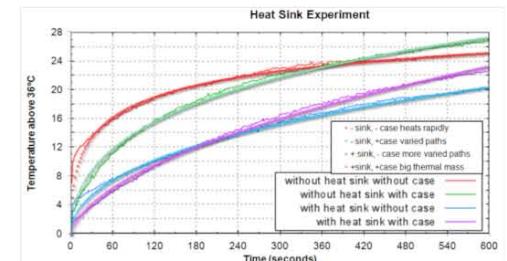
By assuming a coefficient with random variation, a more general thermal dispersion model can be formulated.



scaling
surface

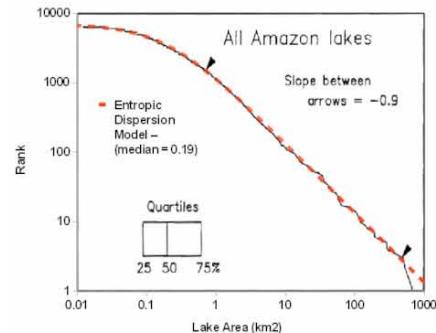
This formulation works well for heat-sinking applications whereby a source of thermal energy is dissipated by an environmental context.

The small-scale version of this application is a CPU heat sink.



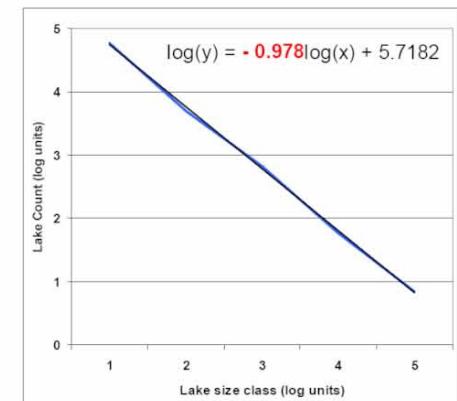
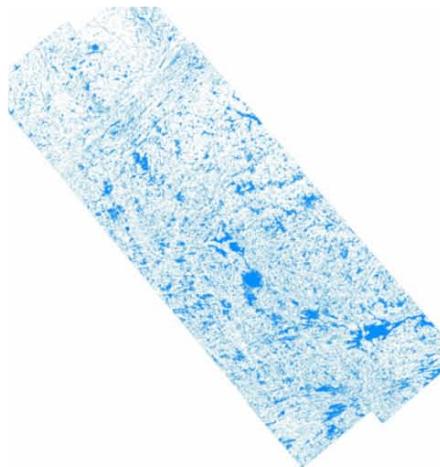
Lake Size Statistics

Key : Maximum entropy estimation for lake size distributions within regions

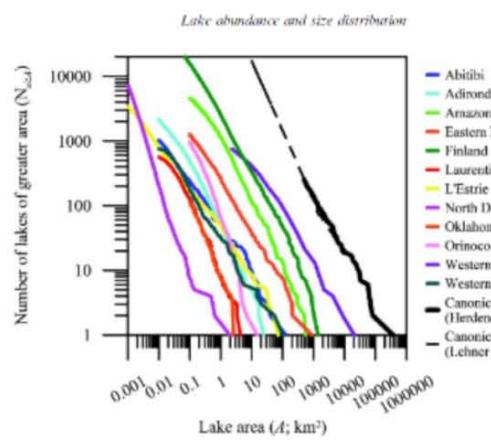


Amazon region

Lake size distributions follow a power-law over several orders of magnitude.

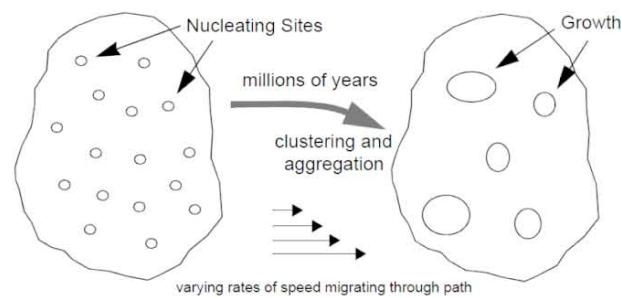


Boundary Waters between USA and Canada



World shows self-similar scaling

Statistics of large-scale environmental obstacles such as lakes and streams are needed for mobility studies. Scaling behaviors provide a concise means of modeling sampling statistics.



Process by which lake sizes aggregate follows a random growth and decline rate scenario. This is a maximum entropy process leading to the characteristic power law observed.

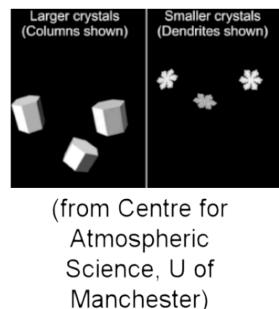
■ Particle Size Statistics

Key : General method for modeling size distribution of particulates such as ash and ice crystals

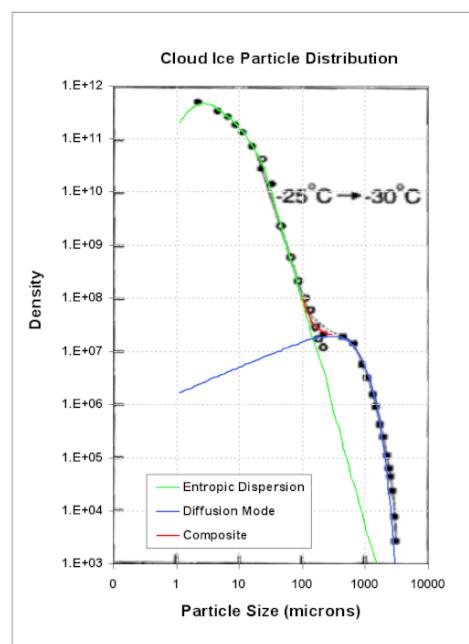
Once characterized, the density and size distribution of airborne particulates follow simple probability density functions (PDF).



