

# A mean-field model for the electron glass

Yoseph Imry , work with  
Ariel Amir and Yuval Oreg

Motivated and influenced by experimental results of  
Zvi Ovadyahu et al, HU.

מכון ויצמן למדע  
WEIZMANN INSTITUTE OF SCIENCE



# Common features of glassy models

- Randomness, frustration.
- Many states close to the ground state.
- Aging and memory effects: relaxation is slower if perturbation lasts longer.

## Physical examples

- Ordinary glasses.
- Various magnetic materials.
- Packing of hard spheres.
- **Electron glass! (long-ranged interactions, but not infinite!)**

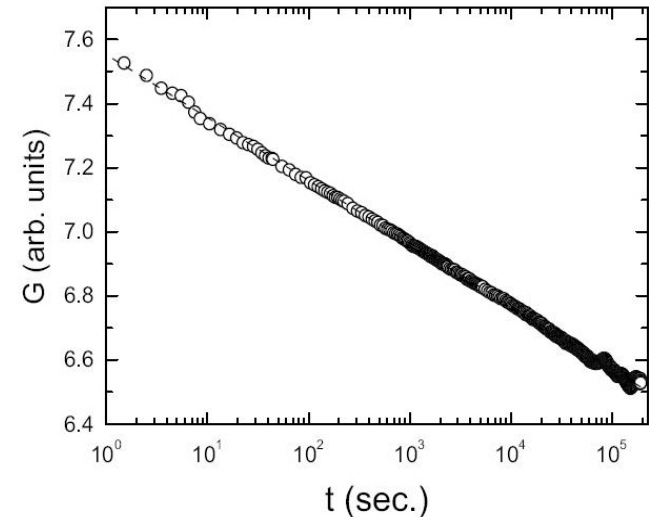
**OUR PROBLEM HERE!**

**General questions: structure of states, out-of-equilibrium dynamics?**

## Experimental motivation

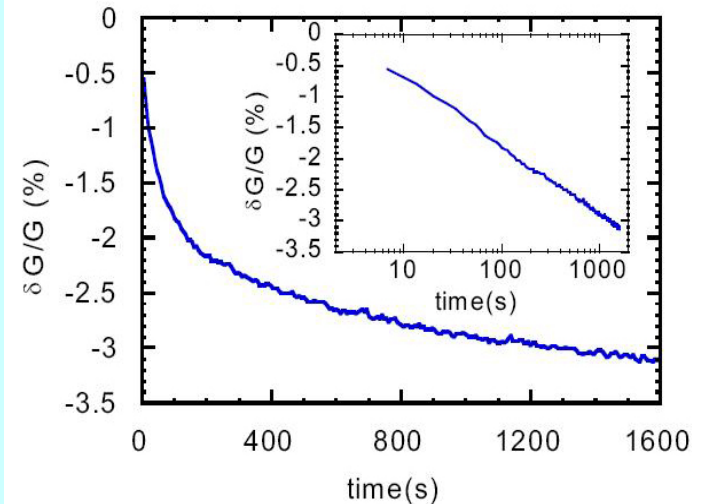
- Disordered samples, excited out-of-equilibrium.
- Measured logarithmic decay of physical observable (conductance, capacitance) from times of a few seconds to a day!
- Similar results for other systems, such as granular Al.

**What are the ingredients leading to slow relaxations?**



Z. Ovadyahu et al. (2003)

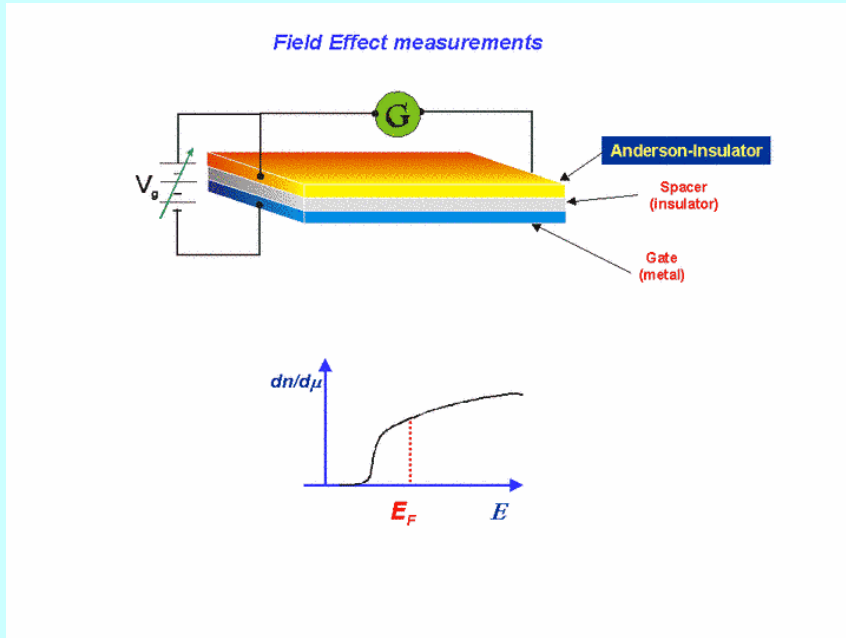
InO samples



T. Grenet (2004)

Granular Al samples

# Electron Glass System



Courtesy of Z. Ovadyahu

- Disordered InO samples
- High carrier densities  $\sim 10^{19-20}/\text{cc}$ , Coulomb interactions may be important.

