

Georgen Lectures . Auke University

October 22-24, 2001

Random Matrix Theory

Lecture 1

It is a great pleasure and honor for me to give the year 2001 Georgen Lectures. I am also very pleased to be here with so many of my friends. My topic is random matrix theory with emphasis on the relationship to integrable systems.

In this the first of 6 lectures, I am going to begin with some

- general remarks about integrable systems.

Then I will

- make some general and historical remarks about random matrix theory (RMT)

- universality conjecture in RMT

In the 2<sup>nd</sup> lecture, I will

- introduce a variety of objects and techniques from modern theory of integrable systems

In the 3<sup>rd</sup> lecture, I will show how

- to use these techniques to solve the universality conjecture following the approach of T. Kriecherbauer, K. McLaughlin, S. Venakides, X. Zhou and P.D.

In the 4<sup>th</sup> and final lecture, I will show how

- to use these techniques to analyze the so-called Ulam problem in the theory of random permutations.

### Integrable systems

The modern theory of integrable systems began in with the discovery of Gardner, Greene, Kruskal and Miura of a method for integrating the Korteweg-de Vries

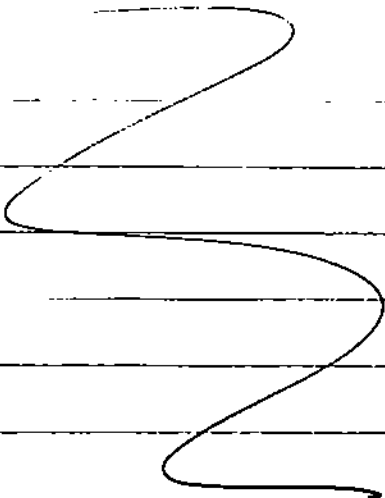
(KdV) equation

$$u_t + uu_x + u_{xxx} = 0, \quad x \in \mathbb{R}, \quad t \geq 0$$

$$u(x, t=0) \rightarrow 0 \text{ as } |x| \rightarrow \infty$$

This equation arises in the theory of water waves

Initially the discovery of Gardner et al was regarded as providing a method of solution for a rather thin set of evolutionary equations, but by the early 1980's it started to become clear that the discovery of Gardner et al was just the first glimpse of a far more general integrable method that would eventually have applications across the broad spectrum of problems in pure and applied mathematics. In the narrowest sense, an integrable system is a



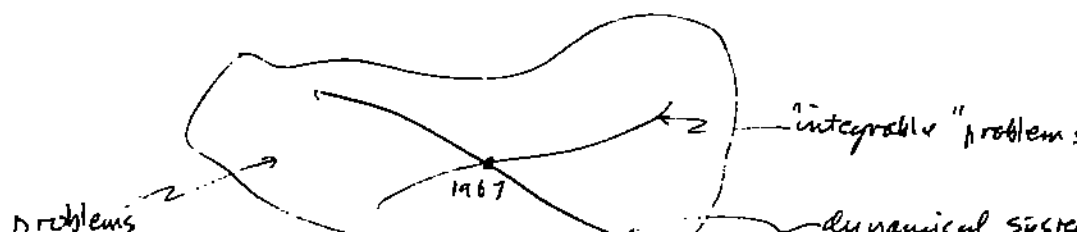
Hamiltonian dynamical system — finite or infinite dimensional — with "enough" integrals of the motion all of whose Poisson brackets are zero, to solve the system in some "explicit" form. Now it has been a rather extraordinary experience in the field over the last 30 years, that many systems which are of great mathematical and physical interest, which may be Hamiltonian, and may not even be dynamical, can be solved "explicitly" using techniques that have a direct link back to the method of solution for the KdV equation discovered by Gardner et al. A kind of developments that I have in mind are for example

- The resolution of the classical Schottky problem in algebraic geometry in terms of the solution of

- the introduction of quantum groups
- integrable statistical models — connection to Jones polynomials
- 2D quantum gravity and the work of Witten and Kontsevich on the KdV hierarchy
- nonlinear special function theory (Painlevé theory)
- conformal field theory — work of Knizhnik, Dotsenko
- \* random matrix theory
- combinatorial problems of Robinson-Schensted-Knuth

The thrust of these lectures is to illustrate how the integrable methods interact with one of these top — random matrix theory.