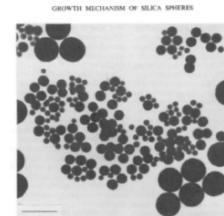
Aerosols are small particles and liquid droplets suspended in air.



(from Centre for Atmospheric Science, U of Manchester)

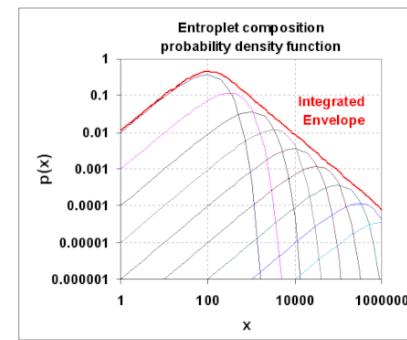


Small particulates outnumber larger particulates according to maximum entropy considerations

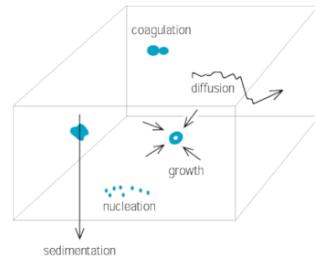


The size distribution of ice crystals can span orders of magnitude in dynamic range. Various nucleation regimes provide superstatistical structure to the empirically measured PDF curves.

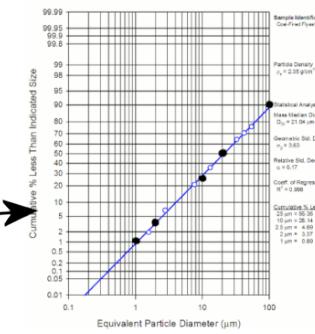
Superstatistical composition envelope



Growth, migration, and nucleate regimes



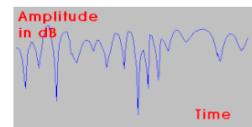
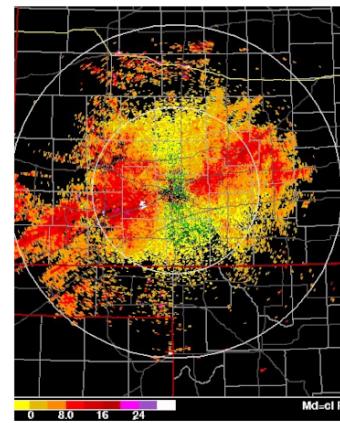
Particulates affect reliability. Particularly damaging to engines are the ash residues left behind volcanic events. Sizes plotted on a log-normal graph.



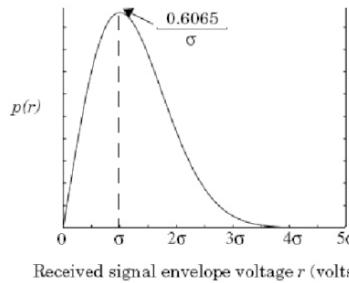
Clutter Modeling

Maximum entropy models for electromagnetic signals and noise

 Vector sum



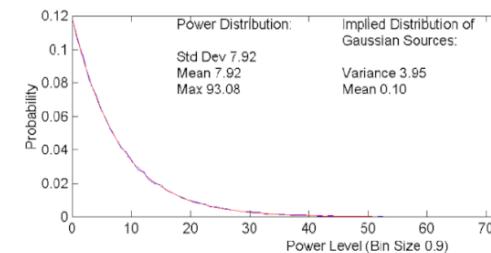
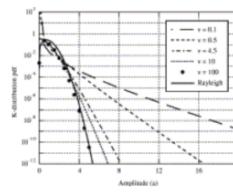
Wave amplitude



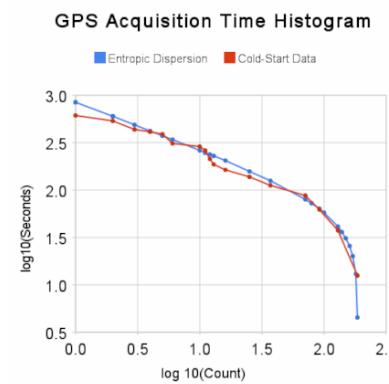
Rayleigh Distribution

In terms of power the spectrum is a maximum entropy damped exponential. This is referred to as noise clutter.

Superposition of Rayleigh noise of different mean frequencies leads to K-Distribution clutter, which is the same distribution to that generated by wind and terrain over a large area.



Maximum
Entropy
Exponential



A practical and very common outcome of noise clutter pertains to "cold start" acquisition times for GPS receivers. The amount of time to lock into satellites depends on the noise level and the quadrature algorithm used to search the satellite doppler frequency spectrum.