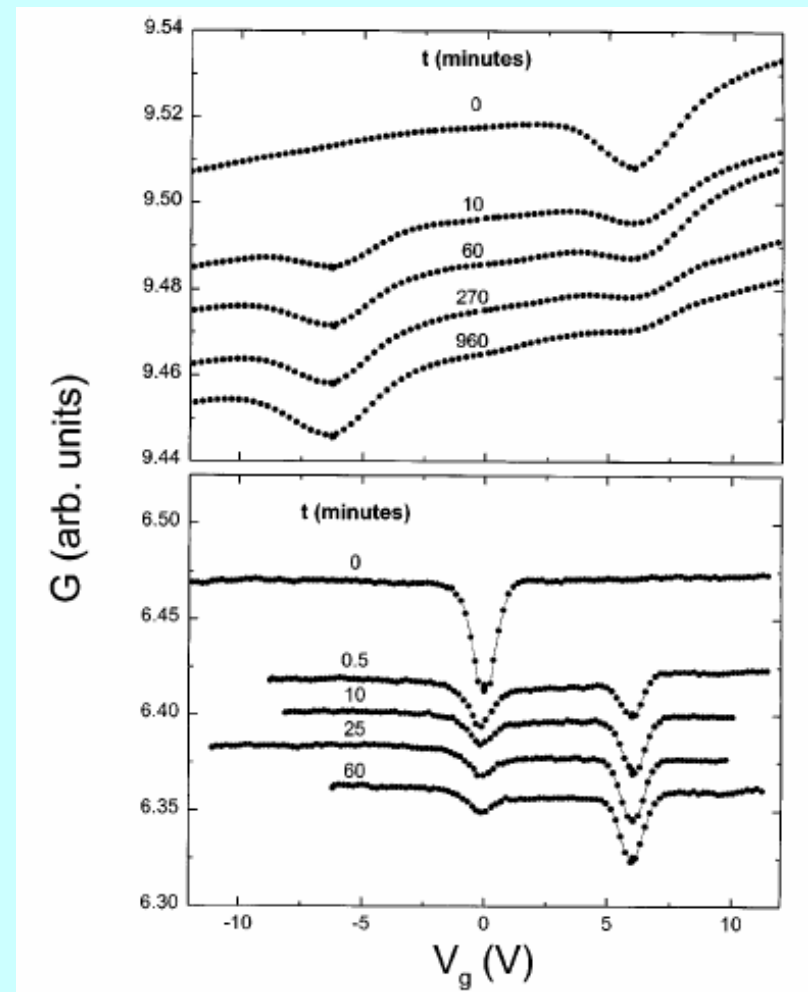


FIG. 1. Conductance versus gate voltage for a TDE of two samples with  $n$  of  $4 \times 10^{20}$  and  $1 \times 10^{20} \text{ cm}^{-3}$  (upper and lower graphs, respectively). In each graph, the first trace was taken  $\approx 12$  hours after the initial cooldown with  $V_g^0$  imposed (6 and 0 V for the upper and lower graphs, respectively). Then,  $V_g^n$  was applied and maintained between subsequent sweeps ( $-6$  and  $6$  V, respectively). The other traces (shifted for clarity) are labeled by the time elapsed since  $V_g^n$  was first applied. Typical scan rate was 0.1–0.2 V/s.

Vaknin, Ovadyahu and Pollak, 1998

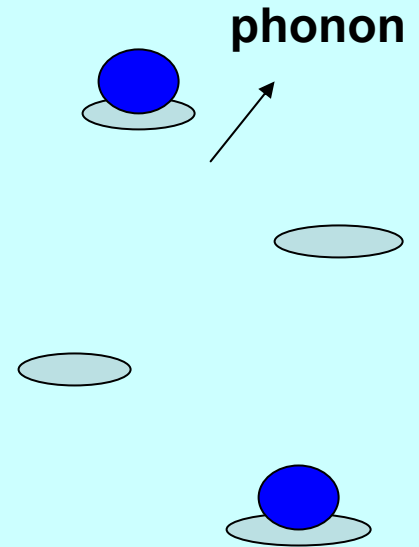
## The model

- Strong localization due to disorder  
→ randomly positioned sites, on-site disorder.
- Coulomb interactions are important (e.g.: Efros-Shklovskii)
- Phonons induce transitions between configurations.
- Interference (quantum) effects neglected.

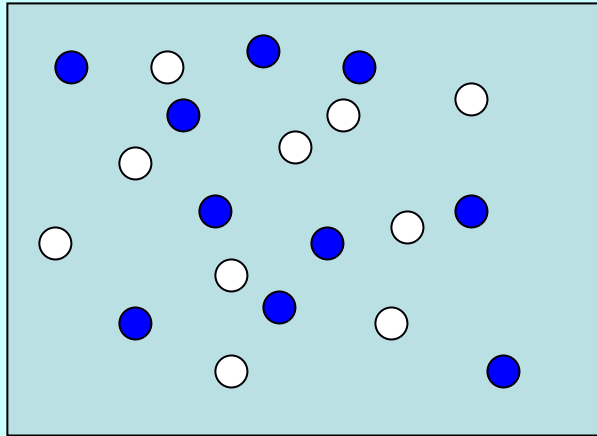
e.g:

**Ovadyahu and Pollak (PRB, 2003)**

**M. Muller and S. Pankov (PRB, 2007)**



## Efros-Shklovskii argument , T=0



● = occupied site

○ = unoccupied site

Cost of moving an electron:  $\Delta E = E_i - E_j - \frac{e^2}{r_{ij}}$

For ground state:  $\Delta E \geq 0$

Assume finite density of states at  $E_f \longrightarrow$  Contradiction.

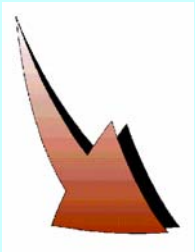
$\longrightarrow$  Upper bound is  $g(E) \leq \alpha |E|^{d-1}$

## Coulomb gap- estimation of scales

- For a 2D systems where the structural disorder is more important:

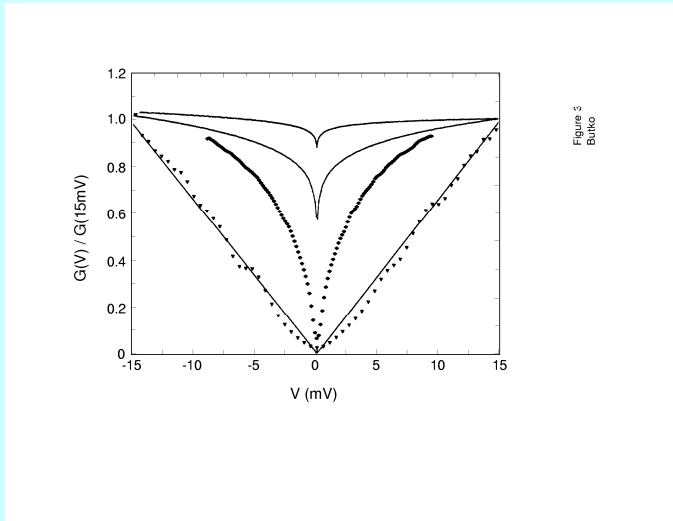
$$T_{washout} \sim \frac{e^2}{r_{nn}}$$

In a typical experiment  $W \sim 10$  nm,  $n \sim 10^{20} \text{cm}^{-3}$ ,  $T \sim 4\text{K}$

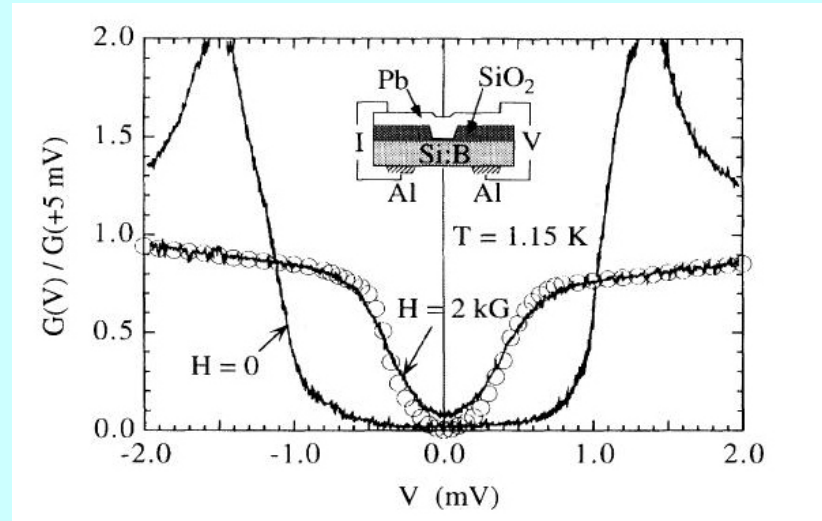


- The Coulomb gap should be observable.
- The system is close to 2D.