Allow me a very schematic moment. Consider the following picture



In 1967, these two "manifolds" had a "transverse intersection". The initial thought was that to derive the ideas of Gardner et al. one should move along the "pde" manifold. But this turned out to be too limiting: we now know that in order to see the full development of the method one should move in the "transverse" direction. And this is the direction we will move in the next 2+ lectures.

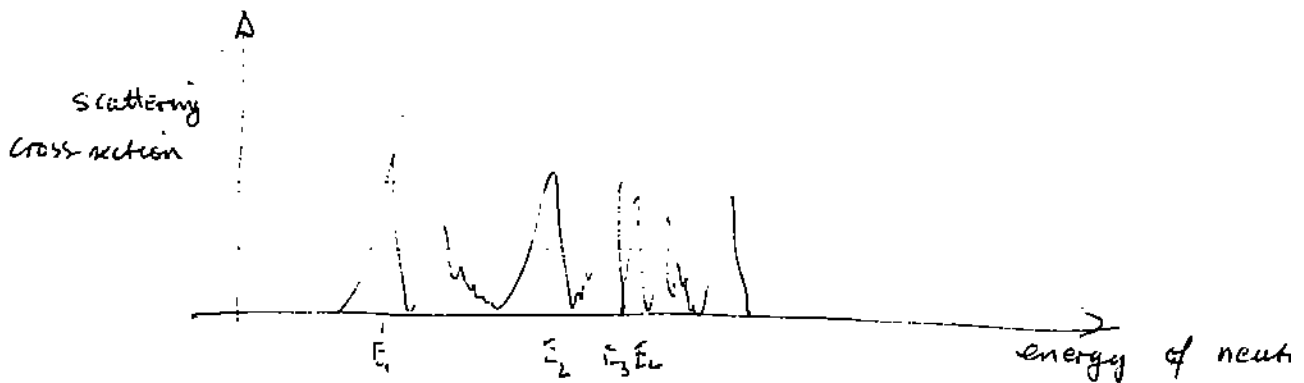
### RMT

Now what is random matrix theory? RMT was first studied in the 1930's in the context of mathematical statistics by Hsu, Wishart and others but it was Wigner in the early 1950's who introduced RMT to mathematical physics. Wigner was concerned with the scattering of neutrons off nuclei.

nuclei



Schematically one sees the following:



The energies  $E_i$  for which one obtains a large scattering cross section are called resonances, and for a heavy nucleus like  $U^{238}$ , say, there can be hundreds of them.

In theory, one could write down non

Schrödinger equation for the neutrons + nucleus and

try to solve it numerically to compute these  $E_i$ .

But in Wigner's time and still in our time.

also into the foreseeable future, this was not a real approach, and so people began to think that it was more appropriate to give the resonances a statistical meaning. But what should the statistical model

At this point Wigner put forward the remarkable hypothesis that the (high) resonances  $E_i$  behave

the ~~the~~ eigenvalues of a (large) matrix. It is to overemphasize what a radical and revolutionary thing

this was: we all recall that when we first were learning some physics, we understood that the detail

of the model were paramount: if you changed the force law in the equations of motion, the behavior

of the system would change. But now all of that

was out the window. The precise mechanism was



no longer important. All that Wigner, and later Dyson, required was that the matrices be Hermitian or real symmetric (so that the eigenvalues were real)

• That the ensemble behaves "naturally" under certain physical symmetry groups ( $N \times N$  Hermitian,  $N \times N$  real symmetric,  $H = J H^T$ )  
GUE, GOE, GSE ensemble  
— more later

During the late 50's, 60's and early 70's, various research (particularly, Bohigas et al) began to test Wigner's hypothesis against real experimental data in a variety of physical situations and the results were pretty impressive. A classic reference for RMT is

• Random matrices — M.L. Mehta, 2nd Ed

and I invite you to look at p17 for a comp

of 1726 nuclear level spacings against predictions of

the GOE ensemble.

procedure: first we must rescale the resonances so

the eigenvalues of the random matrices so that

$$\text{number of resonances / unit interval} =$$

$$\text{expected number of eigenvalues / unit interval}$$

Then we compare the statistics.

