

The Out-of-Equilibrium Anderson impurity model :  
a numerically exact approach with Diagrammatic Quantum Quasi Monte-Carlo

Calculating Feynman diagrams analytically is impractical beyond the first few orders. In this talk, I will discuss recent numerical algorithms that allow one to calculate all diagrams up to order 20 or more. I will show how this technique can be used to study the Anderson model, including for parameters deep into the Kondo regime both at equilibrium (in precise agreement with other techniques) and in out-of-equilibrium situations that were not accessible so far.

Correlations and computational quantum transport: automatic calculation of Feynman diagrams at large orders.

While numerical simulations of quantum transport at the mean field level are by now standard, including even the simplest effects of correlations such as Coulomb blockade is a struggle and one almost always has to resort to uncontrolled approximations or drastic simplifications to account for more complex effects such as Kondo physics.

Feynman diagrams are a natural formalism to express correlations and they circumvent working at the raw Hilbert space level with many-body wavefunctions as is often done numerically (either in quantum Monte-Carlo or tensor network techniques). An immense literature study different ways to find the most important classes of diagrams and calculate them analytically. In this talk, I will discuss the current effort done in my group to design algorithms that numerically sum all diagrams in a systematic way, order by order up to order 20 or more. I will showcase how this technique can capture out-of-equilibrium effects deep in the Kondo regime.

# CORRELATIONS AND COMPUTATIONAL QUANTUM TRANSPORT: AN APPROACH FOR CALCULATING FEYNMAN DIAGRAMS AT LARGE ORDERS

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Olivier Parcollet, Bill Triggs and Serge Florens

# TOOLS FOR QUANTUM TRANSPORT

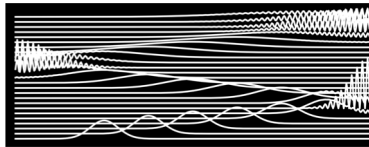
$$\hat{H} = \sum_{ij} H_{ij} c_i^\dagger c_j$$

kwant

[HTTP://KWANT-PROJECT.ORG](http://kwant-project.org)

With TU Delft (Akhmerov, Wimmer et al.)

$$\hat{H}(t) = \sum_{ij} H_{ij}(t) c_i^\dagger c_j$$



[HTTP://TKWANT.KWANT-PROJECT.ORG](http://tkwant.kwant-project.org)

$$\hat{\mathbf{H}}(t) = \hat{\mathbf{H}}_0(t) + U \hat{\mathbf{H}}_{\text{int}}(t)$$

$$\hat{\mathbf{H}}_{\text{int}}(t) = \sum_{ijkl} \mathbf{V}_{ijkl}(t) \hat{\mathbf{c}}_i^\dagger \hat{\mathbf{c}}_j^\dagger \hat{\mathbf{c}}_k \hat{\mathbf{c}}_l.$$

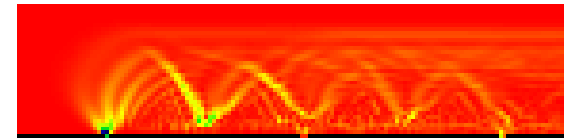
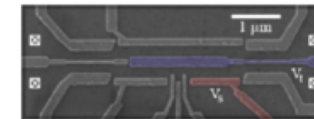
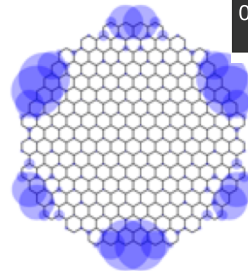
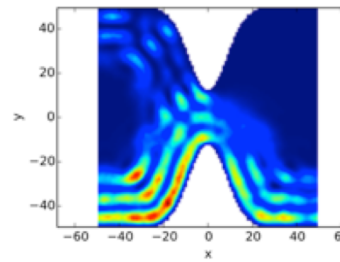
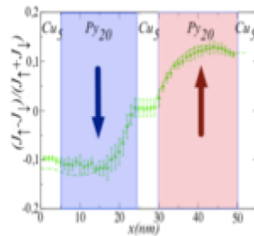
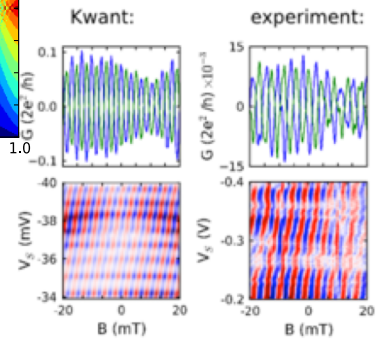
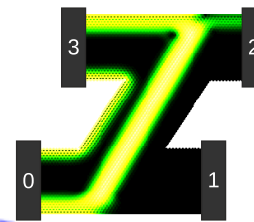
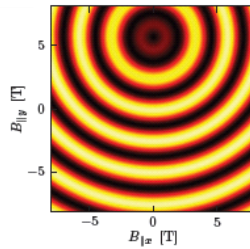
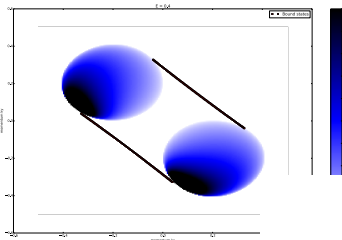
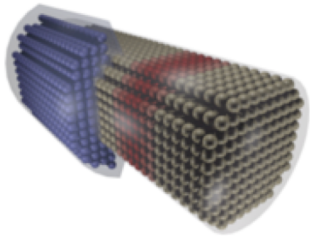
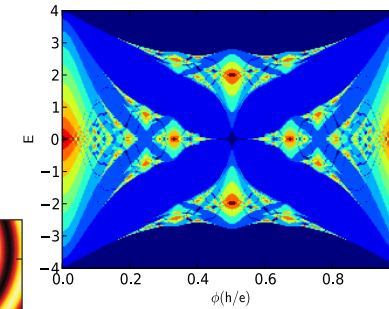
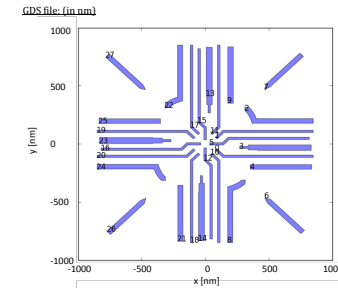
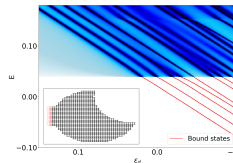
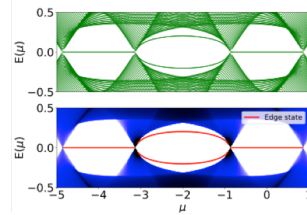
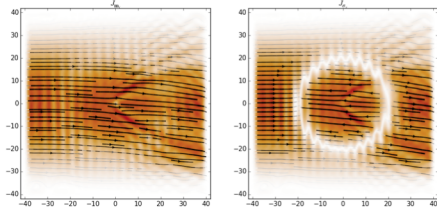


- Kondo physics
- Coulomb blockade, Fermi edge singularity
- 0.7 anomaly
- FQHE
- Quantum computers
- ...

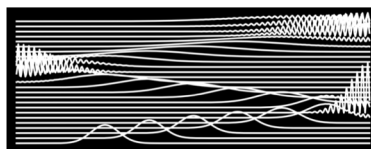
# KWANT GALLERY

With TU Delft (Akhmerov, Wimmer et al.)

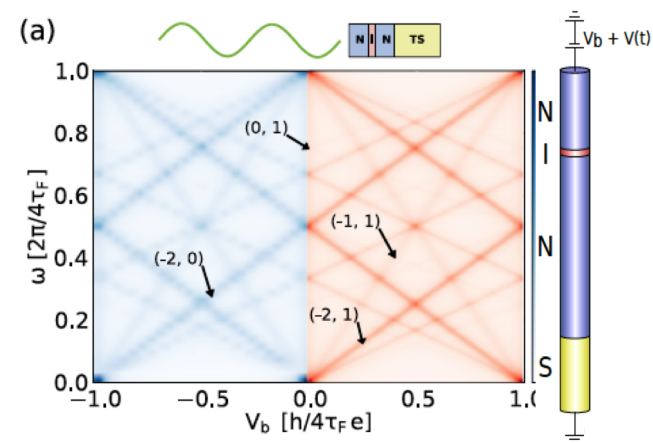
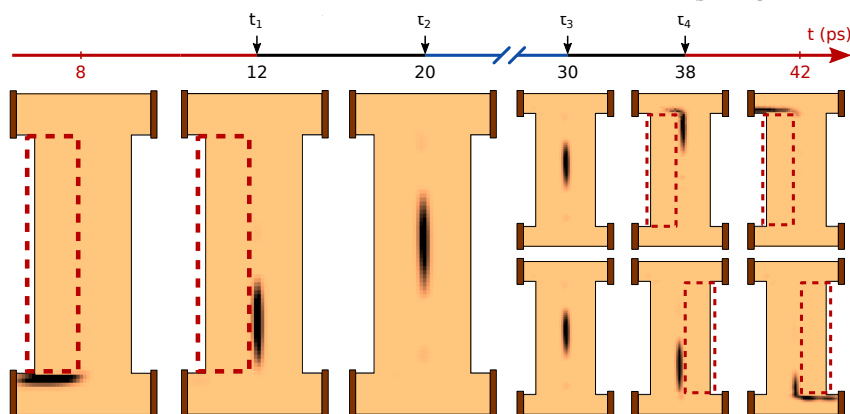
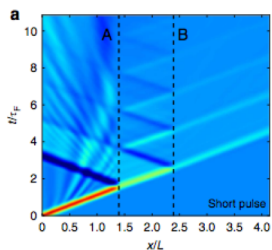
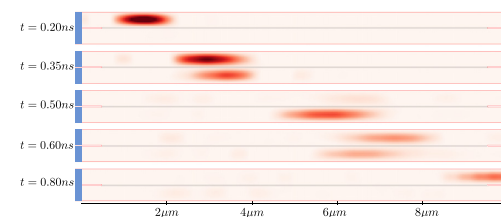
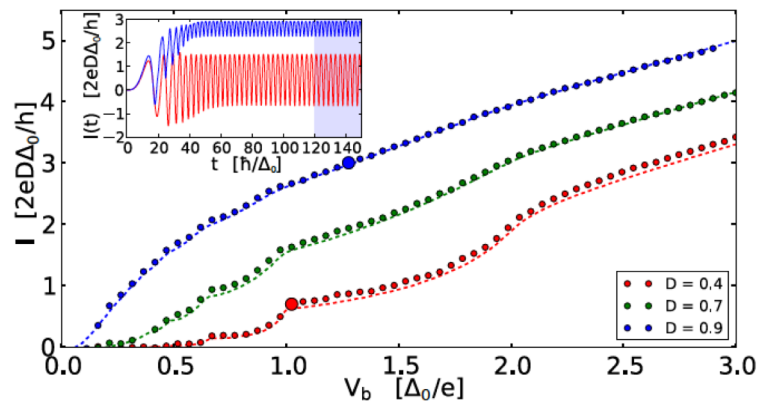
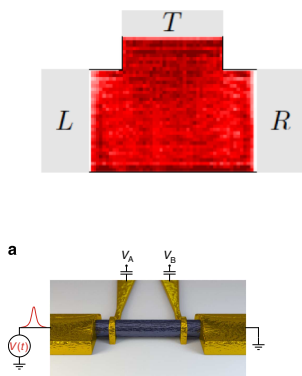
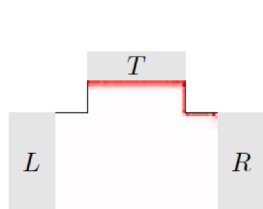
[HTTP://KWANT-PROJECT.ORG](http://kwant-project.org)



# T-KWANT

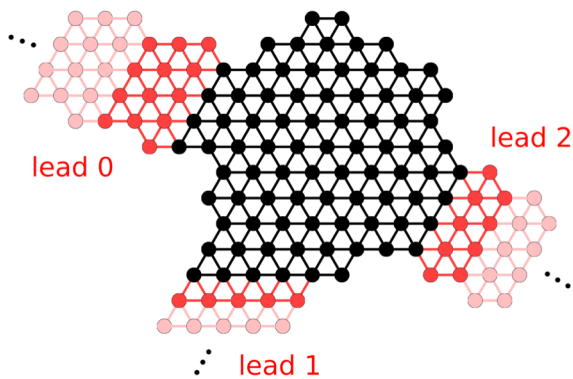


[HTTP://TKWANT.KWANT-PROJECT.ORG](http://tkwant.kwant-project.org)

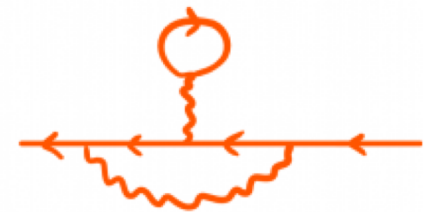


$$\hat{\mathbf{H}}(t) = \hat{\mathbf{H}}_0(t) + U \hat{\mathbf{H}}_{\text{int}}(t)$$

$$\hat{\mathbf{H}}_{\text{int}}(t) = \sum_{ijkl} \mathbf{V}_{ijkl}(t) \hat{\mathbf{c}}_i^\dagger \hat{\mathbf{c}}_j^\dagger \hat{\mathbf{c}}_k \hat{\mathbf{c}}_l.$$

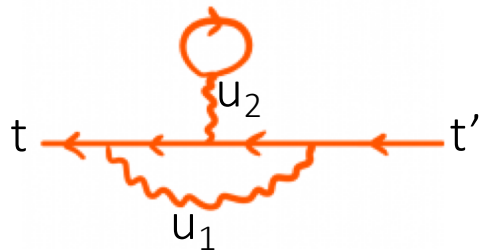


$$Q(U) = \sum_{n=0}^{+\infty} Q_n U^n$$



# A VERY DIRECT APPROACH:

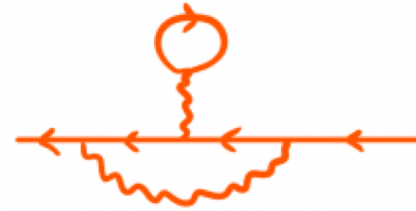
**CALCULATING ALL THE FEYNMAN DIAGRAMS UP TO A GIVEN (LARGE) ORDER**



$$\sim U^2 \int du_1 du_2 g_0(t, u_1) g_0(u_1, u_2) g_0(u_2, u_1) g_0(u_1, t') g_0(u_2, u_2)$$



# A VERY DIRECT APPROACH:



**CALCULATING ALL THE FEYNMAN DIAGRAMS UP TO A GIVEN (LARGE) ORDER**

- PROBLEM #1: There are  $n!$  diagrams.

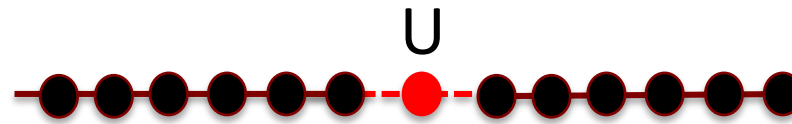
$$F(U) = \sum_n F_n U^n$$

- PROBLEM #2 How to calculate  $n$  dimensional integrals
- PROBLEM #3 How to reconstruct  $F(U)$  from the  $F_n$ .