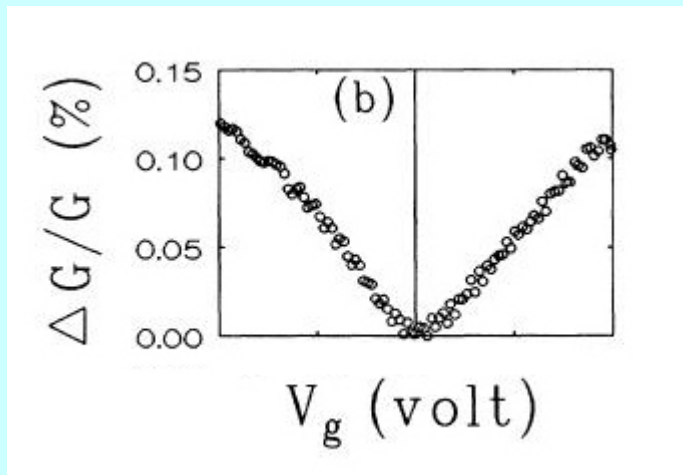
# Experimental manifestation



**2D:** Butko et al. (PRL 84, 2000)



**3D:** Massey et al. (PRL, 1995)



Ben Chorin et al. (PRB, 1992)



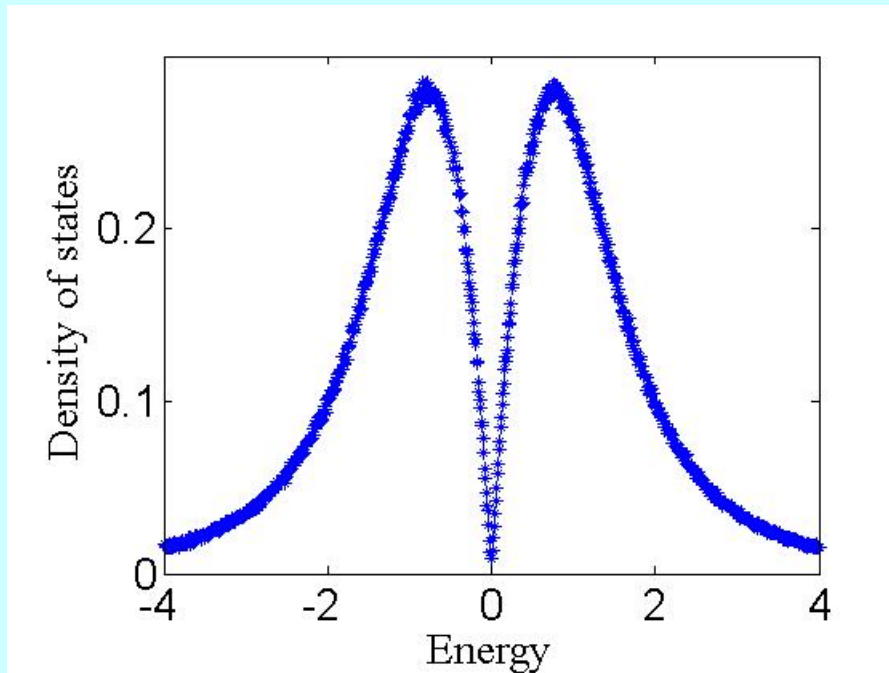


## Mean-field approximation - Equilibrium

- Detailed balance leads to Fermi-Dirac statistics ( $f_j$ ).
- Self-consistent set of equations for the energies:

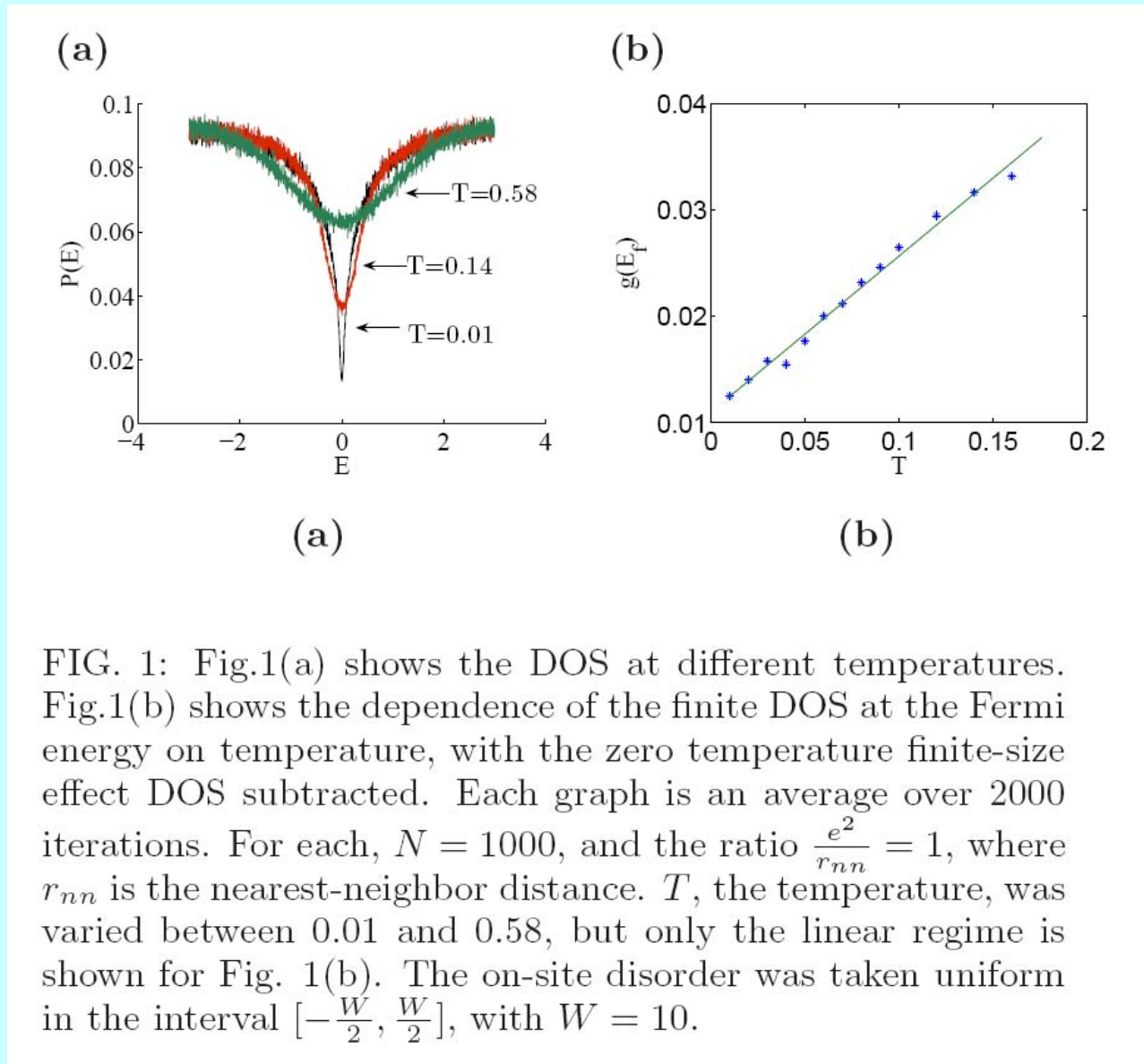
$$E_i = \varepsilon_i - \sum_{j \neq i} \frac{e^2 f_j}{r_{ij}}$$

Many solutions (valleys). Kogan, *PRB* 57, 9736, 1998  
– exp increase with N, in a numerical study.



Produces all features of Coulomb gap, incl temperature dependence (figure shown is for half filling).