More concretely, we typically have a situation where

we are interested in resonances in an interval  $(E, E + \delta E)$

where  $\delta E \ll E$  but  $\delta E$  contains many resonances

We then rescale the resonances in  $\delta E$  so that the

average density is one. We then do a similar

scaling for the eigenvalues in some ensemble, and

only at that point, after we have adjusted

our "microscopes" do we compare statistics. A scientist

approaches such problems with two instruments in

— a microscope and a list of ensembles

- the dials on the microscope are adjusted to account for the macros of the situation, and vary from system to system,
- but once the "slide" is in focus, one sees universal behavior described by one of the entries in the list of ensembles.

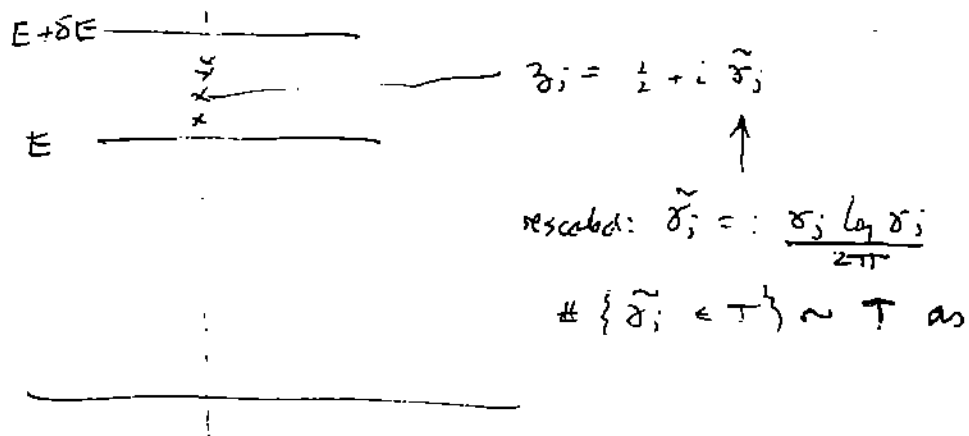
Having given me a schematic moment, I now ask you for a small philosophical moment.

Now it is always something of a mysterious process when begin to think of something that is quite deterministic in a statistical way. Nevertheless, it is a process with which we are very familiar. Take a die, for example, which surely obeys the laws of solid mechanics and aerodynamics, but we readily and intuitively understand as a statistical object. Moreover we "know" what the stochastic model should be: all 6 sides have equal probability. Looked at from this point of view, what Wigner had to do was to see the neutron + nucle

for its stochasticity. In all such problems there is a  
 some singular/asymptotic process <sup>involved</sup> and when we cross some "Bayesian  
 point, phase space opens up, and all bets are off.  
 We are standing, as it were, on the corner of 4<sup>th</sup> & U  
 and 4<sup>th</sup> Street and we are watching this little kid  
 play 3 card Monte: if we are fast enough, we  
 can follow all his moves but then "poof!" the cards  
 are out there, and all 3 <sup>cards</sup> are equally likely.

Now in the early 1970's a very remarkable  
 thing happened. Montgomery, quite independently of the  
 other goings on, began thinking that the zeros of the  
 Riemann zeta function on the critical line  $\text{Re } z = \frac{1}{2}$ ,  
 should also be viewed statistically. And no,  <sup>$\text{Im } z = \frac{1}{2}$</sup>  assuming  
 and retain the imaginary parts of the zeros