# THE OUT-OF-EQUILIBRIUM ANDERSON IMPURITY

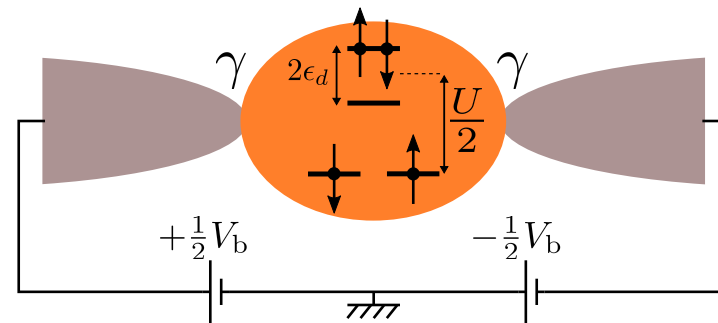


=



$$\hat{H} = \sum_{i=-\infty}^{+\infty} \sum_{\sigma} \gamma_i \hat{c}_{i,\sigma}^{\dagger} \hat{c}_{i+1,\sigma} + h.c. + \epsilon_d (\hat{n}_{\uparrow} + \hat{n}_{\downarrow}) + U \theta(t) \left( \hat{n}_{\uparrow} - \frac{1}{2} \right) \left( \hat{n}_{\downarrow} - \frac{1}{2} \right).$$

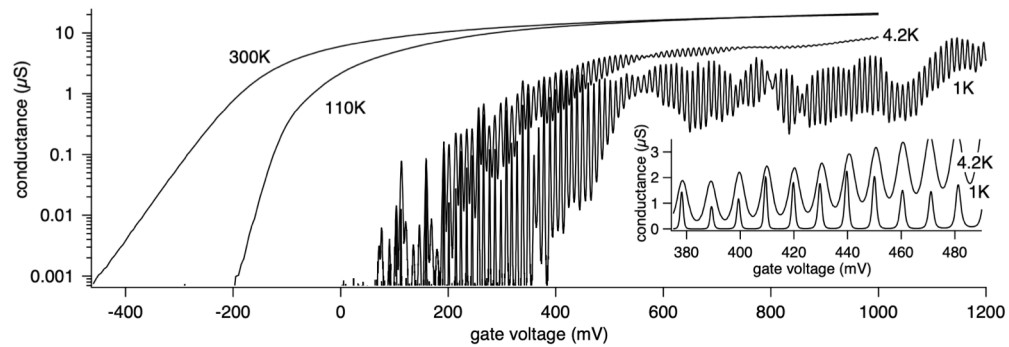
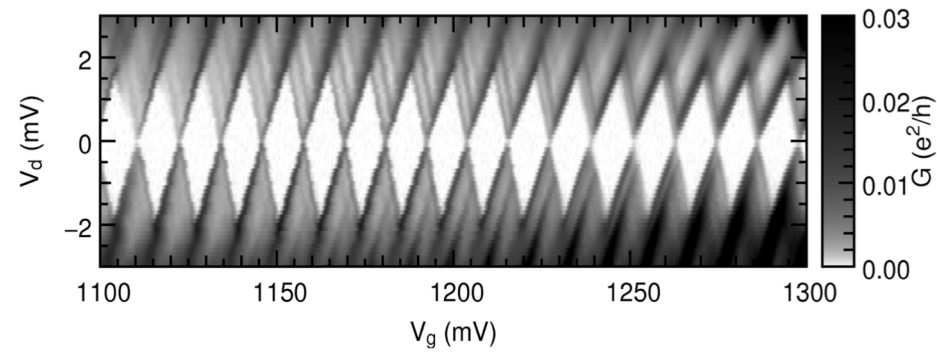
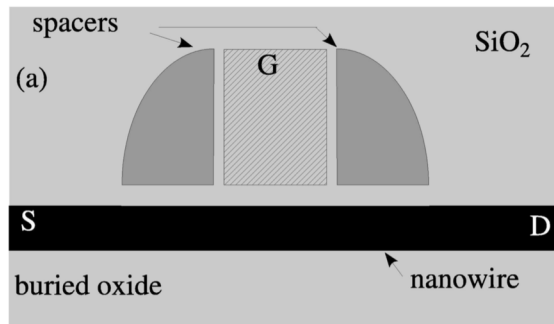
# COULOMB BLOCKADE 101



$$E(N) = \frac{U}{2}N^2 + (\epsilon_d - U)N + \frac{U}{2}$$

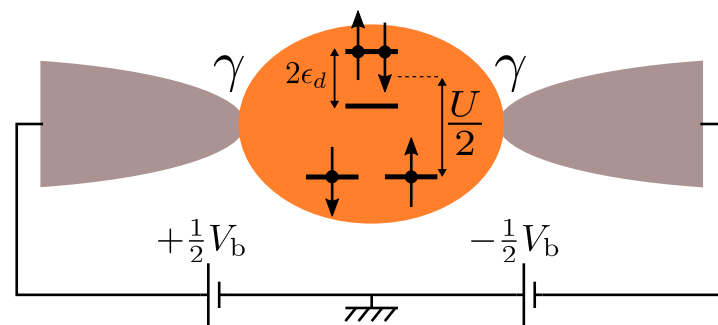
$$E(N + 1) - E(N) = (\epsilon_d - U) + (2N + 1)\frac{U}{2}$$

# COULOMB BLOCKADE 101



Hofheinz et al. [arXiv:cond-mat/0609245](https://arxiv.org/abs/cond-mat/0609245)

# KONDO 101



**U**

**$\gamma^2/U$**

REPULSIVE  
INTERACTION

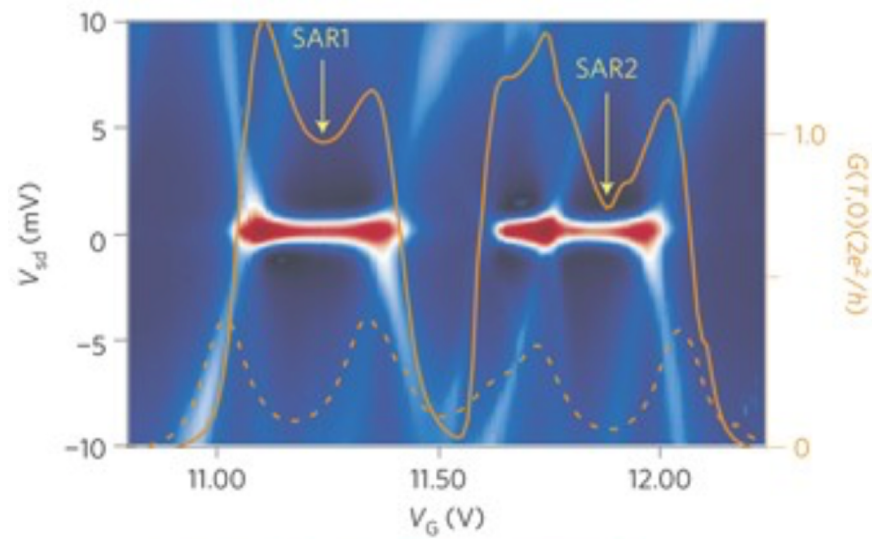


ANTIFERROMAGNETIC  
COUPLING



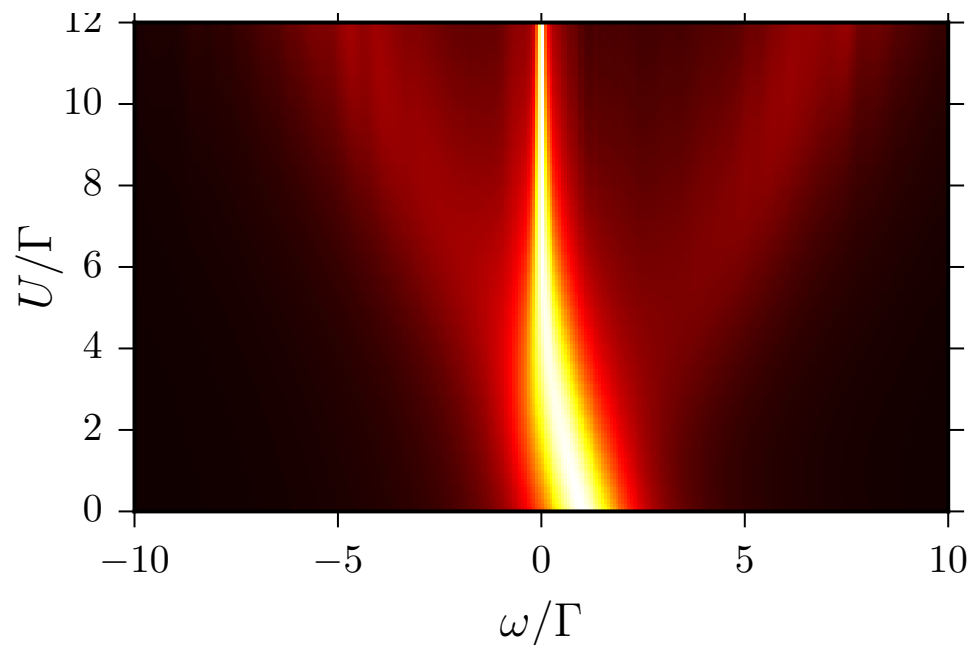
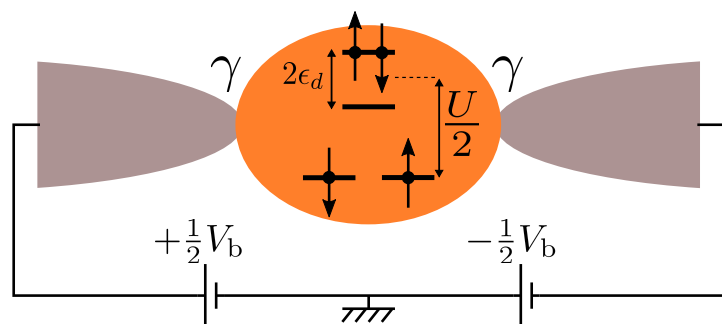
ATTRACTIVE  
INTERACTION

# KONDO 101



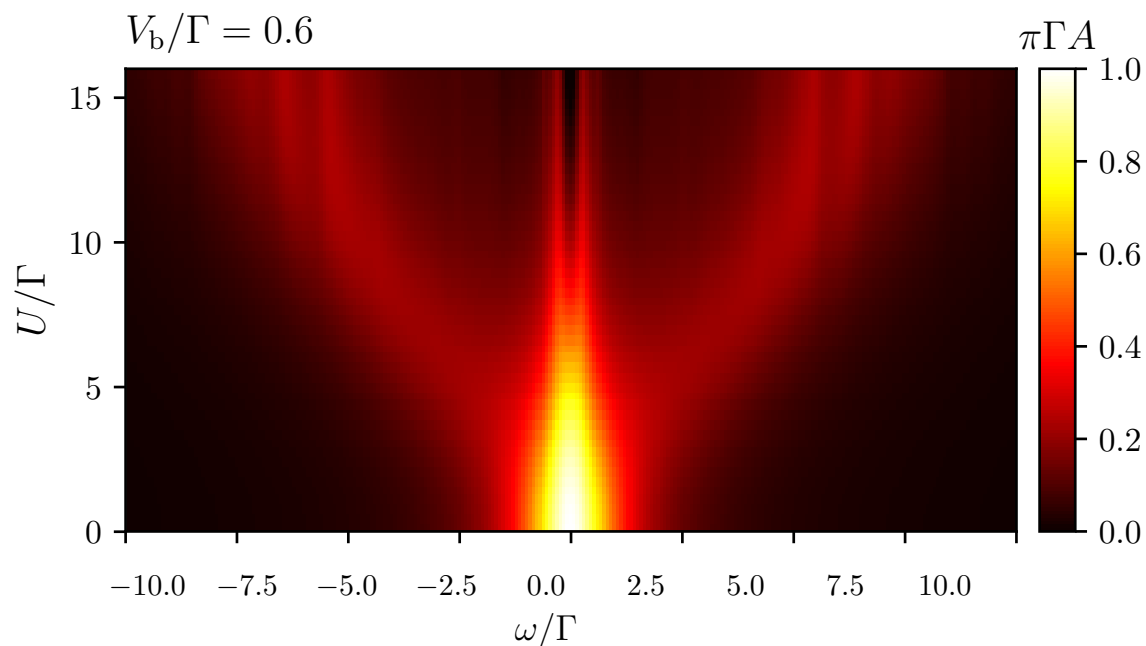
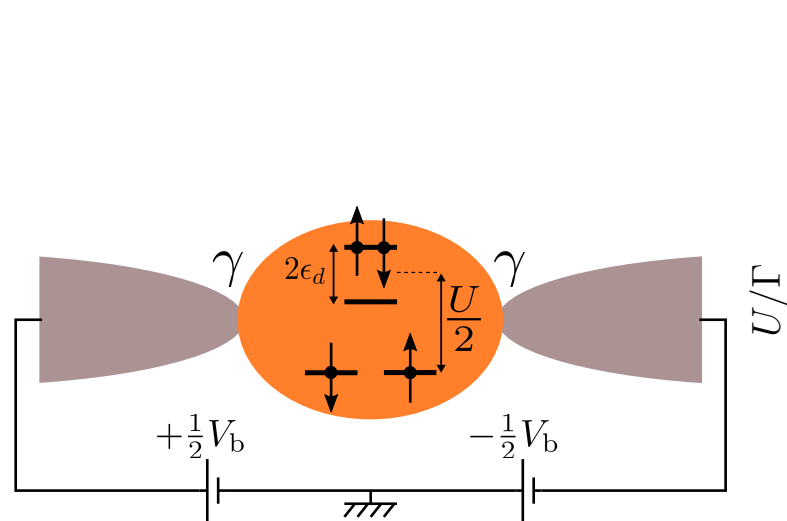
*Nature Physics* **5**, 208 (2009)

# TEASER: AT EQUILIBRIUM



$$\hat{\mathbf{H}} = \sum_{i=-\infty}^{+\infty} \sum_{\sigma} \gamma_i \hat{\mathbf{c}}_{i,\sigma}^{\dagger} \hat{\mathbf{c}}_{i+1,\sigma} + h.c. + \epsilon_d (\hat{\mathbf{n}}_{\uparrow} + \hat{\mathbf{n}}_{\downarrow}) + U\theta(t) \left( \hat{\mathbf{n}}_{\uparrow} - \frac{1}{2} \right) \left( \hat{\mathbf{n}}_{\downarrow} - \frac{1}{2} \right).$$

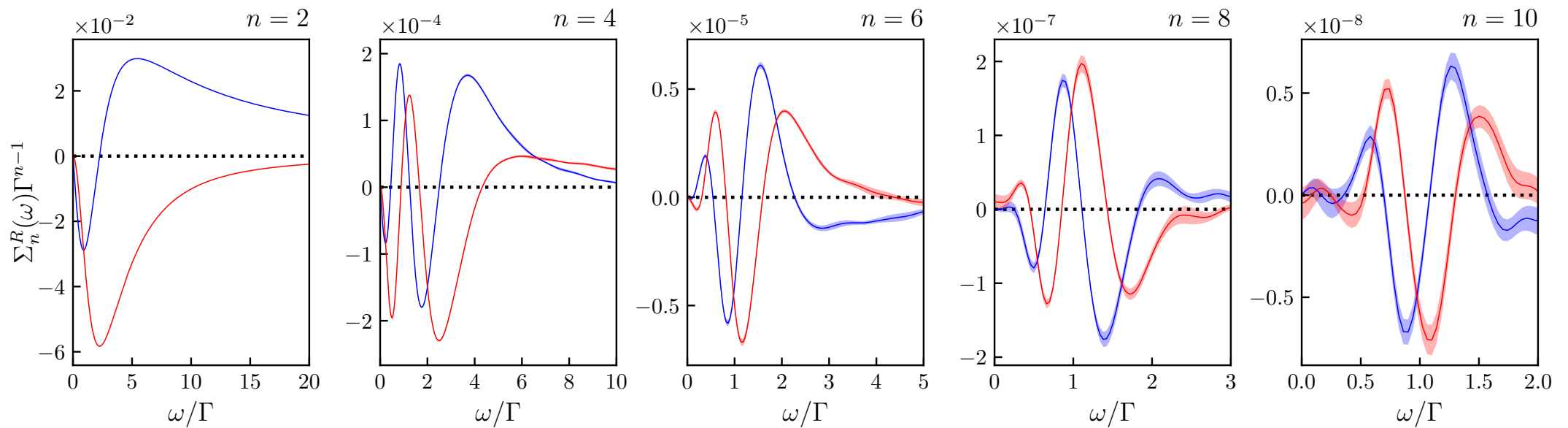
# TEASER: OUT-OF-EQUILIBRIUM



$$\hat{H} = \sum_{i=-\infty}^{+\infty} \sum_{\sigma} \gamma_i \hat{\mathbf{c}}_{i,\sigma}^{\dagger} \hat{\mathbf{c}}_{i+1,\sigma} + h.c. + \epsilon_d (\hat{\mathbf{n}}_{\uparrow} + \hat{\mathbf{n}}_{\downarrow}) + U\theta(t) \left( \hat{\mathbf{n}}_{\uparrow} - \frac{1}{2} \right) \left( \hat{\mathbf{n}}_{\downarrow} - \frac{1}{2} \right).$$