# Temp dependence of Coulomb gap, mean field



## Mean-field approximation - Dynamics

$$n_i \rightarrow \langle n_i \rangle, \frac{dn_i}{dt} = \sum_j (-\gamma_{i,j} + \gamma_{j,i})$$

$$\gamma_{i,j} \sim n_i(1 - n_j)[N(|\Delta E|) + \theta(\Delta E)]$$

$\Delta E$  contains all the interaction information

We saw: Mean-field works well for statics!

- The dynamics will make the system ‘dig’ its Coulomb gap, eventually.
- Many locally stable points (‘glassy’)

## Solution near locally stable point

Close enough to the equilibrium (locally) stable point,  
one can linearize the equations, leading to the equation:

$$\frac{d\vec{\delta n}}{dt} = A \cdot \vec{\delta n}$$

$$A_{i,j} = \frac{\gamma_{i,j}^0}{n_j^0(1-n_j^0)} - \frac{e^2}{T} \sum_{l \neq i,j} \gamma_{i,k}^0 \left( \frac{1}{r_{i,j}} - \frac{1}{r_{i,k}} \right)$$

$(i \neq j)$

**Negligible (checked numerically),  
when  $T \gg 0$ .**

The sum of every column vanishes (particle number conservation)

Off diagonals are positive

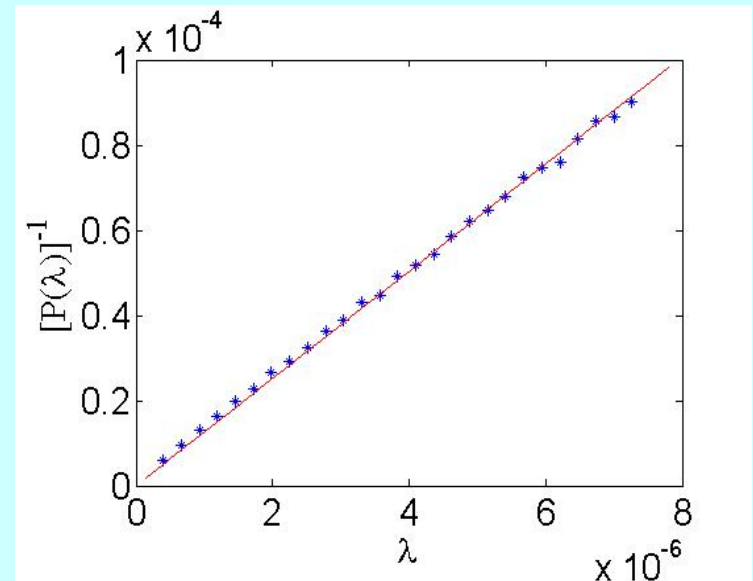
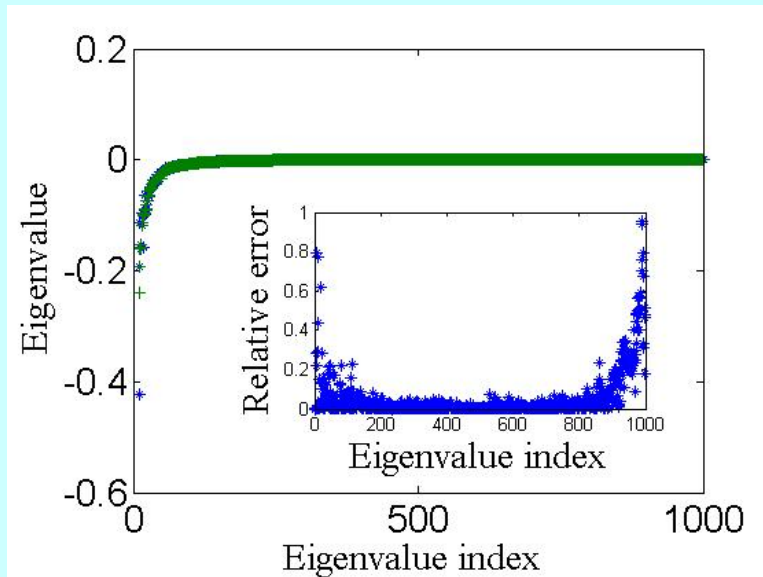
→ all  $\text{Re}(\text{eigenvalues})$  are negative (Stability!)



## Solution for dynamics

The eigenvalue distribution will determine the relaxation.

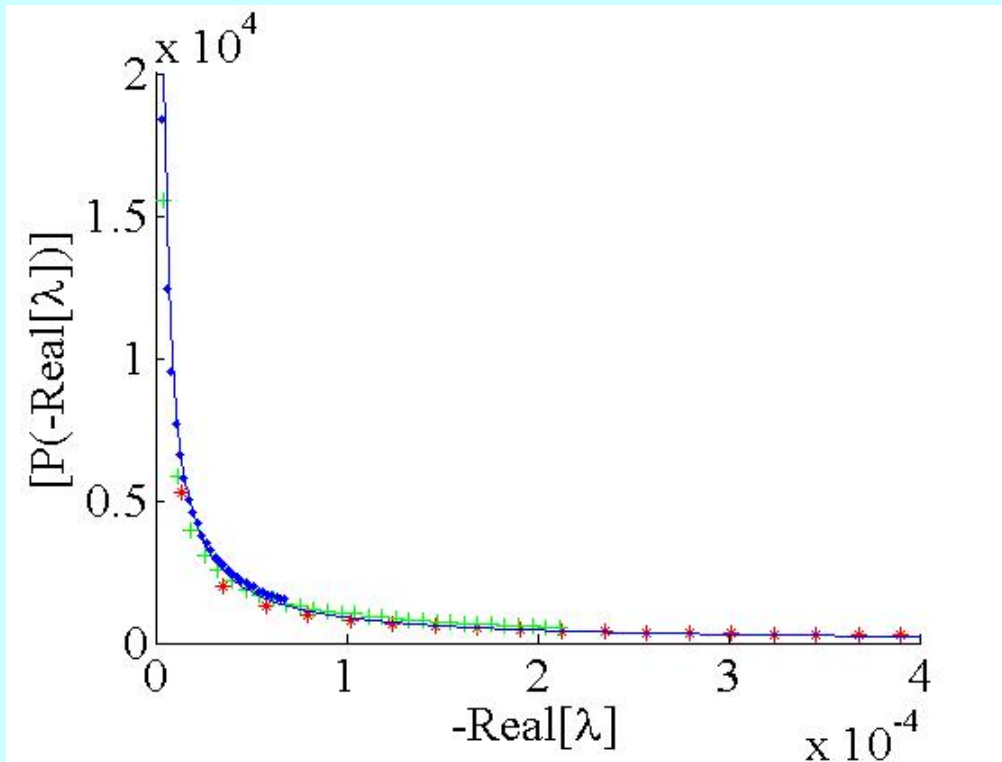
Solving numerically shows a distribution proportional to  $1/\lambda$  :



Implication on dynamics:

$$\delta n(t) = \int P(\lambda) e^{-\lambda t} d\lambda \sim -A \log(t)$$

## Effect of Interactions



- Slow relaxations occur also without interactions.
- Interactions push the distribution to (s)lower values.

# Distance matrices- definition

( cf. M. Mezard, G. Parisi, and A. Zee, Nucl. Phys. B 3, 689, 1999).