(  
 $= \frac{1}{N} \# \{ \text{pairs } (i, j), 1 \leq i \neq j \leq N, \text{ s.t. } a \leq \tilde{\gamma}_i - \tilde{\gamma}_j \leq b \}$   
 for the (rescaled) zeroes in an interval  $(E, E + \delta E) \subset$   
 $E \gg 1,$



and he found a limiting formula  $R = R(a, b)$

$$\left( \int_a^b \left[ 1 - \left( \frac{\sin \pi u}{2\pi u} \right)^2 \right] du \right)$$

for the 2 pt. function as  $E \rightarrow \infty$ ,  $\# \tilde{\gamma}_j \in \delta E \rightarrow \infty$ . W

happened next is very well-known (I even asked Dyson

to authenticate this version). Montgomery met Dyson at

at the Institute in Princeton, and when he told him

about his calculations, Dyson immediately wrote down

a formula and asked Montgomery, "Did you get this

Now he knew the answer, Dyson said "Well, if the zeros  
 of the zeta function behaved like the eigenvalues of  
 a random GUE matrix, this would have to be the  
 answer!" Indeed what Montgomery had obtained  
 for  $R$  was precisely the 2-pt function for the  
 eigenvalues of a (large) random (GUE) matrix.

At this point, the cat was out of the bag.  
 People began asking whether their favorite list of #'s  
 behaved like the eigenvalues of a random matrix.

Through the 80's an extraordinary variety of systems  
 were investigated from this point of view with  
astounding results: for example, if we take a region

in the plane

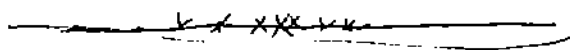


and look at the eigenvalues  $0 \leq \lambda_1 \leq \lambda_2 \leq \dots$  of  
the Dirichlet Laplacian in this region they they too  
(after the standard scaling) behave like the eigenvalues  
of a random GUE matrix (see Mehta p13)

In the late 1980's, Montgomery's work was  
up numerically by Odlyzko, who then confirmed  
Montgomery's work to high accuracy & also investigate  
other statistics such as the nearest neighbor spacing  
again there was incredible agreement with random  
matrix theory. In recent years, the work of Montgomery  
Odlyzko has been a wonderful springboard for  
Sarnak - Rudnick and then Sarnak - Katz, to prove all  
kinds of GUE ( & GSE) random matrix properties  
for the zeros of all kinds of automorphic L functions

Now up until very recently, the physical and mathematical phenomena which were investigated, concerned

the eigenvalues of a random matrix in the bulk of the spectrum.



But in the last 2 yrs or so a very interesting class of problems have started to appear in combinatorics and also in statistical particle models, which concern the eigenvalues at the top of the spectrum. For

example, consider the following version of solitaire called "patience sorting". Suppose we have N cards numbered linearly  $1, 2, \dots, N$  for convenience

Shuffle the cards

1 2 3 . . . N

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Now take the top card  $\pi(1)$  ...

Eventually you end up with  $P_N(\pi)$ . <sup>plies</sup> Question

Putting uniform distribution on the set of

~~permut~~ shuffles  $\{\pi\} = S_N$ , how does  $P_N(\pi)$

behave as  $N \rightarrow \infty$ ? Or naïve, <sup>differently</sup> how

big a table do I need typically to play the game with  $N$  cards?

To state the answer to this question we need  
introduce some notation

(which we will discuss in the 4<sup>th</sup> lecture,

The Theorem is the following (Baik, Johansson, P.D.)

As  $N \rightarrow \infty$ ,  $P_N$ , suitably centered and scaled

behaves statistically like the largest eigenvalue of

a random GUE matrix

More precisely  $\lambda_1^{(N)}(H_N) \approx \lambda_1^{(N)}(H_N) \rightarrow \lambda_1^{(N)}(H_N)$

are the eigenvalues of a (large)  $N \times N$  GUE matrix  
Then

$$\lim_{N \rightarrow \infty} \text{Prob} \left( \frac{P_N - 2\sqrt{N}}{N^{1/6}} \leq t \right) = \lim_{N \rightarrow \infty} \text{Prob} \frac{|\sum_{i=1}^{(N)} b_i| - \nu}{2^{-1/2} N^{-1/6}}$$

$$\equiv F(t)$$

$F(t)$  is called the Tracy - Widom distribution after it  
discovered and can be expressed explicitly in terms of  
a solution of the Painlevé II equation.