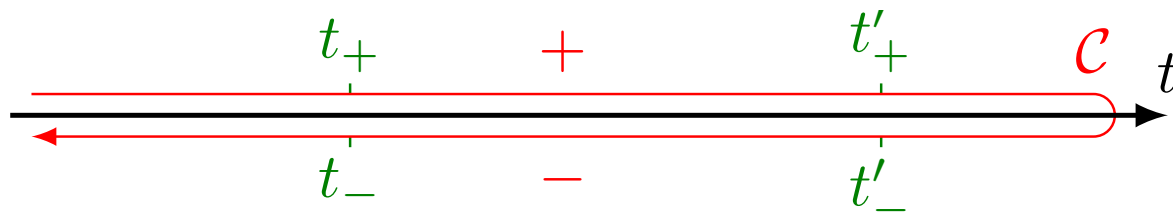
# PROBLEM #1: THE $n!$ DIAGRAMS



# KELDYSH FORMALISM IN A NUTSHELL

$$\langle \mathcal{O}(t) \rangle = \langle \mathcal{U}^\dagger(t) \mathcal{O} \mathcal{U}(t) \rangle$$

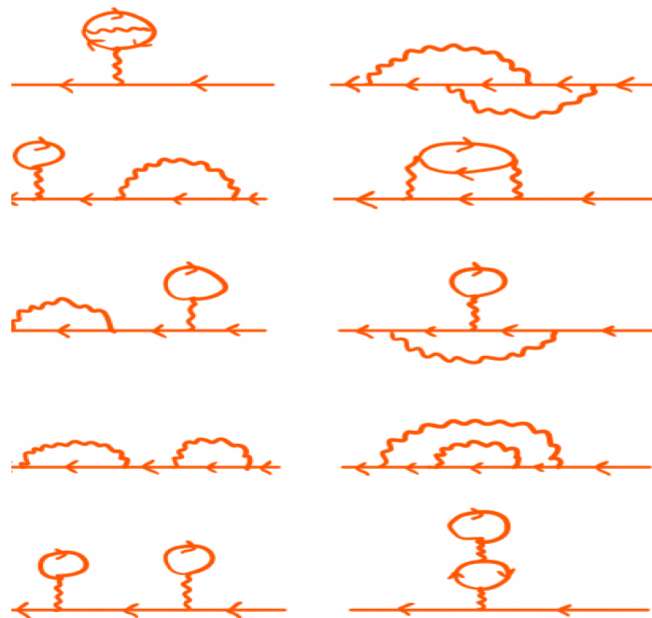
$$\mathcal{U}(t) = T \exp \left( -i \int_0^t \hat{H}_{\text{int}}(u) du \right)$$



$$\langle \mathcal{O}(t) \rangle = \left\langle T_C \hat{\mathcal{O}}(t) \exp \left( -i \int_C \hat{H}_{\text{int}}(u) du \right) \right\rangle$$

# WICK DETERMINANTS

$$\langle c_1^+ c_1 c_2^+ c_2 c_3^+ c_3 c_4^+ c_4 c_5^+ c_5 \rangle = \sum_P (-1)^{|P|} \langle c_1^+ c_{P(1)} \rangle \langle c_2^+ c_{P(2)} \rangle \langle c_3^+ c_{P(3)} \rangle \langle c_4^+ c_{P(4)} \rangle \langle c_5^+ c_{P(5)} \rangle$$



# WICK DETERMINANTS

$$\langle c_1^+ c_1 c_2^+ c_2 c_3^+ c_3 c_4^+ c_4 c_5^+ c_5 \rangle = \sum_P (-1)^{|P|} \langle c_1^+ c_{P(1)} \rangle \langle c_2^+ c_{P(2)} \rangle \langle c_3^+ c_{P(3)} \rangle \langle c_4^+ c_{P(4)} \rangle \langle c_5^+ c_{P(5)} \rangle$$

$$\langle c_1^+ c_1 c_2^+ c_2 c_3^+ c_3 c_4^+ c_4 c_5^+ c_5 \rangle = \det \langle c_i^+ c_j \rangle$$

# A « VERY SIMPLE » FORMULA

$$G_{ij}^c(\bar{t}, \bar{t}') = \sum_{n=0}^{+\infty} \frac{i^n}{n!} U^n \sum_{\{a_i\}} (-1)^{\sum_i a_i} \int du_1 du_2 \dots du_n \sum_{i_1 j_1 k_1 l_1} V_{i_1 j_1 k_1 l_1}(u_1) \dots \sum_{i_n j_n k_n l_n} V_{i_n j_n k_n l_n}(u_n) \det \mathbf{M}_n$$

2<sup>n</sup> sum cancels disconnected diagrams

$$\mathbf{M}_n = \begin{pmatrix} g_{k_1 i_1}^<(\bar{u}_1, \bar{u}_1) & g_{k_1 j_1}^<(\bar{u}_1, \bar{u}_1) & g_{k_1 i_2}^c(\bar{u}_1, \bar{u}_2) & \dots & g_{k_1 j}^c(\bar{u}_1, \bar{t}') \\ g_{l_1 i_1}^<(\bar{u}_1, \bar{u}_1) & g_{l_1 j_1}^<(\bar{u}_1, \bar{u}_1) & g_{l_1 i_2}^c(\bar{u}_1, \bar{u}_2) & \dots & g_{l_1 j}^c(\bar{u}_1, \bar{t}') \\ g_{k_2 i_1}^c(\bar{u}_2, \bar{u}_1) & g_{k_2 j_1}^c(\bar{u}_2, \bar{u}_1) & g_{k_2 i_2}^<(\bar{u}_2, \bar{u}_2) & \dots & g_{k_2 j}^c(\bar{u}_2, \bar{t}') \\ \dots & \dots & \dots & \dots & \dots \\ g_{k_n i_1}^c(\bar{u}_n, \bar{u}_1) & g_{k_n j_1}^c(\bar{u}_n, \bar{u}_1) & g_{k_n i_2}^c(\bar{u}_n, \bar{u}_2) & \dots & g_{k_n j}^c(\bar{u}_n, \bar{t}') \\ g_{l_n i_1}^c(\bar{u}_n, \bar{u}_1) & g_{l_n j_1}^c(\bar{u}_n, \bar{u}_1) & g_{l_n i_2}^c(\bar{u}_n, \bar{u}_2) & \dots & g_{l_n j}^c(\bar{u}_n, \bar{t}') \\ g_{i i_1}^c(\bar{t}, \bar{u}_1) & g_{i j_1}^c(\bar{t}, \bar{u}_1) & g_{i i_2}^c(\bar{t}, \bar{u}_2) & \dots & g_{i j}^c(\bar{t}, \bar{t}') \end{pmatrix}$$

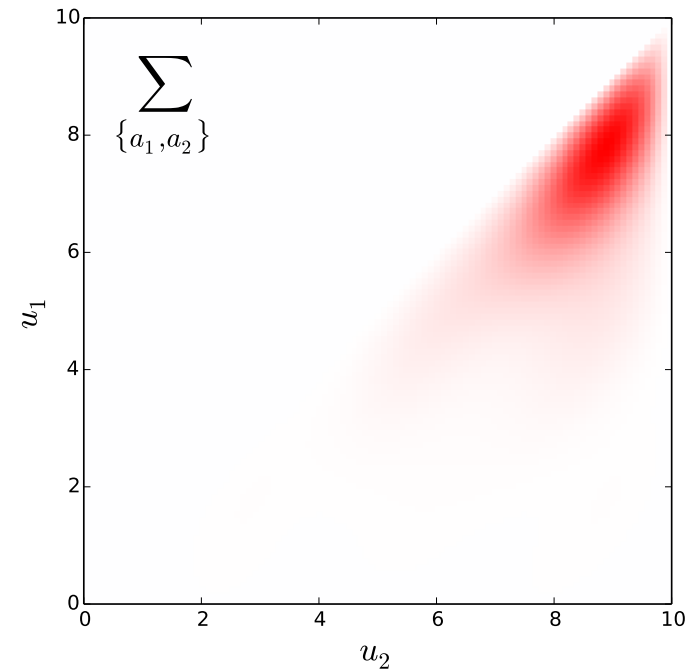
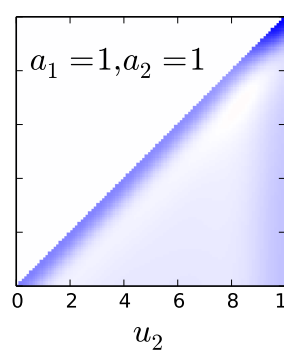
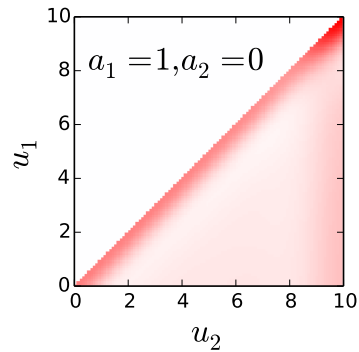
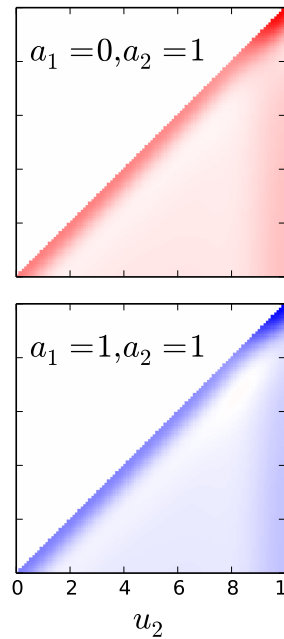
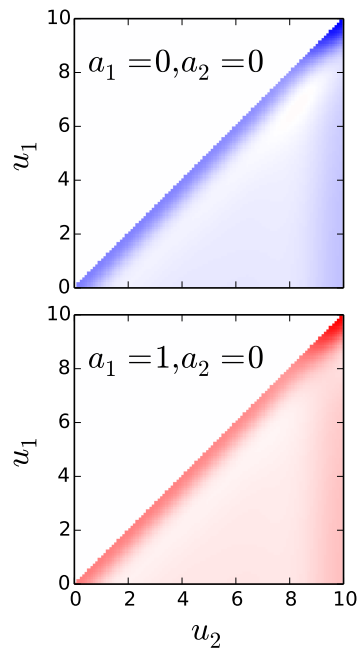


$$g_{ij}^<(t, t') = i \sum_{\alpha} \int \frac{dE}{2\pi} f_{\alpha}(E) \Psi_{\alpha E}(t, i) \Psi_{\alpha E}^*(t', j) + i \sum_n f(E_n) \Psi_n(t, i) \Psi_n^*(t', j)$$

**Known non-interacting functions.**

$$G_{ij}^c(\bar{t}, \bar{t}') = \sum_{n=0}^{+\infty} \frac{i^n}{n!} U^n \sum_{\{a_i\}} (-1)^{\sum_i a_i} \int du_1 du_2 \dots du_n \sum_{i_1 j_1 k_1 l_1} V_{i_1 j_1 k_1 l_1}(u_1) \cdots \sum_{i_n j_n k_n l_n} V_{i_n j_n k_n l_n}(u_n) \det \mathbf{M}_n$$

2<sup>n</sup> sum cancels disconnected diagrams



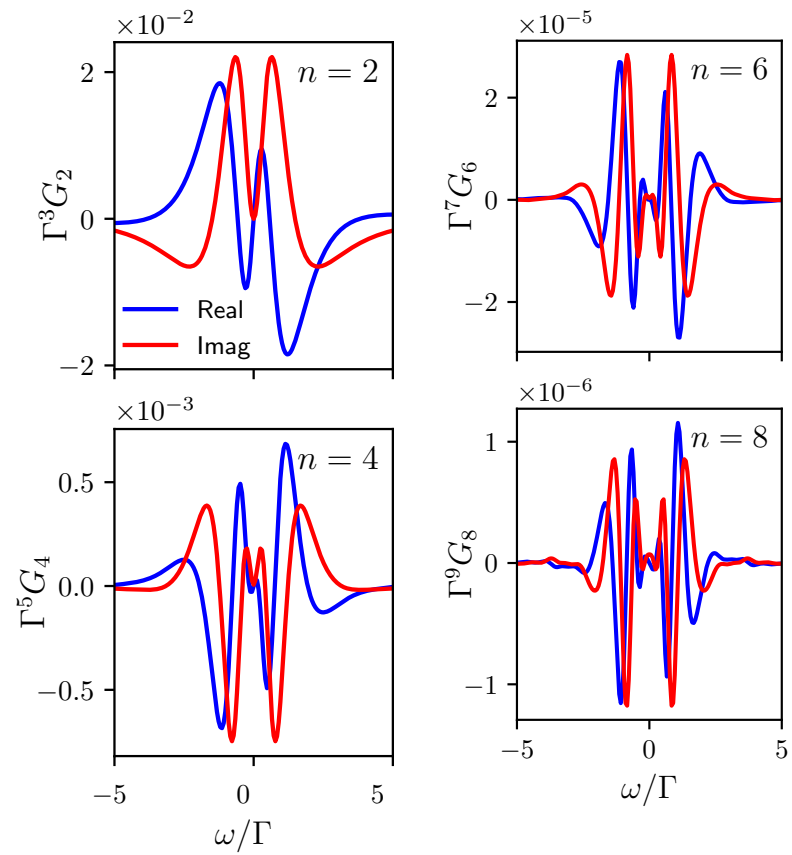
# KERNEL EXPANSION

$$\mathbf{M}_n = \begin{pmatrix} g_{k_1 i_1}^<(\bar{u}_1, \bar{u}_1) & g_{k_1 j_1}^<(\bar{u}_1, \bar{u}_1) & g_{k_1 i_2}^c(\bar{u}_1, \bar{u}_2) & \dots & g_{k_1 j}^c(\bar{u}_1, \bar{t}') \\ g_{l_1 i_1}^<(\bar{u}_1, \bar{u}_1) & g_{l_1 j_1}^<(\bar{u}_1, \bar{u}_1) & g_{l_1 i_2}^c(\bar{u}_1, \bar{u}_2) & \dots & g_{l_1 j}^c(\bar{u}_1, \bar{t}') \\ g_{k_2 i_1}^c(\bar{u}_2, \bar{u}_1) & g_{k_2 j_1}^c(\bar{u}_2, \bar{u}_1) & g_{k_2 i_2}^<(\bar{u}_2, \bar{u}_2) & \dots & g_{k_2 j}^c(\bar{u}_2, \bar{t}') \\ \dots & \dots & \dots & \dots & \dots \\ g_{k_n i_1}^c(\bar{u}_n, \bar{u}_1) & g_{k_n j_1}^c(\bar{u}_n, \bar{u}_1) & g_{k_n i_2}^c(\bar{u}_n, \bar{u}_2) & \dots & g_{k_n j}^c(\bar{u}_n, \bar{t}') \\ g_{l_n i_1}^c(\bar{u}_n, \bar{u}_1) & g_{l_n j_1}^c(\bar{u}_n, \bar{u}_1) & g_{l_n i_2}^c(\bar{u}_n, \bar{u}_2) & \dots & g_{l_n j}^c(\bar{u}_n, \bar{t}') \\ \hline g_{i_1}^c(t, \bar{u}_1) & g_{j_1}^c(t, \bar{u}_1) & g_{i_2}^c(t, \bar{u}_2) & \dots & g_{j}^c(t, \bar{t}') \end{pmatrix}$$

→ Expand the determinant over the last row.  
(all times in a single calculation)

$$G_{xx'}^{aa'}(t, t') = g_{xx'}^{aa'}(t, t') + \int du \sum_{b,y} (-1)^b g_{xy}^{ab}(t, u) K_{yx'}^{ba'}(u, t')$$

# BARE DATA



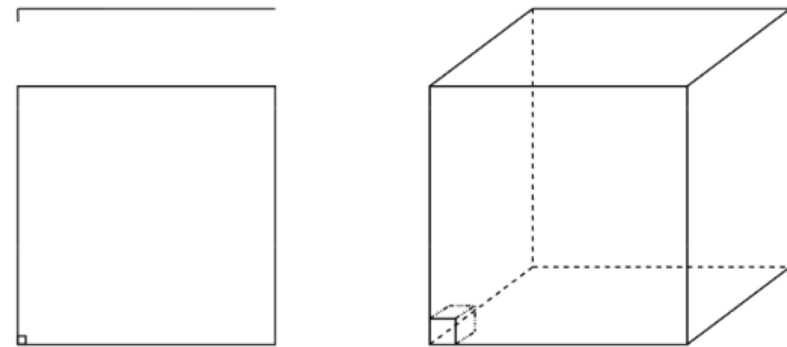


# PROBLEM #2 THE N DIMENSIONAL INTEGRAL

→ The dimensionality curse

$$F(U) = \sum_{n=0}^{\infty} F_n U^n,$$

$$F_n = \int d^n \mathbf{u} f_n(u_1, u_2, \dots, u_n).$$



The standard approach: Metropolis Monte-Carlo.