- Off diagonals:

$$A_{i,j} = e^{-\frac{r_{ij}}{\xi}}$$

(Euclidean distances in D dimensions)

- Sum of every column vanishes

**GREAT CIRCLE DISTANCES, IN MILES,  
BETWEEN TEXAS AND SELECTED NORTH AMERICAN CITIES**

City	Geographic coordinates		Distances to Texas cities																
	Latitude	Longitude	10000	8000	6000	4000	2000	1000	500	250	125	62.5	31.25	15.625	7.8125	3.90625	1.953125		
Atlanta	33.45	84.39	861	888	916	952	997	1052	1117	1192	1276	1369	1470	1580	1698	1834	2001	2199	
Baltimore	39.30	76.29	1371	1436	1504	1575	1651	1732	1818	1909	2004	2104	2208	2316	2428	2544	2664	2791	
Birmingham	33.51	86.58	1372	1402	1436	1474	1515	1559	1607	1658	1712	1770	1831	1895	1961	2030	2101	2175	
Boston	42.35	71.05	1379	1450	1524	1601	1681	1764	1850	1939	2030	2124	2220	2318	2418	2520	2624	2731	
Buffalo	42.84	78.47	1400	1472	1547	1624	1704	1787	1873	1961	2051	2143	2237	2332	2428	2526	2626	2729	
Charlotte	35.22	80.84	1380	1430	1483	1538	1595	1655	1717	1781	1847	1914	1983	2052	2122	2193	2264	2336	
Chicago	41.88	87.63	889	908	929	953	981	1011	1043	1077	1113	1151	1191	1232	1274	1318	1363	1409	
Cincinnati	39.10	84.50	969	988	1009	1033	1060	1089	1120	1153	1188	1224	1261	1299	1338	1378	1419	1460	
Cleveland	41.50	81.50	1132	1160	1190	1221	1254	1289	1325	1362	1400	1439	1478	1518	1558	1598	1639	1680	
Denver	39.74	104.99	664	688	715	744	776	811	848	888	929	971	1014	1058	1103	1149	1195	1242	
Detroit	42.33	83.0	1142	1173	1206	1241	1278	1316	1355	1394	1434	1474	1514	1554	1594	1634	1674	1714	
Houston	29.76	95.36	1277	1288	1299	1310	1321	1332	1343	1353	1363	1373	1383	1393	1403	1413	1423	1433	
Indianapolis	39.83	86.16	991	1011	1032	1054	1077	1101	1126	1152	1178	1204	1230	1257	1284	1311	1338	1365	
Jacksonville	30.53	81.52	1275	1316	1359	1404	1451	1499	1548	1598	1648	1698	1748	1798	1848	1898	1948	1998	
Kansas City	39.10	94.58	1071	1094	1119	1145	1172	1200	1228	1257	1286	1315	1344	1373	1402	1431	1460	1489	
Los Angeles	34.02	118.24	1672	1707	1744	1783	1824	1865	1908	1952	1997	2042	2088	2134	2180	2226	2272	2318	
Los Angeles	34.02	118.24	1672	1707	1744	1783	1824	1865	1908	1952	1997	2042	2088	2134	2180	2226	2272	2318	
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Los Angeles	34.02	118.24	1672	1707	1744	1783	1824	1865	1908	1952	1997	2042	2088	2134	2180	2226	2272	2318	
Los Angeles	34.02	118.24	1672	1707	1744	1783	1824	1865	1908	1952	1997	2042	2088	2134	2180	2226	2272	2318	

What is the eigenvalue distribution?

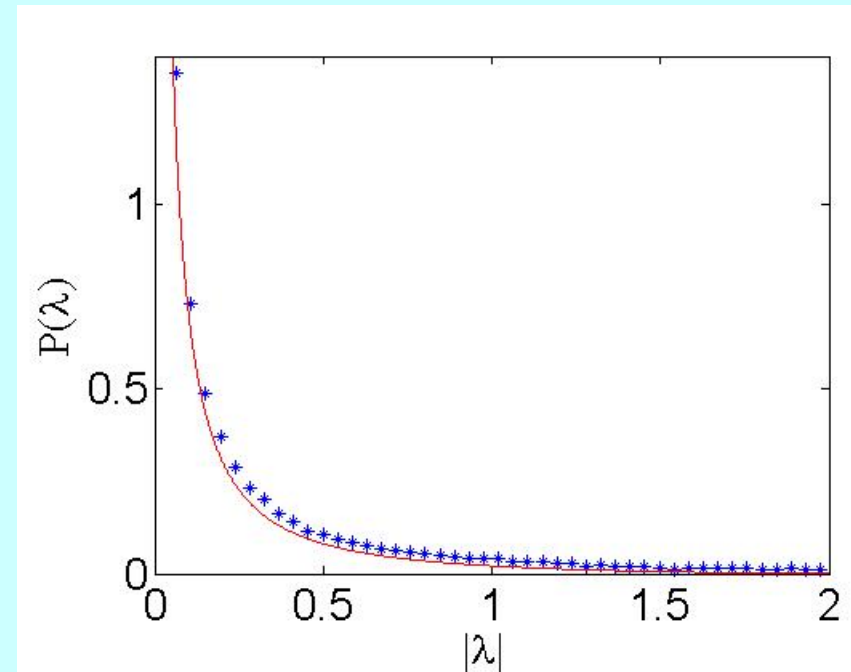
# Distance (Euclidean) matrices- eigenvalue distribution

- Low density system  $\longrightarrow$  basically a set of independent dipoles.
- Calculation of the nearest-neighbor probability distribution will give approximately the eigenvalue distribution.
- Calculation gives, for **exponential** dependence on the distance:

$$P(\lambda) \approx \frac{N\xi}{L} \frac{1}{\lambda} \quad (1D)$$

$$P(\lambda) \approx \frac{\pi N\xi^2}{L^2} \frac{\log[-\lambda/2]}{\lambda} \quad (2D)$$

Heuristic calc—almost the same!





## Is this more general?

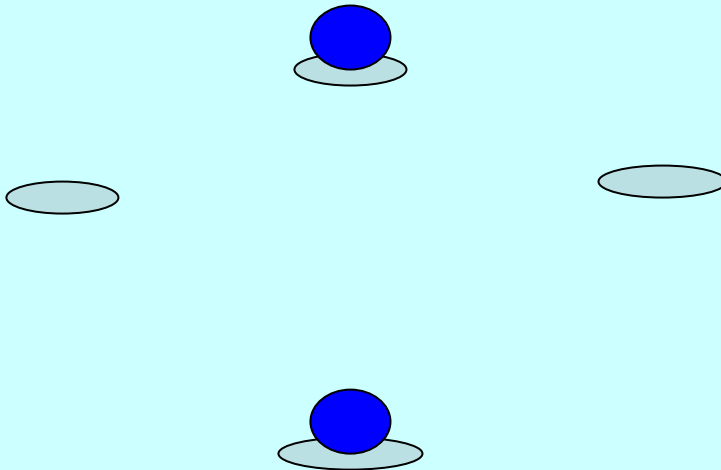
Benford's law: for many physical properties the *log* is uniformly distributed.

Examples: river lengths, phone bills, **1/f** noise etc.