There are many equivalent formulations of the  
patience sorting problem - longest increasing subsequence  
of a random permutation, the height of a nucleated  
droplet in a supersaturated medium in the no-called Polynuclear  
Growth (PNG) model, the number of boxes in the first  
row of a Young diagram under Plancherel measure



out that the # of boxes in the 2<sup>nd</sup> row behaves statistically like the 2<sup>nd</sup> largest eigenvalue  $\lambda_{-}^{(N)}(M)$  of a random GUE matrix, and so on.

Now what are these ~~three basic~~ <sup>three basic</sup> distributions mentioned above that ~~were~~ singled out by Wigner and Dyson (on the basis of the behavior of the system under time reversal and change of ~~reference~~ reference frame)?

\* 1) Gaussian Unitary Ensemble (GUE)  
consisting of

(a)  $N \times N$  Hermitian matrices  $M = (M_{ij})$

(b) with probability distribution

$$P(M) dM = P(M) \prod_i^N dM_{ii} \prod_{i < j} d\text{Re} M_{ij} \prod_{i < j} d\text{Im} M_{ij}$$

which is invariant under ~~automorphism~~ <sup>normalization density</sup> unitary conjugation  $U$

$$M \mapsto U^{-1} M U = M', \quad U \text{ unitary}$$

$$\text{i.e. } P(M') dM' = P(M) dM$$

(c)  $\{M_{ii}, \text{Re} M_{ij}, \text{Im} M_{ij}\}$  are independent so

$$P(M) = \prod_i^N \mathcal{I}^{(0)}(M_{ii}) \prod_{i < j} \mathcal{I}^{(1)}(\text{Re} M_{ij}) \prod_{i < j} \mathcal{I}^{(1)}(\text{Im} M_{ij})$$

2) Gaussian Orthogonal ensemble (GOE)

(a)  $N \times N$  real symmetric

3) Gaussian Symplectic ensemble (GSE)

(a)  $2N \times 2N$  Hermitian self dual  $M = M^u$

$$M = J M^T J^{-1}$$

$$J = \begin{pmatrix} 0 & I \\ -I & 0 \end{pmatrix}$$

(b) . . .

invariant under  $M \mapsto U^{-1} M U$ ,  $U$  unit symplectic

$$U^T J U = J$$

(c) . . .

The analysis of GUE is the simplest and I am going to

restrict myself to this case ~~in~~ throughout these lectures

So focusing on GUE, the first Theorem in the lecture is that if  $P(M)$  satisfies (a) (b) (c) then necessarily

$$P(M) = \text{const } e^{-\alpha(M^2 + \beta M + \gamma)}$$

where  $\alpha > 0, \beta, \gamma \in \mathbb{R}$ . Centering and rescaling we have the

GUE distribution  
↑  
Gaussian

$$(2) \quad P(M) = \frac{1}{Z_N} e^{-\alpha M^2} dM$$

Now here comes the problem: whereas conditions (a) & (b) are physical, (c) is just a device and has no physical basis.

If we just assume (a) and (b) we find

that

$$(3) \quad P(M) dM = \frac{1}{Z_N} e^{-\text{tr} V(M)} dM$$

for some real valued function  $V$ ,  $V(x) \rightarrow \infty$  as  $|x| \rightarrow \infty$

are called Unitary Ensembles

Ensembles of matrices of type (3). Now physicists turn

this all on its head and say "as  $\neq$  there is no physical

way to distinguish between different choices of  $V$ , then

whatever answers we compute, the answer must be independent

of  $V$ ." This is a rough form of what is meant by

universality in random matrix theory. In the

coming lectures we will focus on a particular basic statistical

The probability  $P(a,b)$  that a (large) matrix has no eigen in an interval  $(a,b)$ , and we will show that

$P(a,b)$  indeed has a universal form independent of  $V$ . And

in proving this, we will employ techniques that have

arisen in integrable theory over the years and can

be traced back in some form to the original method

of solution for the KdV equation discovered in 1967 by

Gardner, Greene, Kruskal and Miura.

To summarize, what has emerged, somewhat mysteriously and to everyone's surprise is a very powerful

heuristic: all kinds of things physical and mathematical,

believe, in some limit, like random matrices. Next time

you have a system that you want to approximate in such a way model as a large matrix, so for it. You have a good chance of be