→ Intrinsic very slow convergence  $N^{-1/2}$  (in one dimension:  $1/N^{15}$  or even exponential)

→ We do not build any knowledge of  $f(u)$

# MACHINE LEARNING THE INTEGRAND

$$f_n(u_1 \dots u_n) \approx p_n(u_1 \dots u_n)$$

Special class of warper functions that we know how to integrate

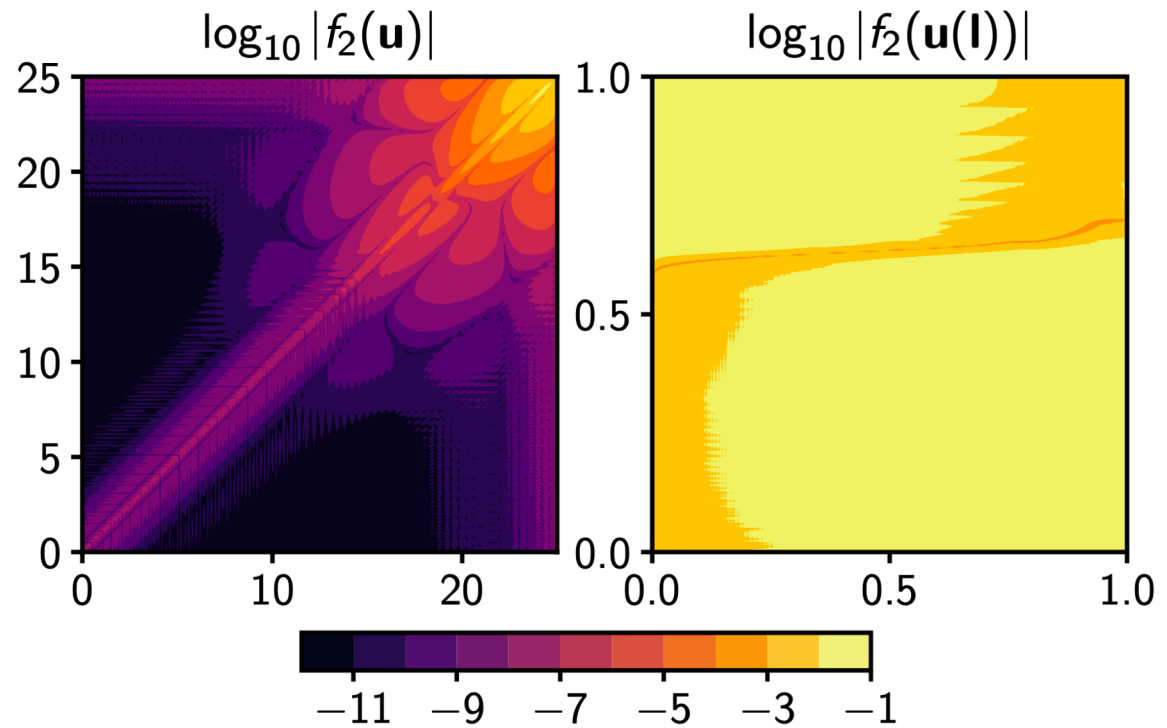
Construct a change of variable  $\mathbf{u}(\mathbf{x})$  such that  $\frac{\partial \mathbf{u}}{\partial \mathbf{x}} = \frac{1}{p_n(u_1 \dots u_n)}$

$$F_n = \int_{[0,1]^n} d^n \mathbf{x} f_n[\mathbf{u}(\mathbf{x})] \left| \frac{\partial \mathbf{u}}{\partial \mathbf{x}} \right|$$

And get an almost constant integrand

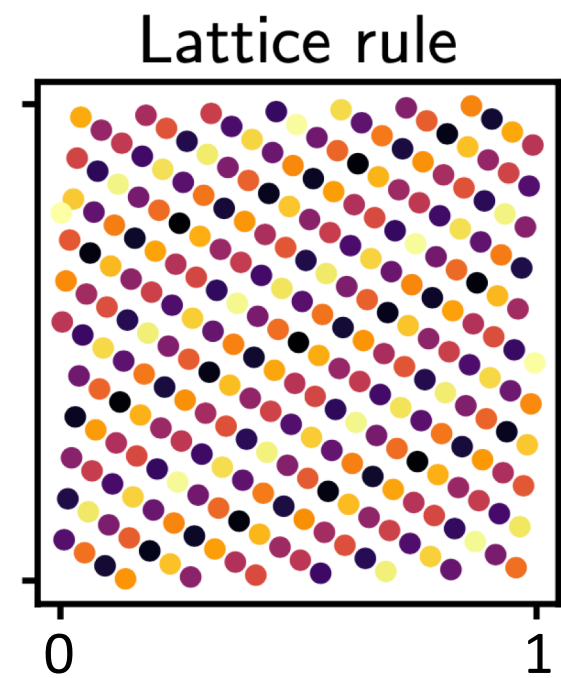
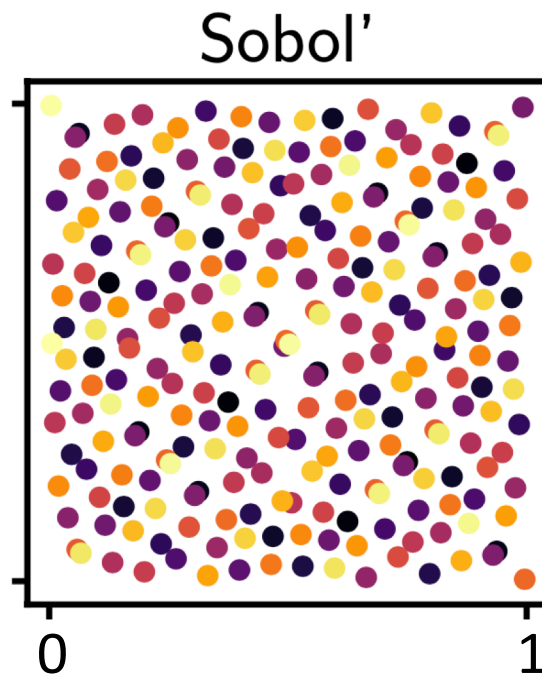
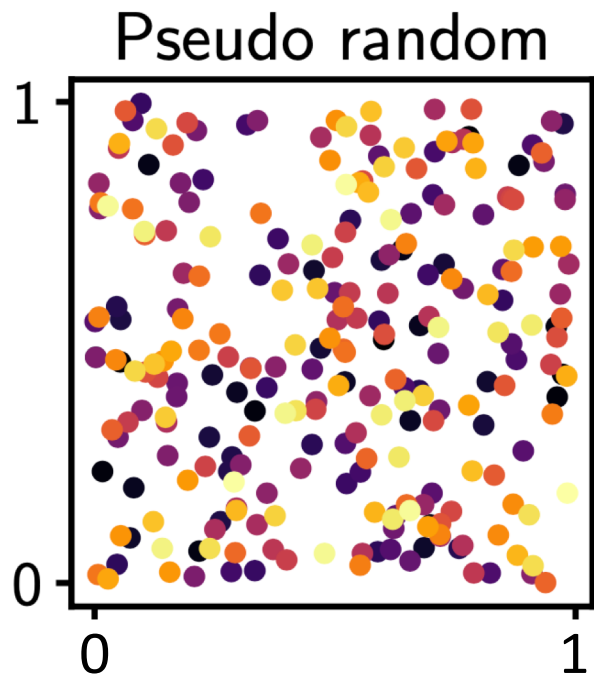
$$p_n(\mathbf{u}) = \prod_{i=1}^n h^{(i)}(v_i),$$

$$v_i = u_{i-1} - u_i,$$

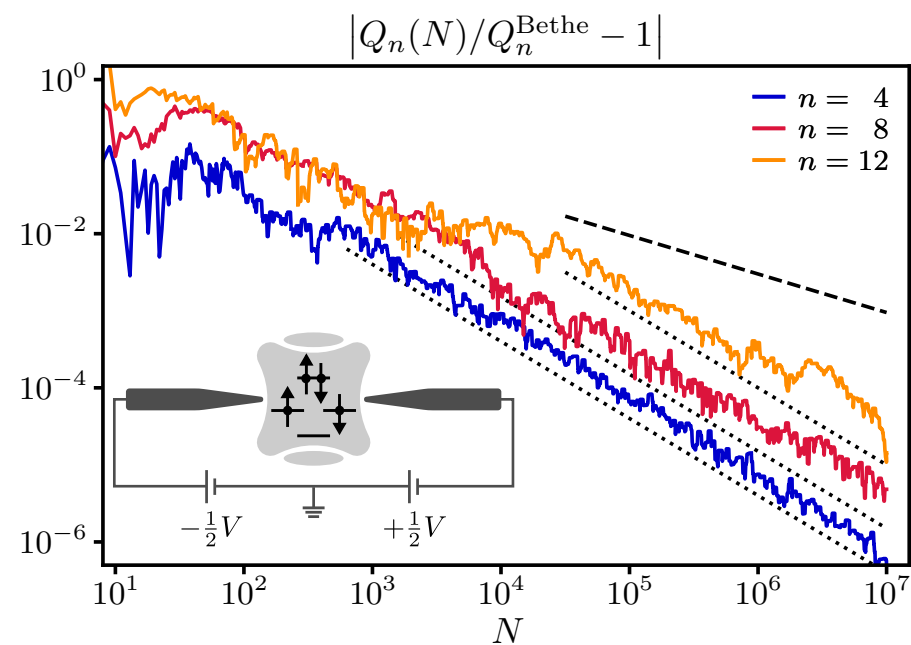
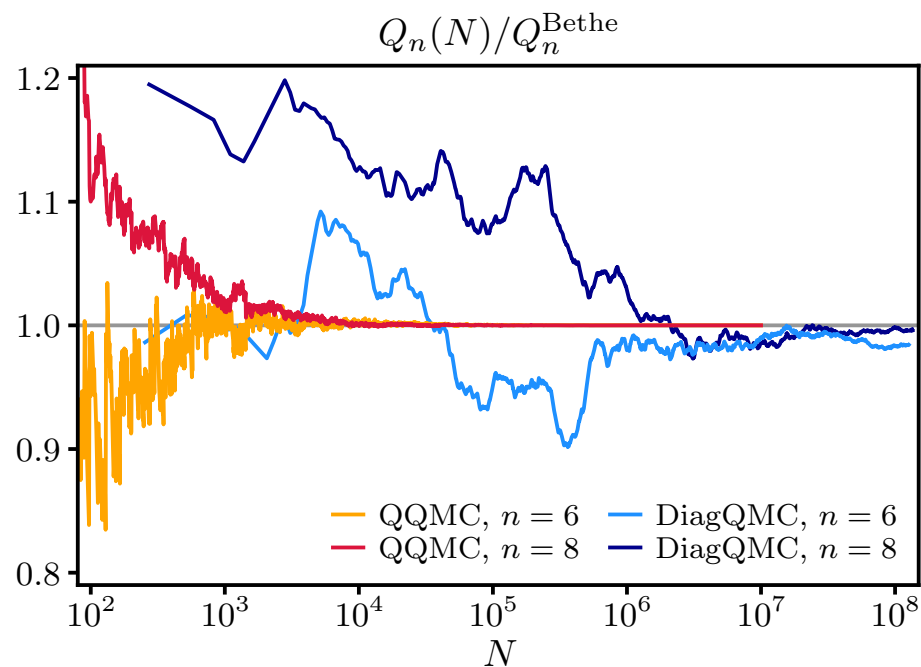


Coming up soon: 
$$p_n(\mathbf{u}) = h_a^{(1)}(v_1) h_{ab}^{(2)}(v_2) \cdots h_{cd}^{(n-1)}(v_{n-1}) h_d^{(n)}(v_n)$$

# USE QUASI RANDOM NUMBERS TO ESTIMATE THE INTEGRAL

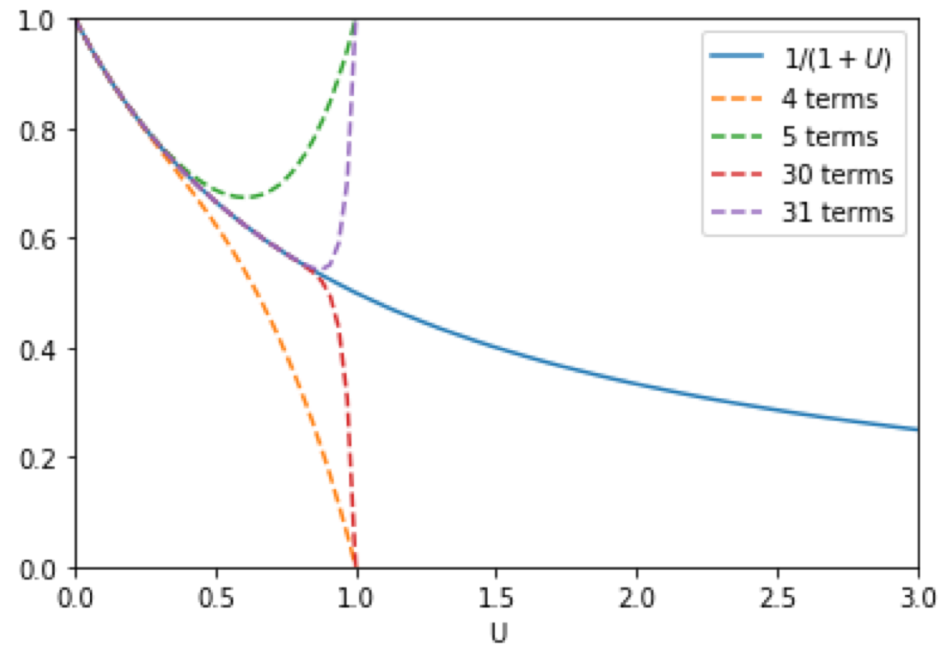
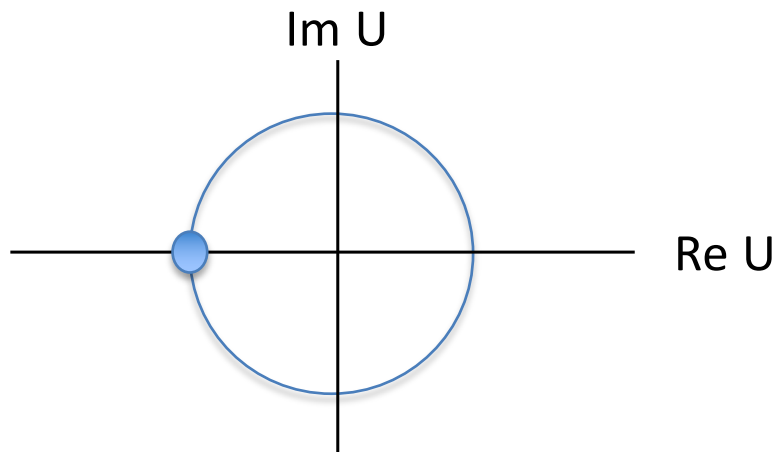