Implies: distribution of first digit  $\sim \log\left(\frac{1}{d} + 1\right)$

More concretely:

*What happens when we take multi-particle relaxations into account?*



## Connection to conductance relaxation?

If we assume:

$$\delta\sigma \sim g(\mu)$$

The energies of sites are changed, we get a finite DOS at the chemical potential.

$$\delta E = \sum \frac{e^2 \delta n_i}{r_{ij}}$$

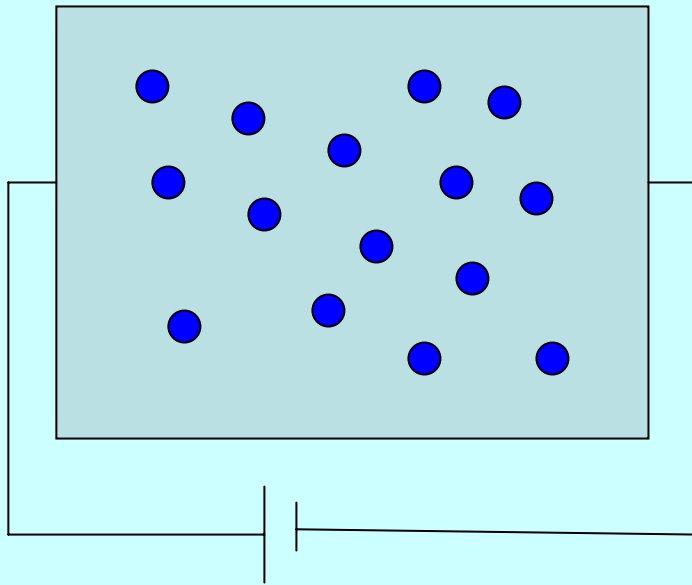
$g(\mu) \sim \delta n \rightarrow$  logarithmic decay of the conductance?

Other mechanisms might be involved.

## Mean-field approximation – steady state

Miller-Abrahams resistance network: essentially mean-field

Find  $n_i$  and  $E_i$  such that the systems is in steady state.



$$I = \sum_j \frac{U_i - U_j}{R_{ij}}, \quad R_{ij} = \frac{T}{e^2 \gamma_{ij}^0}$$

↑  
equilibrium rates

$$U_i = -\vec{E} \cdot \vec{r} + e^2 \sum_{k \neq i} \frac{\delta n_k}{r_{ik}} + \delta \mu_i$$

Leads to variable-range-hopping:  $\sigma \sim e^{CT \frac{1}{d+1}}$

**A. Miller and E. Abrahams, (Phys. Rev. 1960)**

## Variable Range Hopping – back of the envelope derivation

Einstein formula:

$$\sigma = e^2 \frac{\partial n}{\partial \mu} D$$

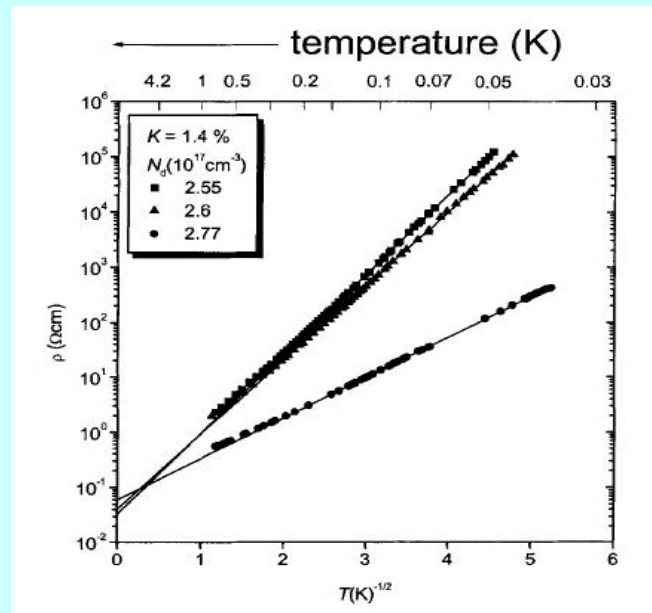
The typical diffusion coefficient:

$$r(T) \sim T^{-\frac{1}{d+1}}, \quad E(T) \sim T^{\frac{d}{d+1}}$$

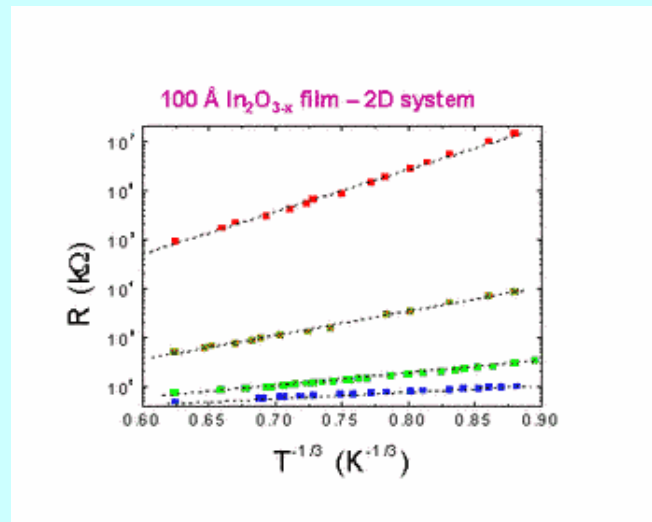
$$\sigma \sim e^{CT^{\frac{1}{d+1}}}$$

Repeating the optimization with a Coulomb gap:

$$\sigma \sim e^{CT^{\frac{1}{2}}}$$

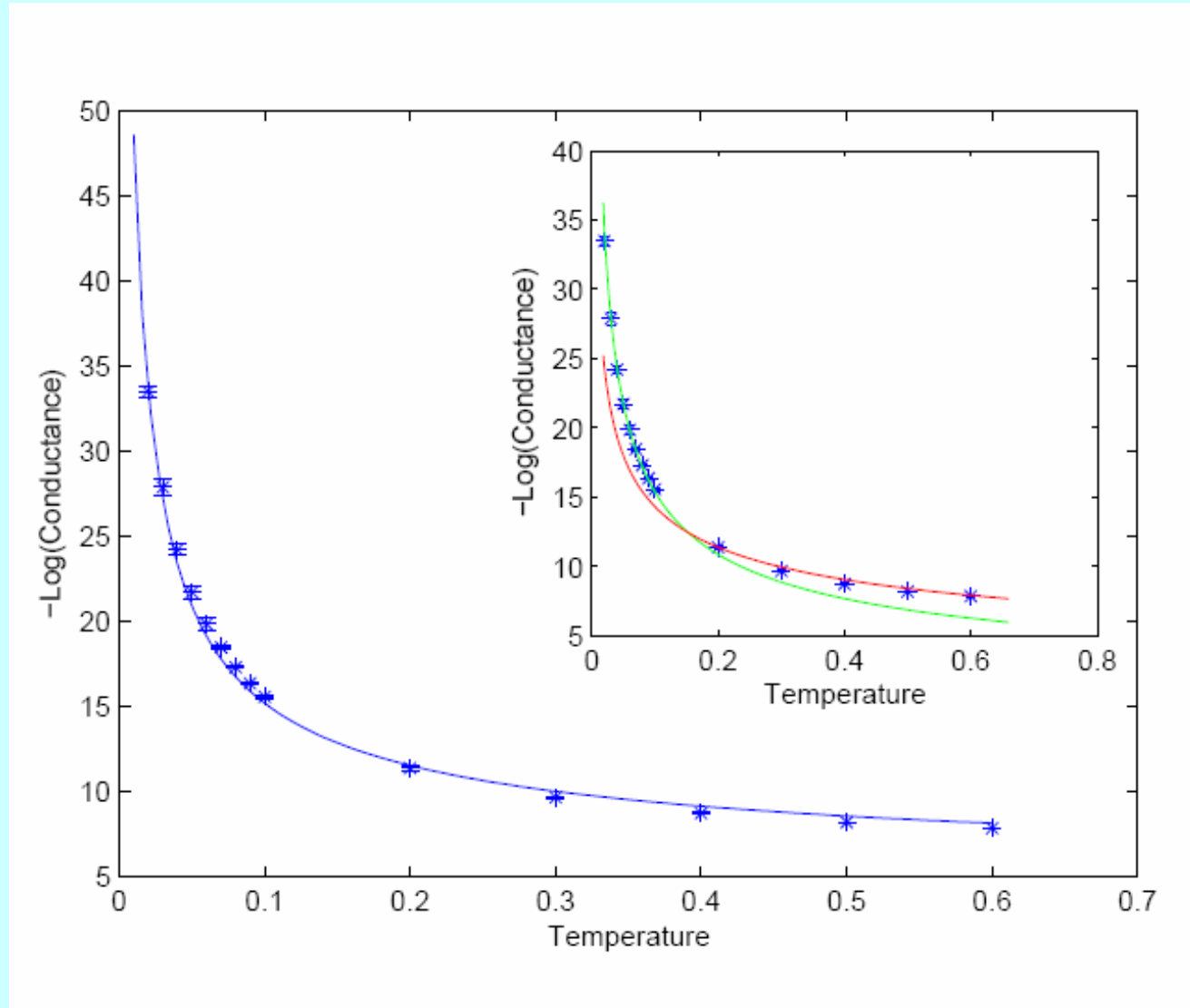


Rentzsch et al. (2001)



Ovadyahu (2003)

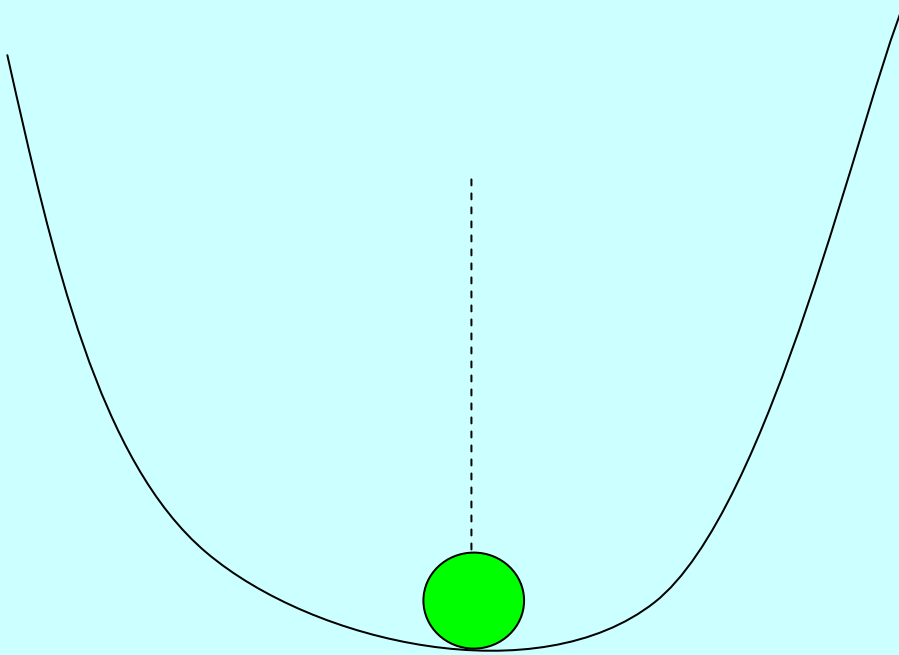
# VRH (Mott) to E-S Crossover, from mean-field



# Aging

Assume a parameter of the system is slightly modified (e.g:  $V_g$ )

After time  $t_w$  it is changed back. What is the response?

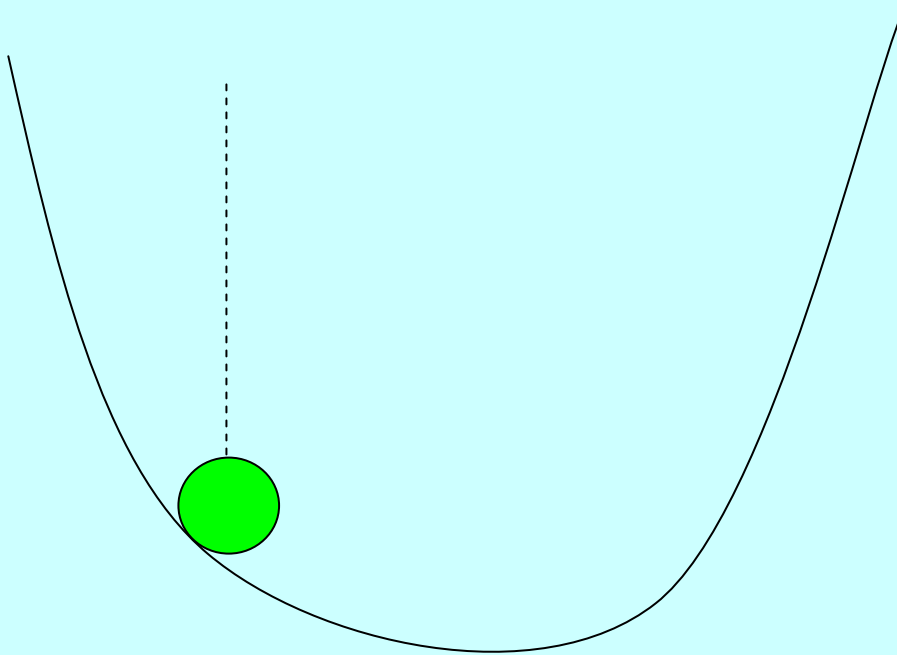


**Initially, system is at some local minimum**

# Aging

Assume a parameter of the system is slightly modified (e.g:  $V_g$ )

After time  $t_w$  it is changed back. What is the response?