# ENJOY THE SPEED-UP



# PROBLEM #3: RECONSTRUCTION.

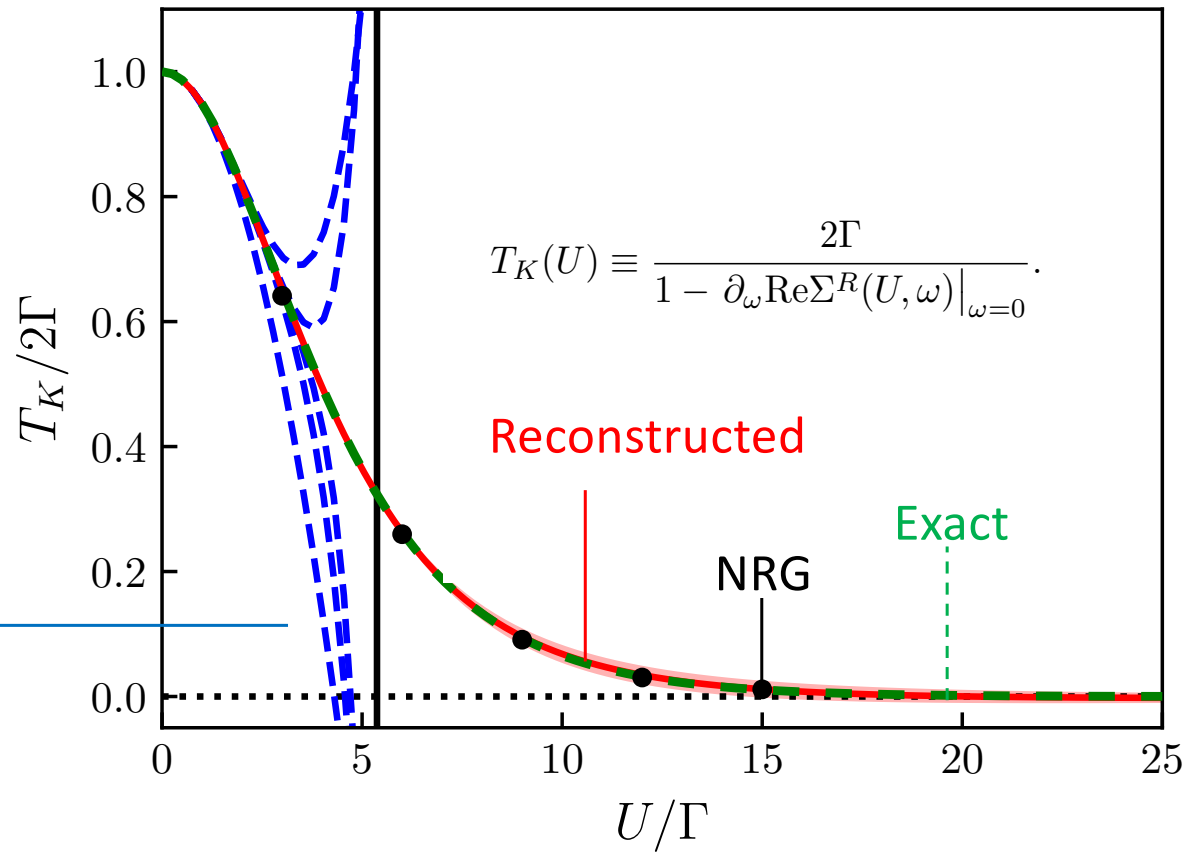
$$\frac{1}{1+U} = \sum_n (-1)^n U^n$$



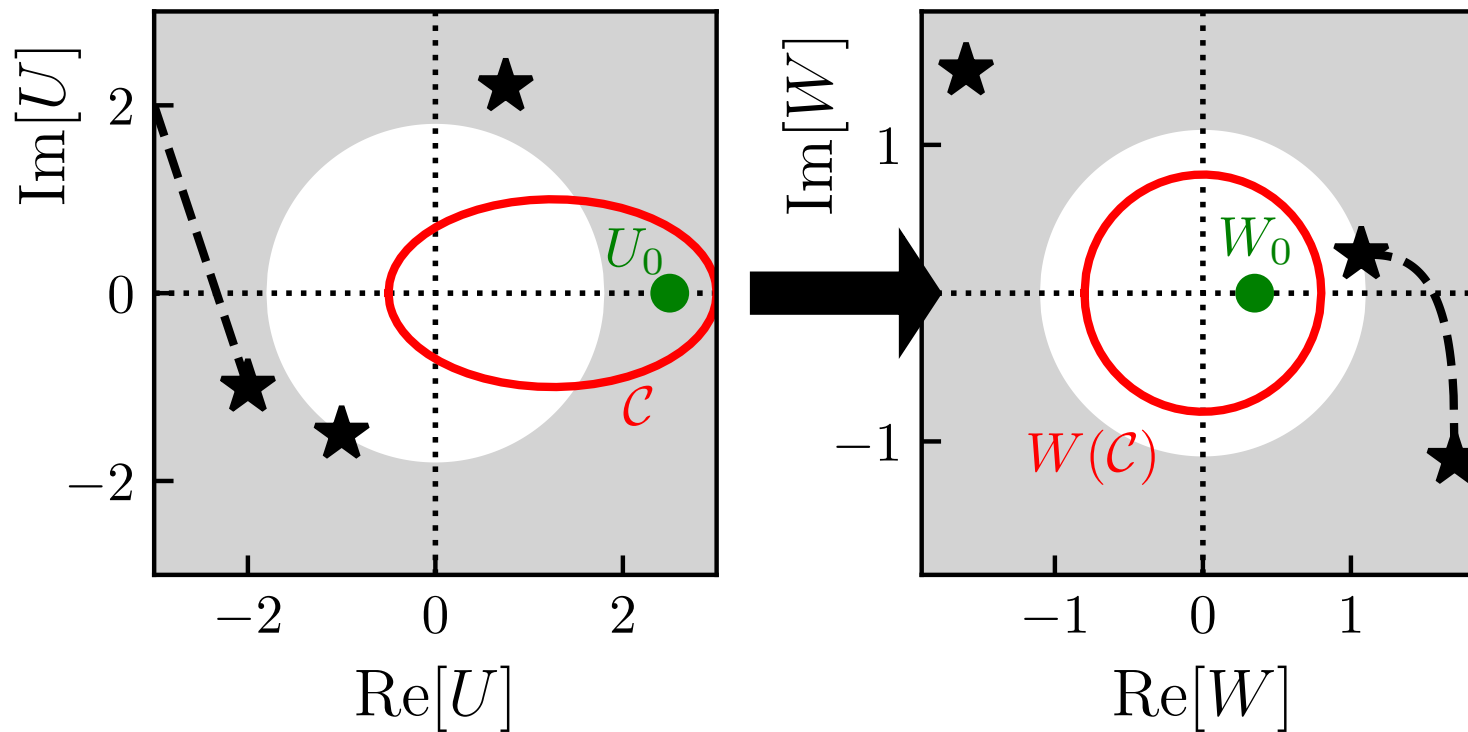
# PROBLEM #3: RECONSTRUCTING $F(U)$ FROM $F_N$

$$F(U) = \sum_n F_n U^n$$

Truncated sum

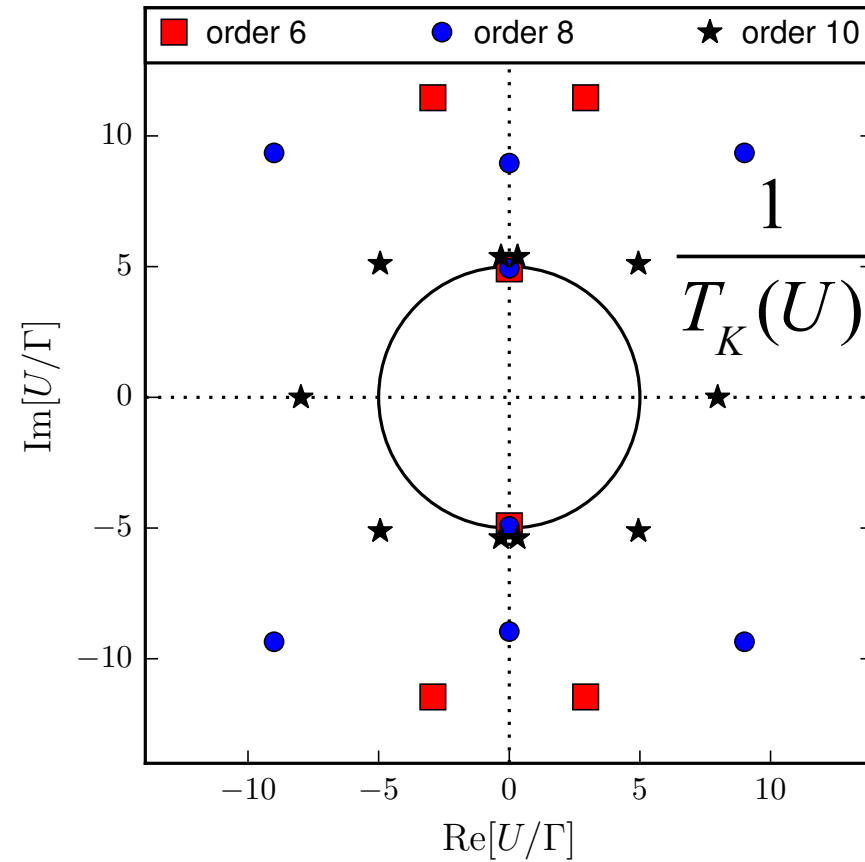
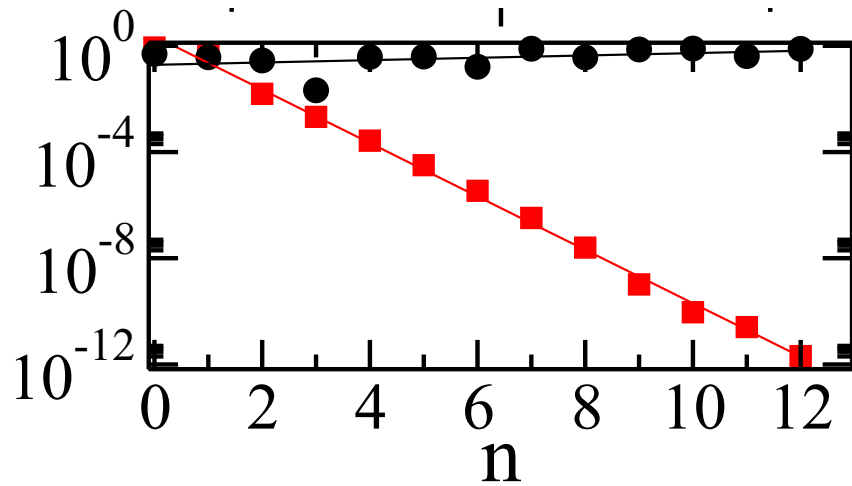


# SEPARATION PROPERTY HYPOTHESIS

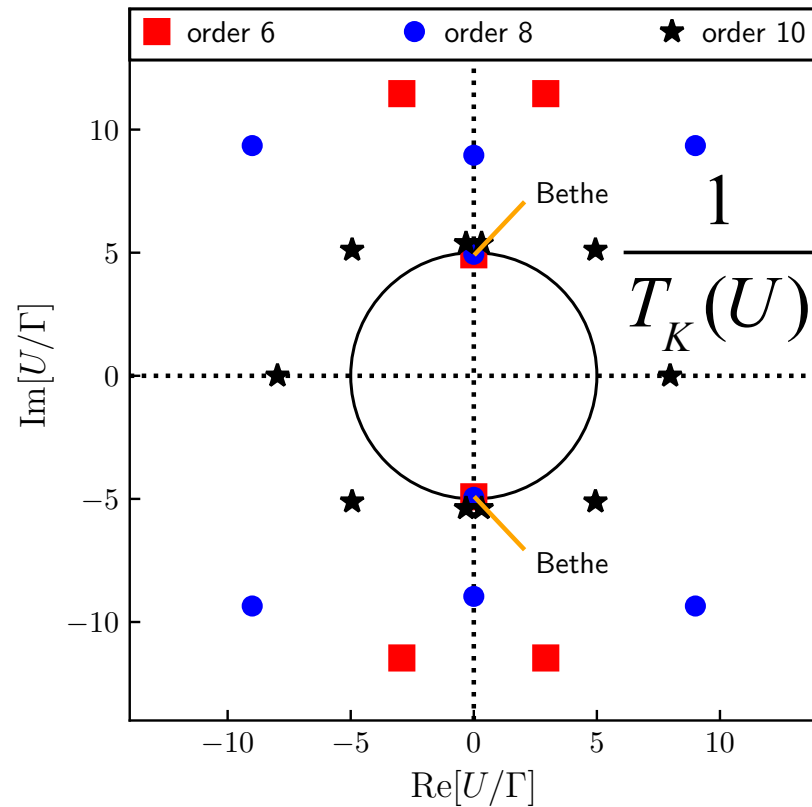
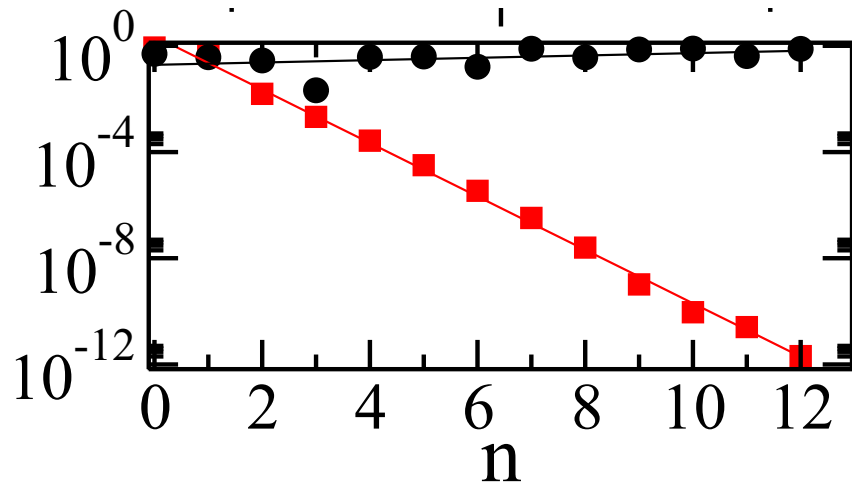




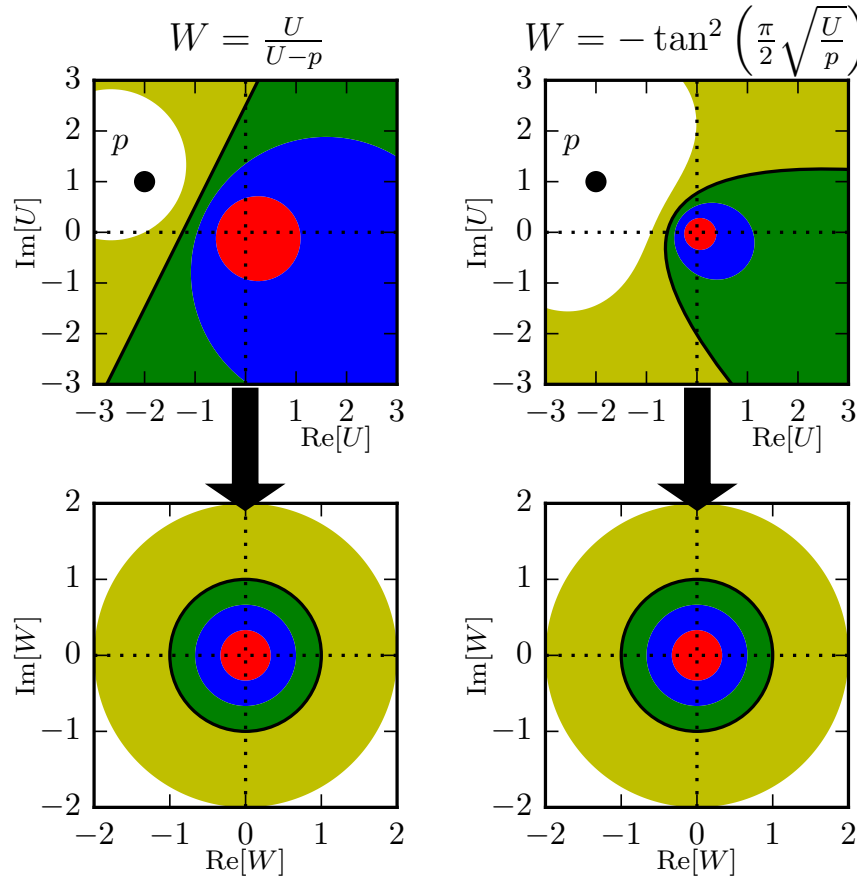
# 1) FIND THE POLES / ZEROS OF THE INVERSE



# 1) FIND THE POLES / ZEROS OF THE INVERSE



# 2) DESIGN THE CONFORMAL TRANSFORM



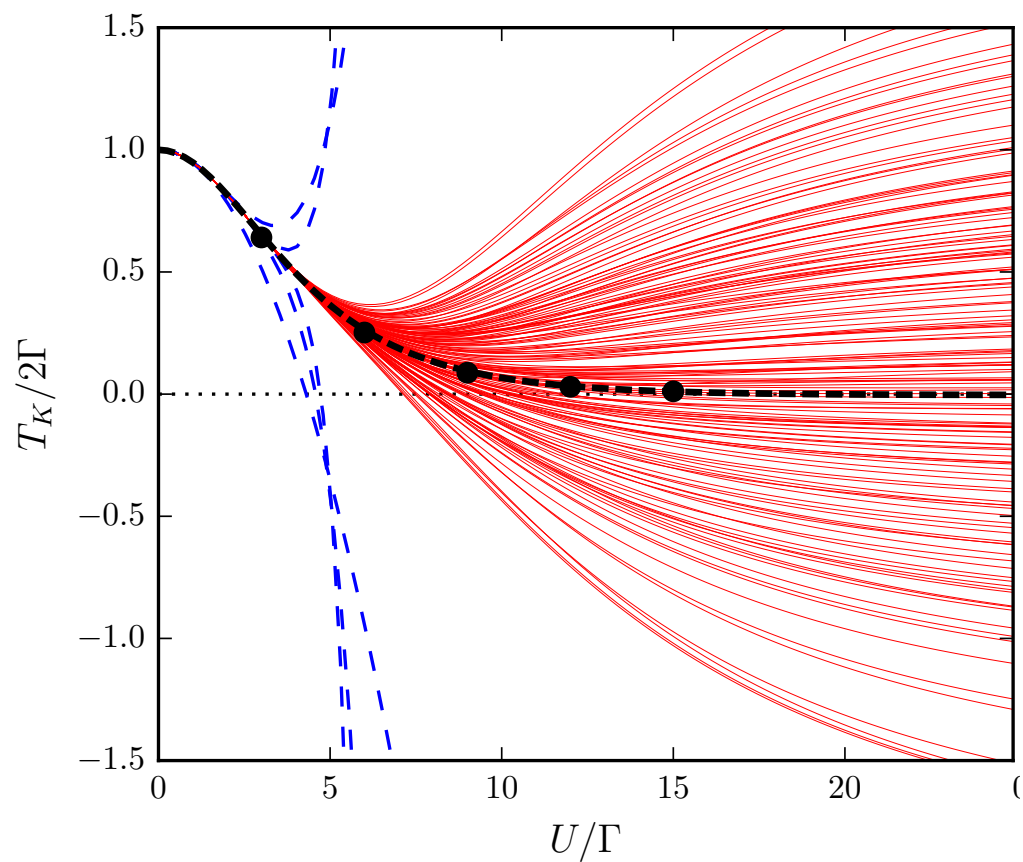
$$F(U(W)) = \sum_n \bar{F}_n W^n$$

$W(U)$ : exact

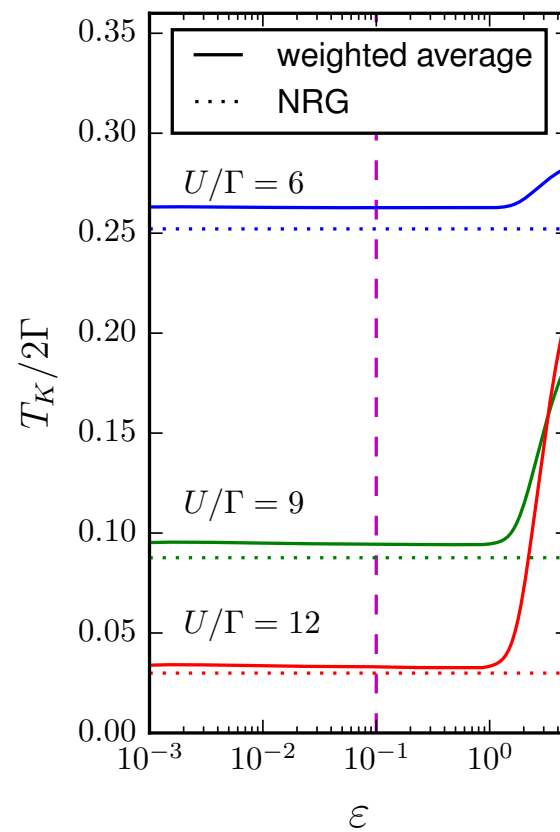
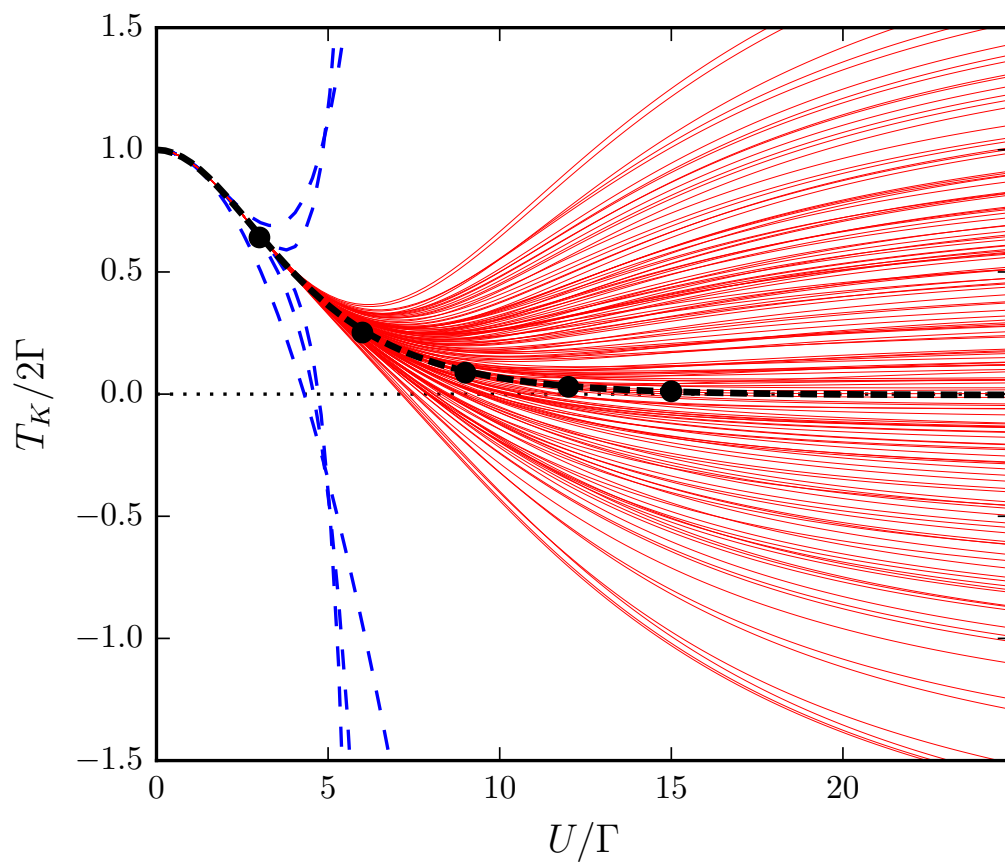
$U(W)$ : truncated

$$F(U) = \sum_n \bar{F}_n [W(U)]^n$$

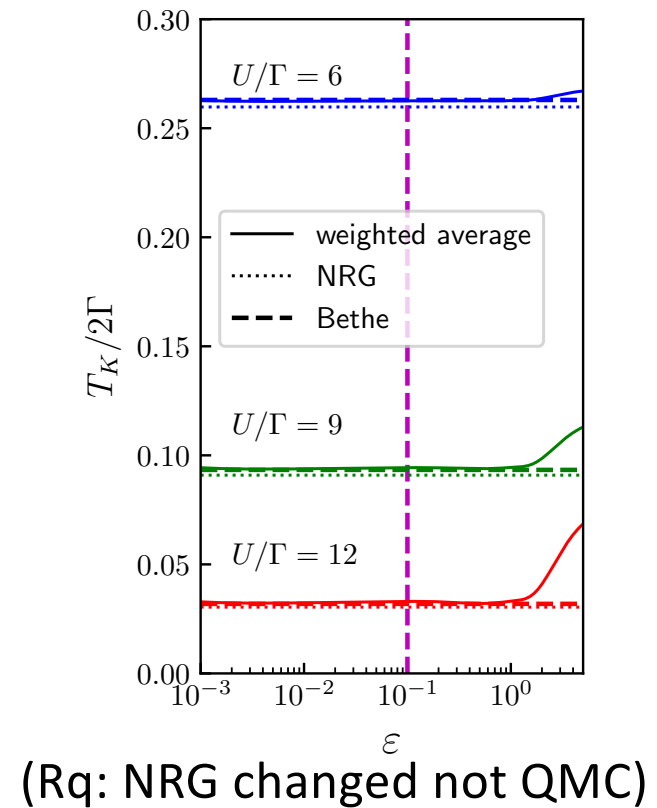
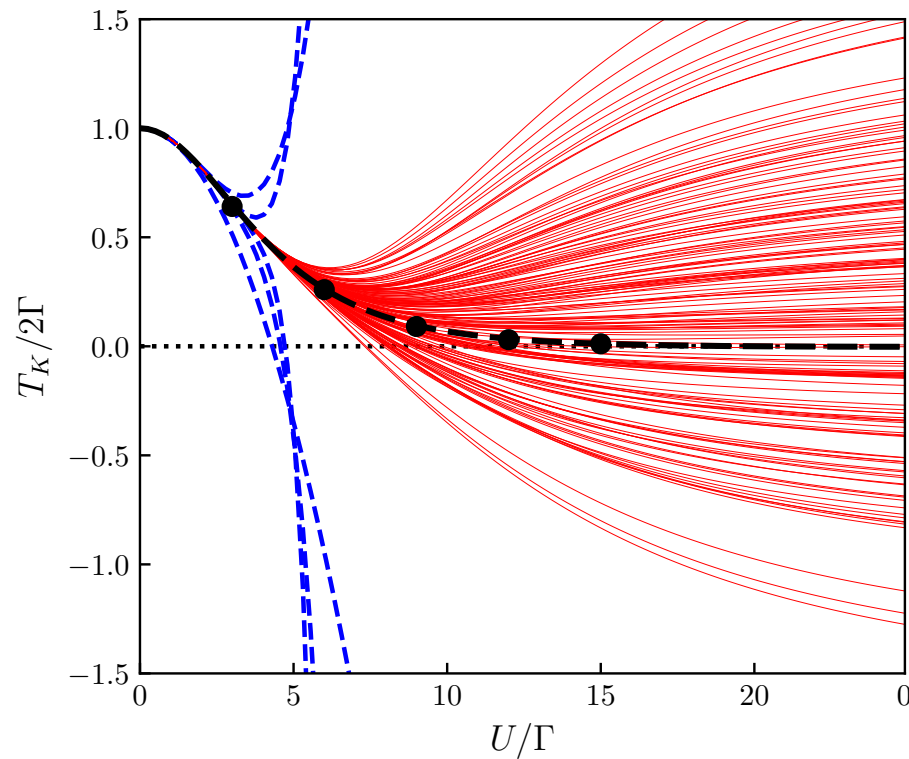
# 3) ADD BAYESIAN INFERENCE