**At time  $t=0$  the potential changes,**

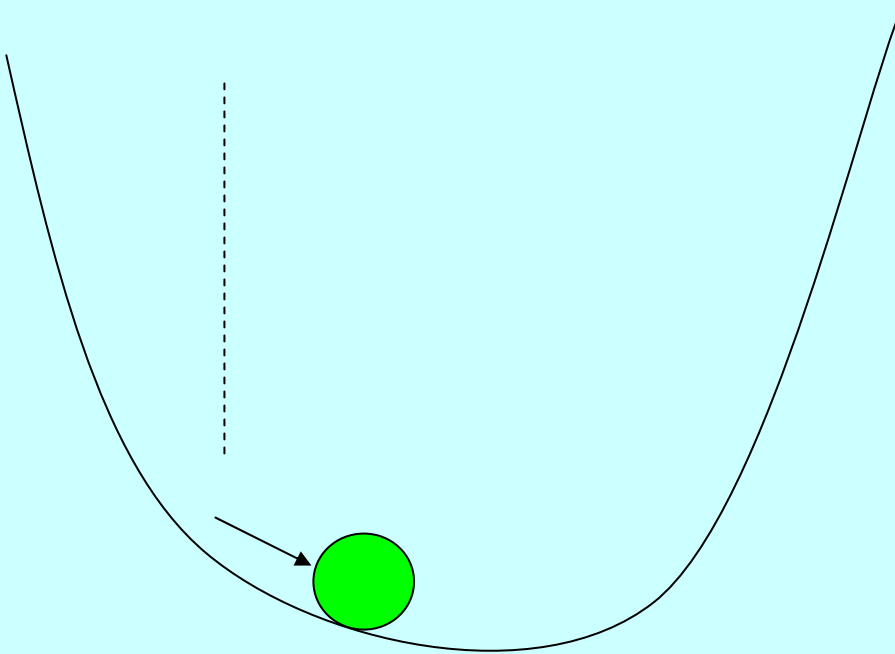
**and the system begins to roll towards the new minimum**



# Aging

Assume a parameter of the system is slightly modified (e.g:  $V_g$ )

After time  $t_w$  it is changed back. What is the response?

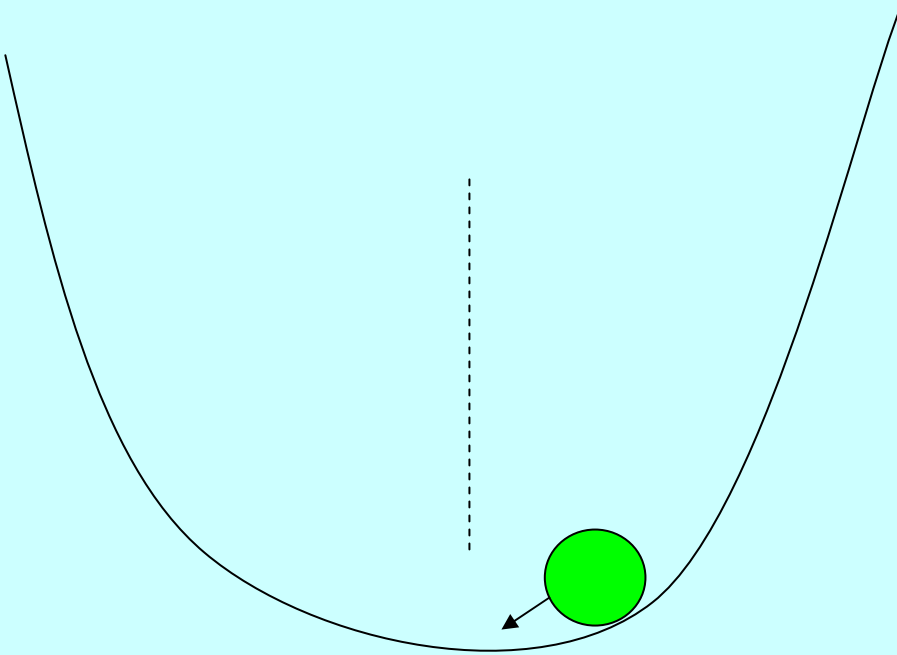


**At time  $t_w$  the system reached some new configuration**

# Aging

Assume a parameter of the system is slightly modified (e.g:  $V_g$ )

After time  $t_w$  it is changed back. What is the response?



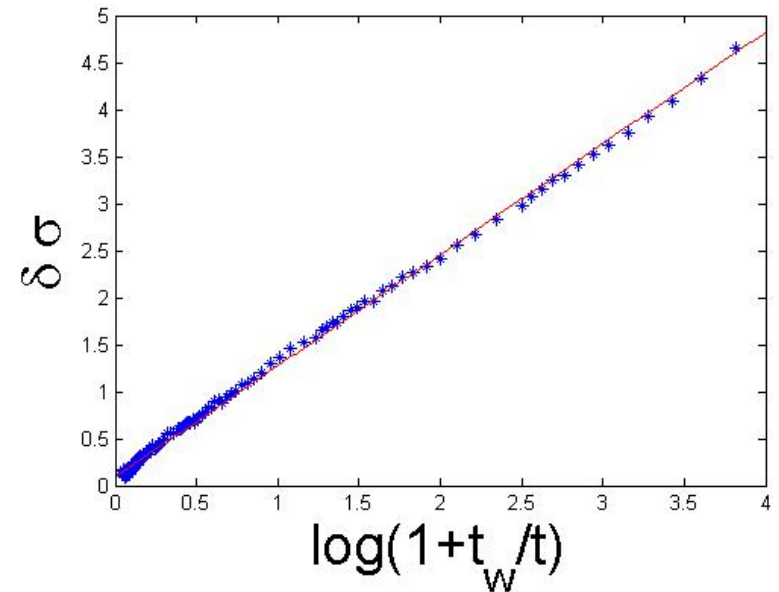
**Now the potential is changed back to the initial form-  
the particle is not in the minima!**

**The longer  $t_w$ , the further it got away from it.**

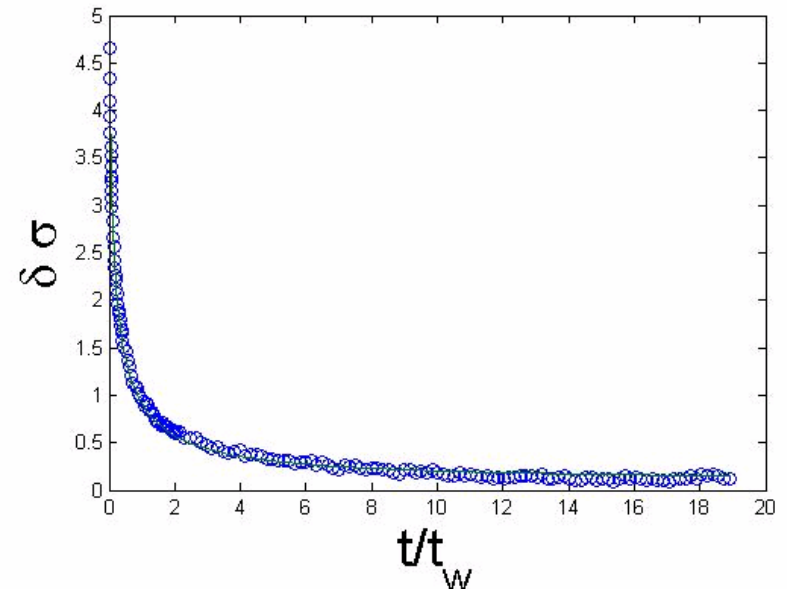
**It will begin to roll down the hill.**

## Results

- Simple aging (function of  $t/t_w$ )
- $\log(t)$  turns to a power-law at large times
- Not stretched exponential!
- Fits experimental data!



Data courtesy of Z. Ovadyhau



# Conclusions and future work

- Dynamics near locally stable point: slow,  $\log t$ , relaxations. How universal is the  $1/\lambda$  distribution? **a very relevant RMT class.**
- Slow dynamics may arise without transitions between *different* metastable states. (work in progress). How will the inter-state transitions connect with intra-state ones?  
**We believe: They'll push the distribution to lower  $\lambda$ !**
- It is interesting to see if the mean-field model can predict the 'two-dip' experiment, where the system shows memory effects.  
**More details: arXiv 0712.0895, Phys. Rev. B 77, 1, 2008**