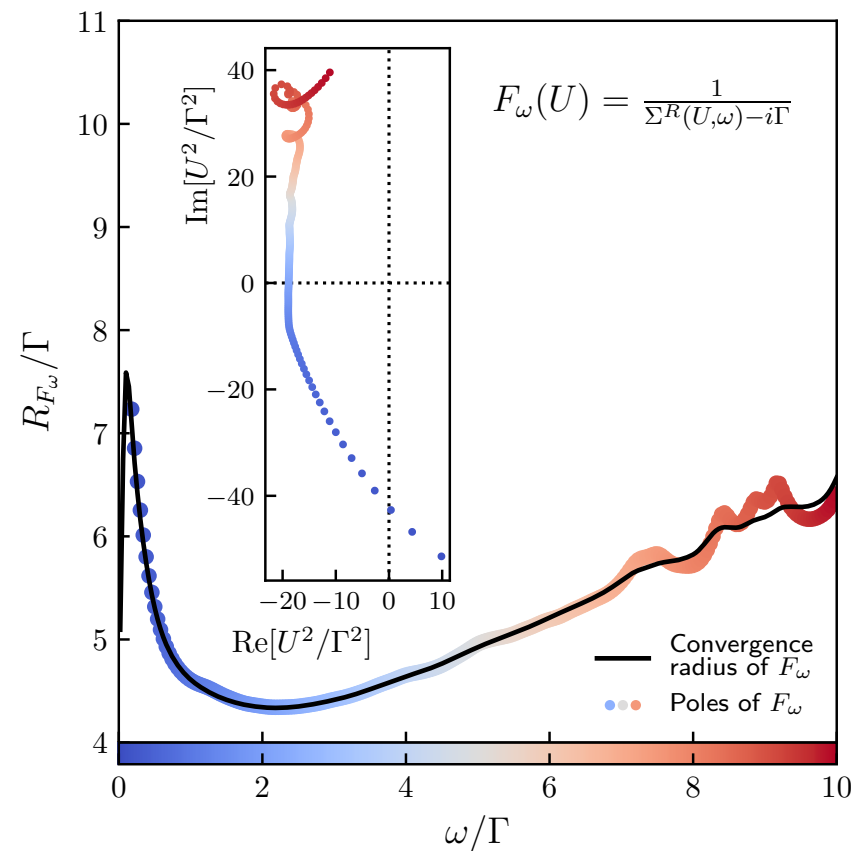
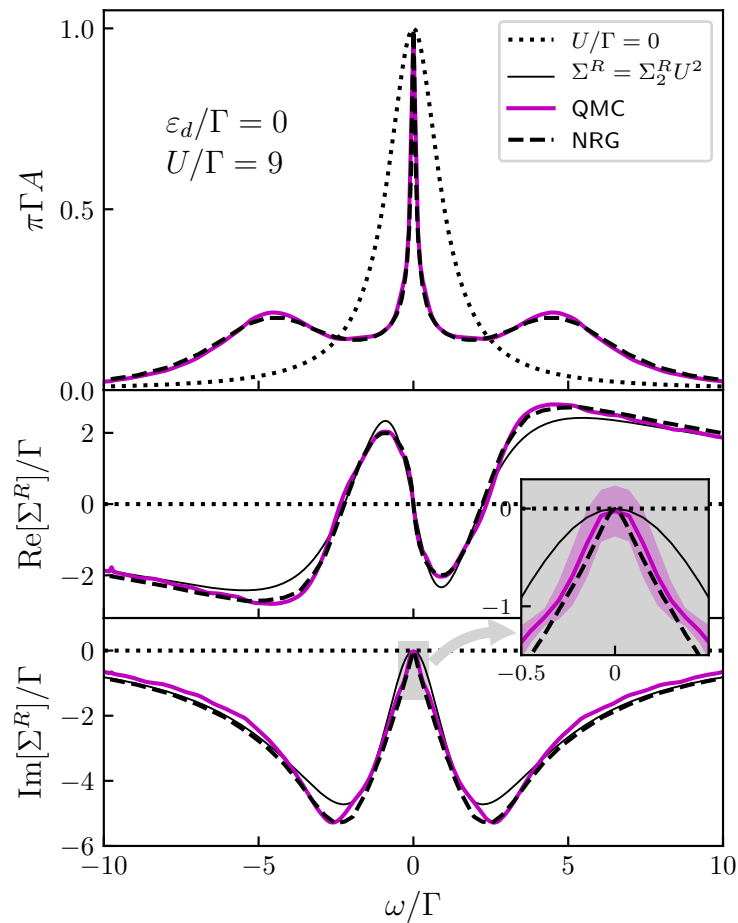
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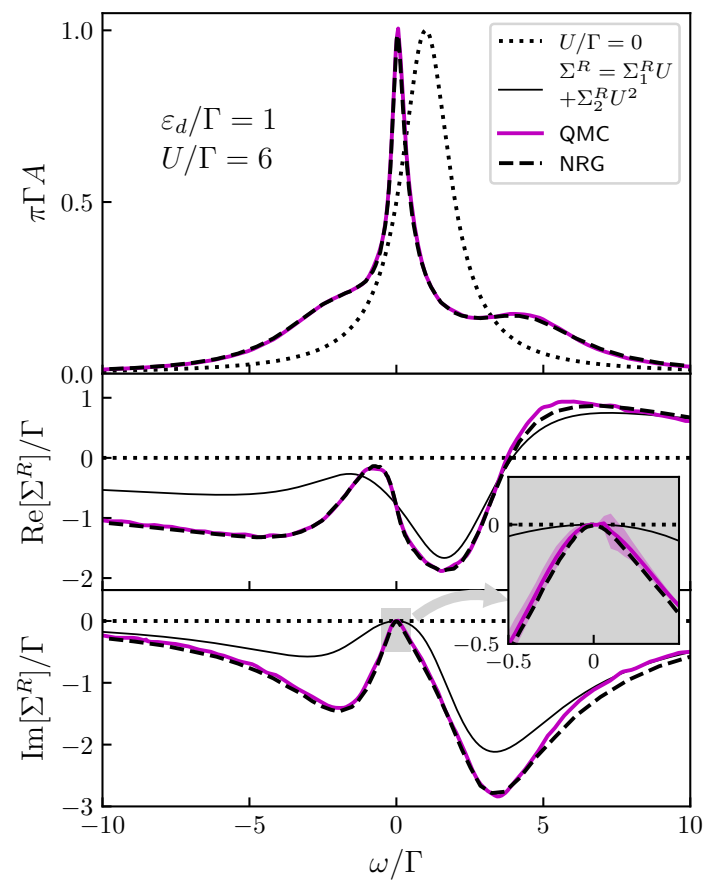
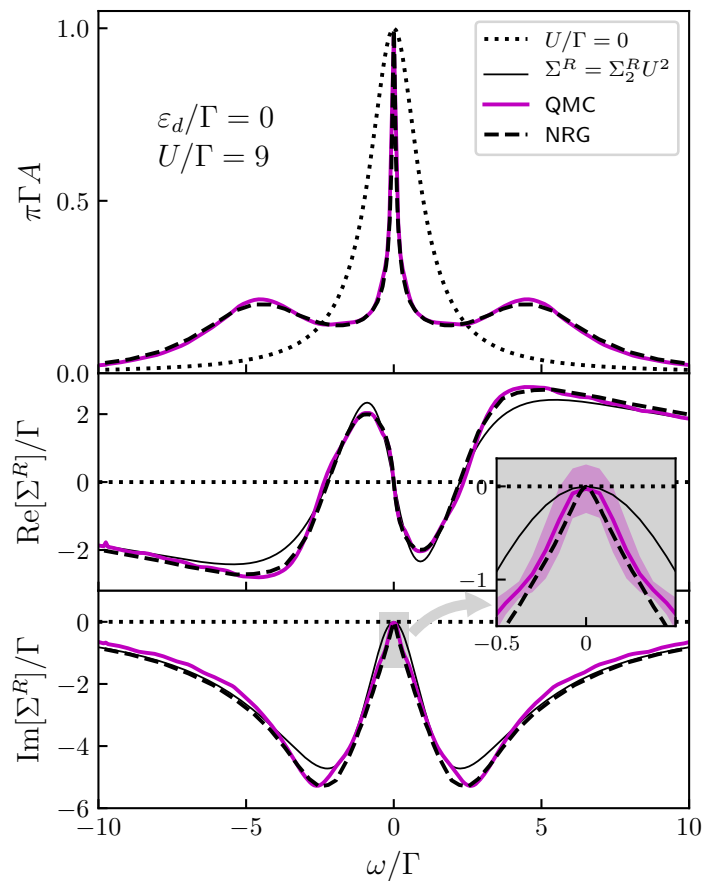
# 3) ADD BAYESIAN INFERENCE



# SOME EQUILIBRIUM BENCHMARKS

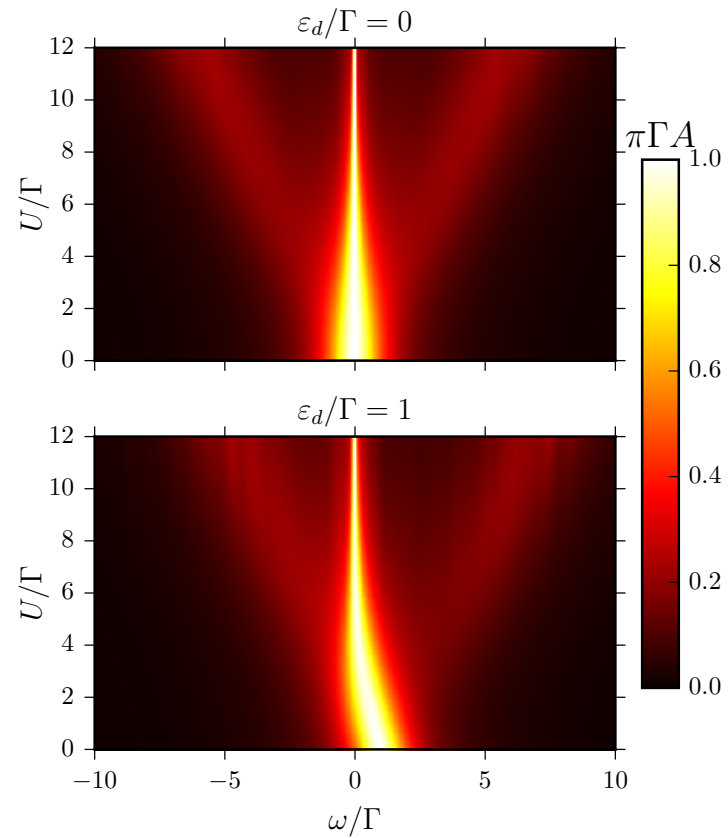


# SOME EQUILIBRIUM BENCHMARKS

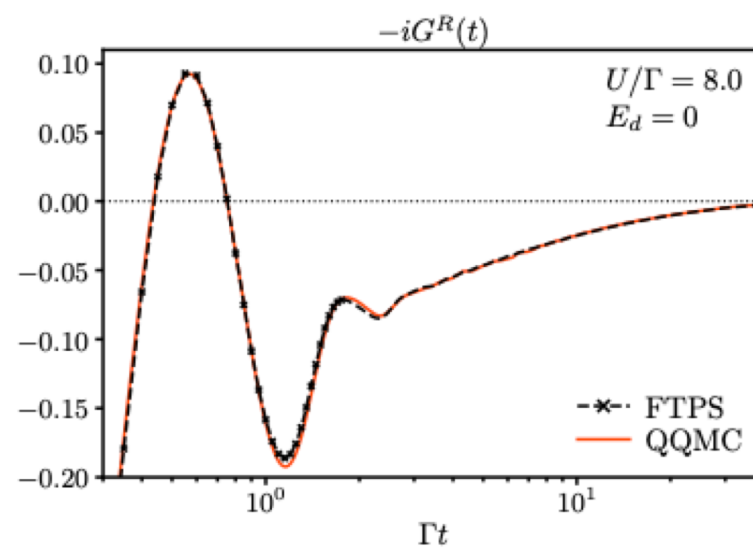
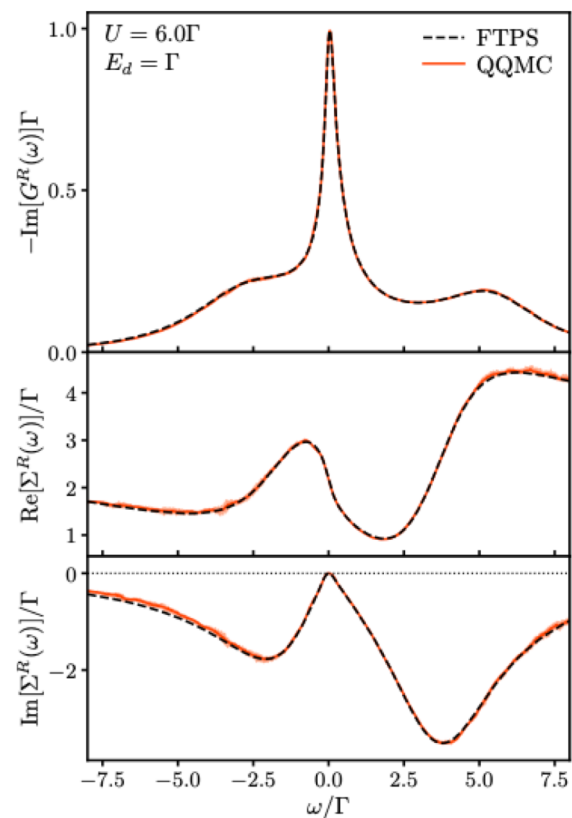
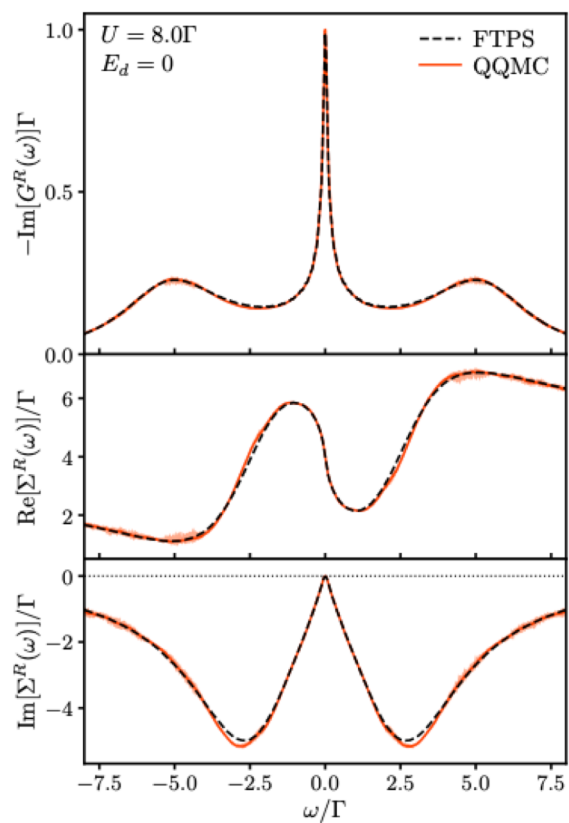




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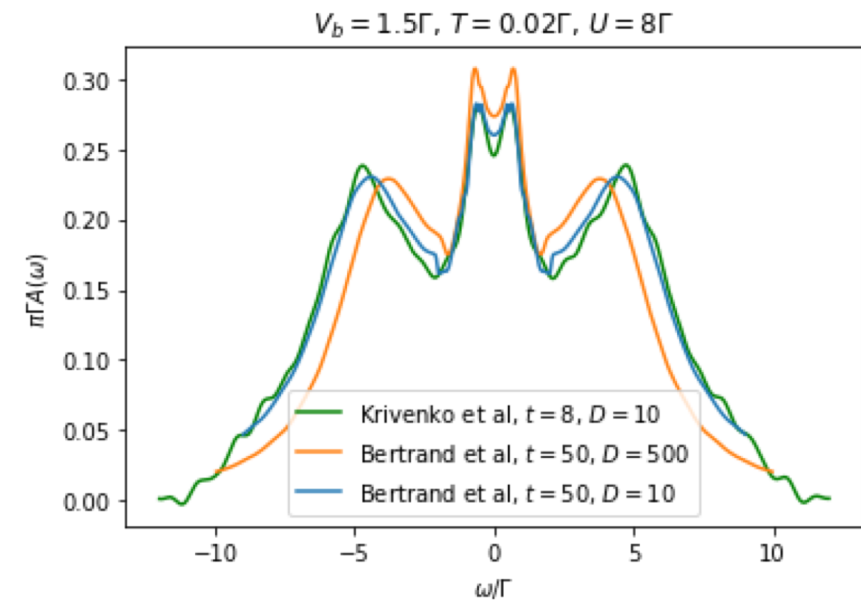
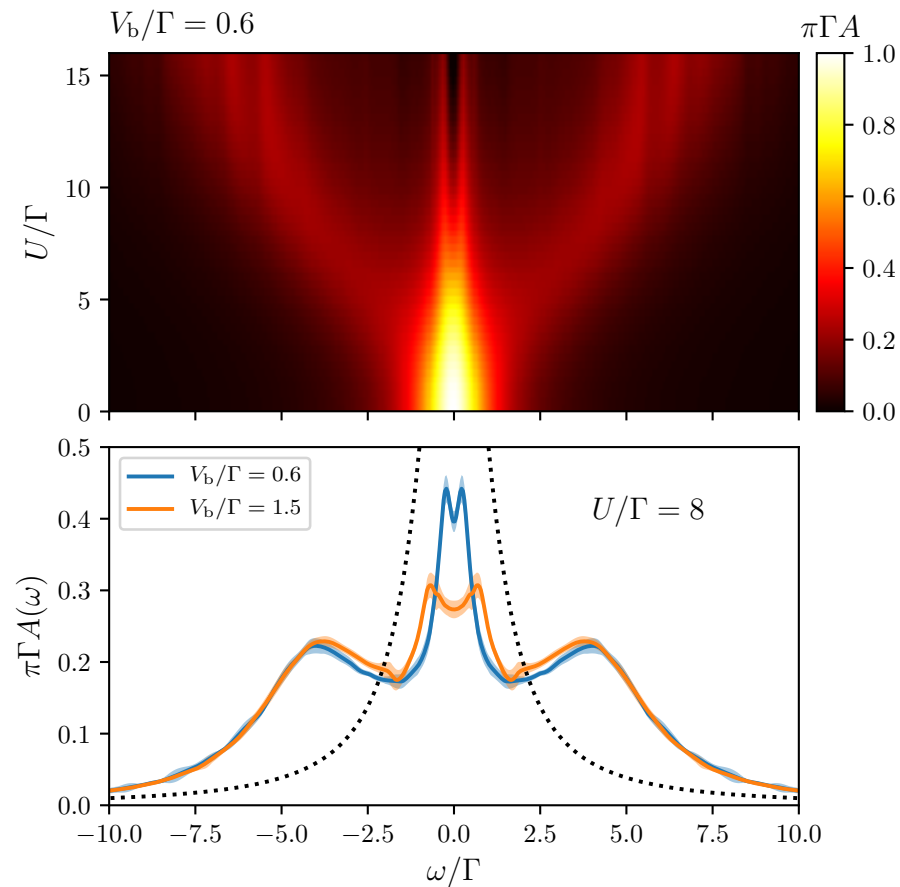


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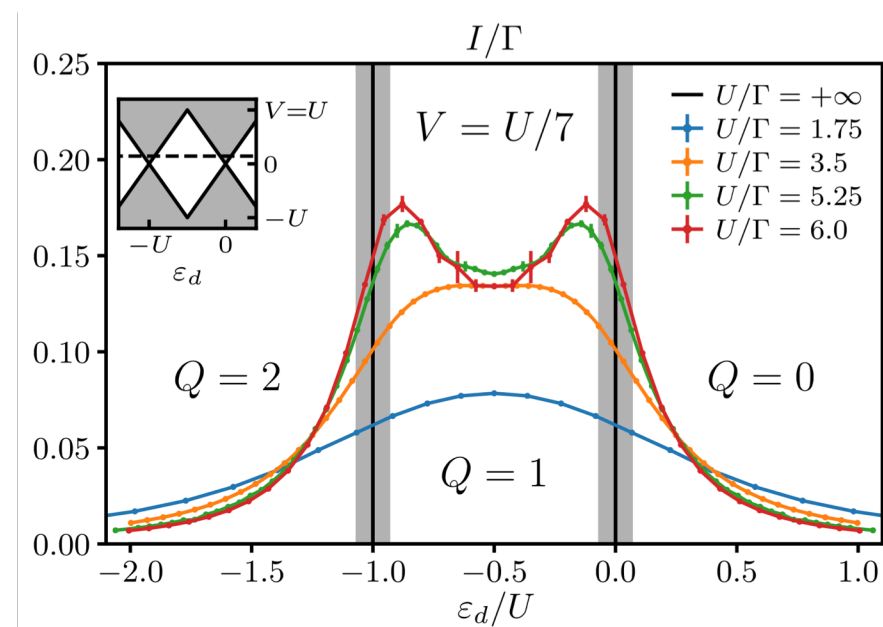
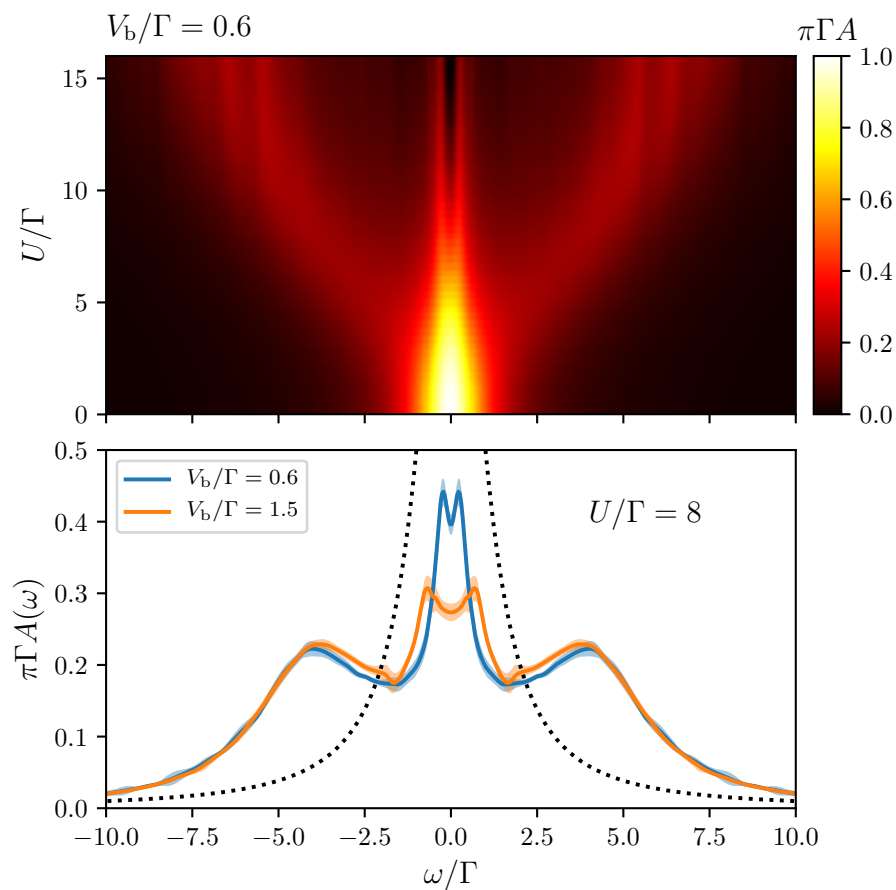
# AND BACK TO NON-EQUILIBRIUM

Igor Krivenko,<sup>1</sup> Joseph Kleinhenz,<sup>1</sup> Guy Cohen,<sup>2,\*</sup> and Emanuel Gull  
arXiv:1904.11527v1

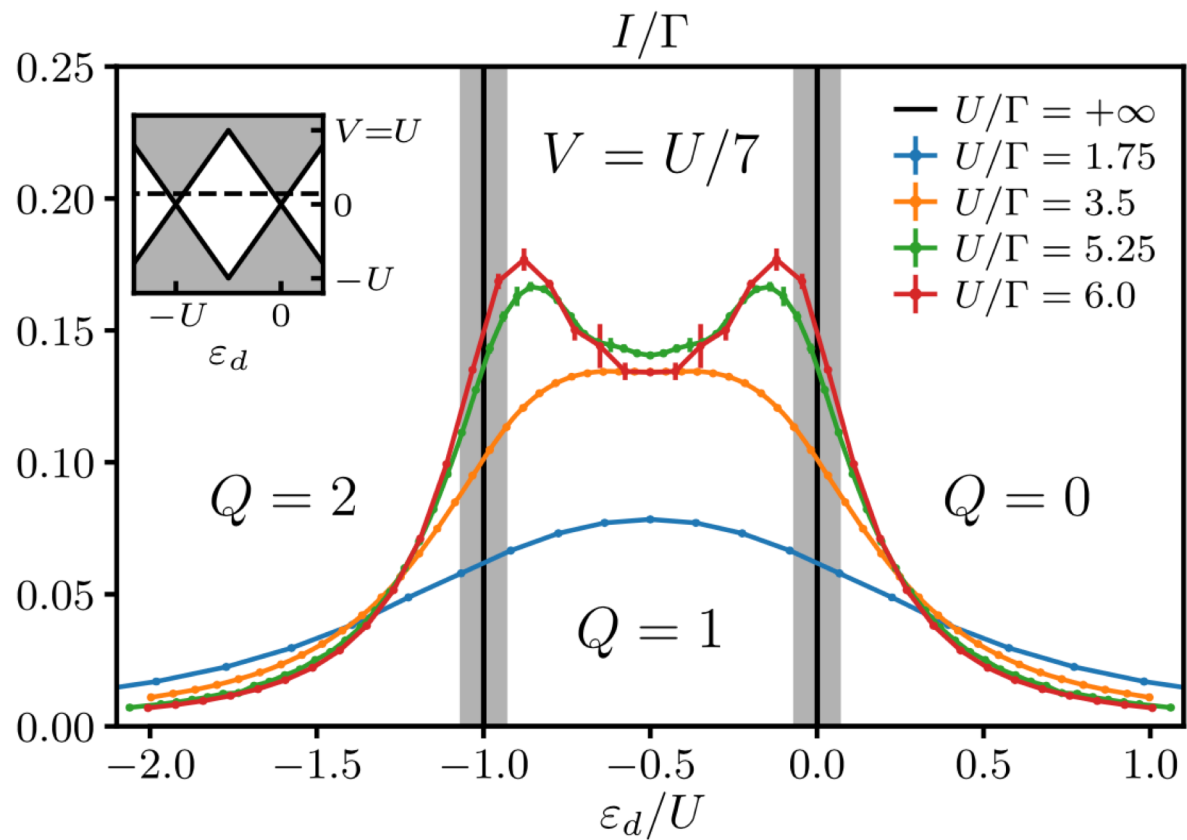


(PRELIMINARY COMPARAISON  
DIFFERENT BATH AND TIMES)

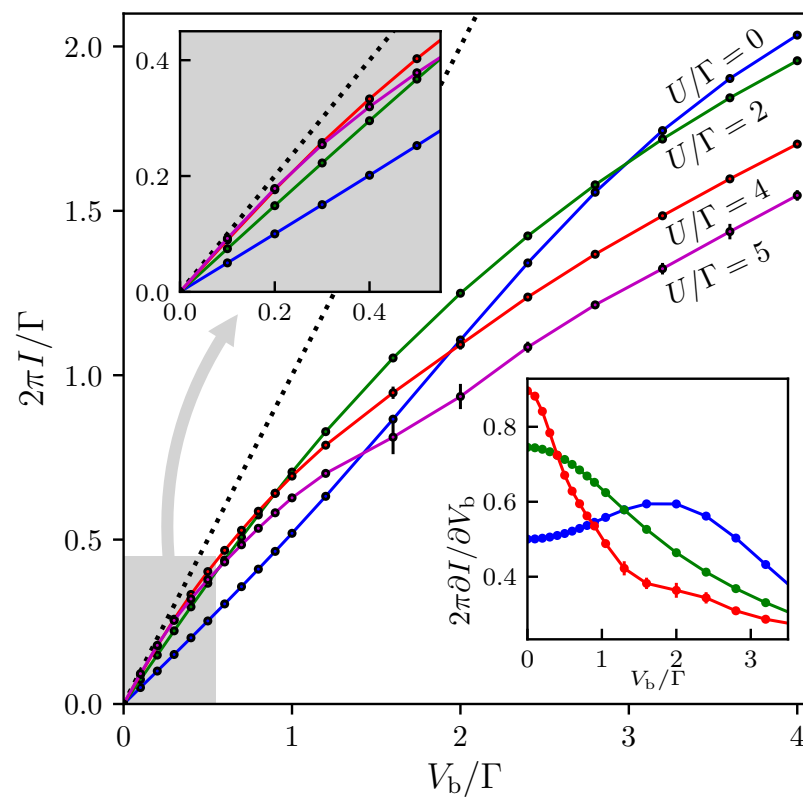
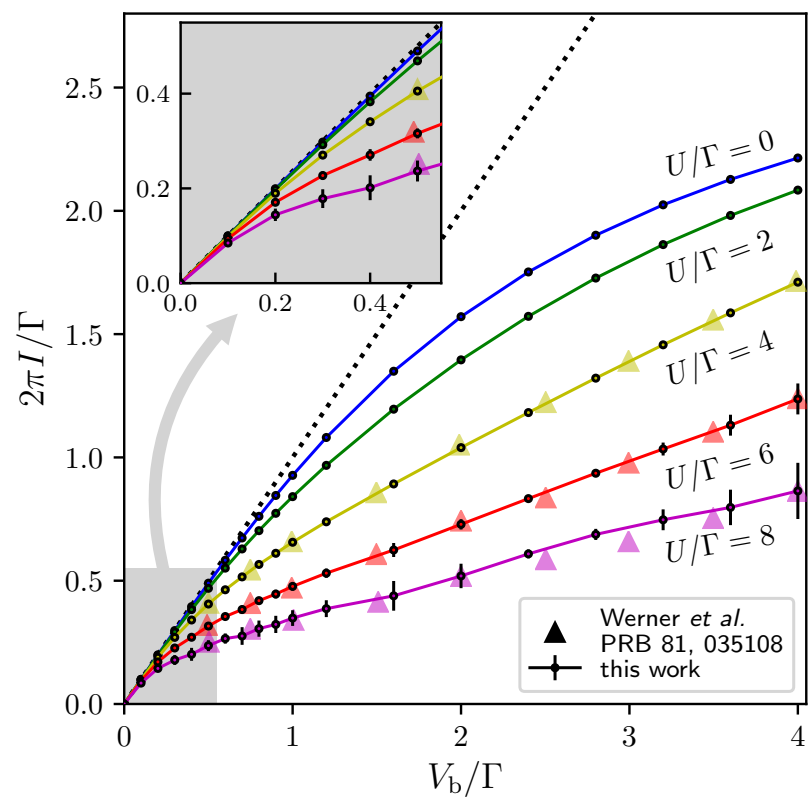
# AND BACK TO NON-EQUILIBRIUM



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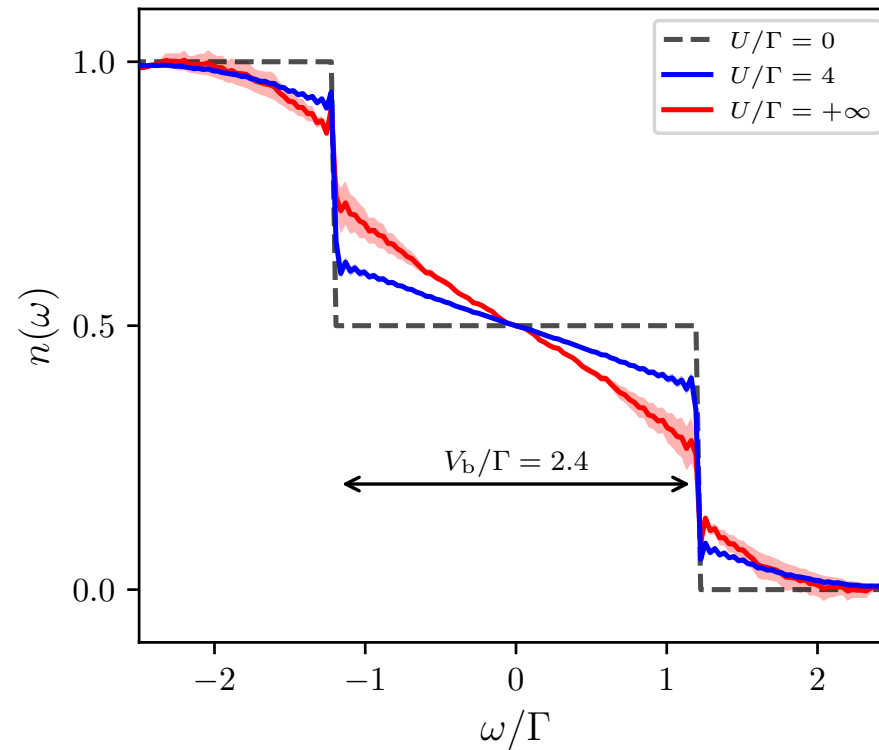


# AND BACK TO NON-EQUILIBRIUM



# NON-EQUILIBRIUM DISTRIBUTION FUNCTION

$$n(\omega) = \frac{iG^<(\omega)}{2 \text{Im}[G^R(\omega)]}$$



# CONCLUSION

## MORE TECHNICAL PROGRESS NEEDED

- Improved sampling & implementation (coming up soon)
- Terms and Counter-terms

## MORE APPLICATIONS

- Lattice (in progress)
- DMFT (also in progress)
- Nanoelectronics (0.7 anomaly)
- Few qubits & their baths
- Eventually an open source software

PHYSICAL REVIEW LETTERS **125**, 047702 (2020)

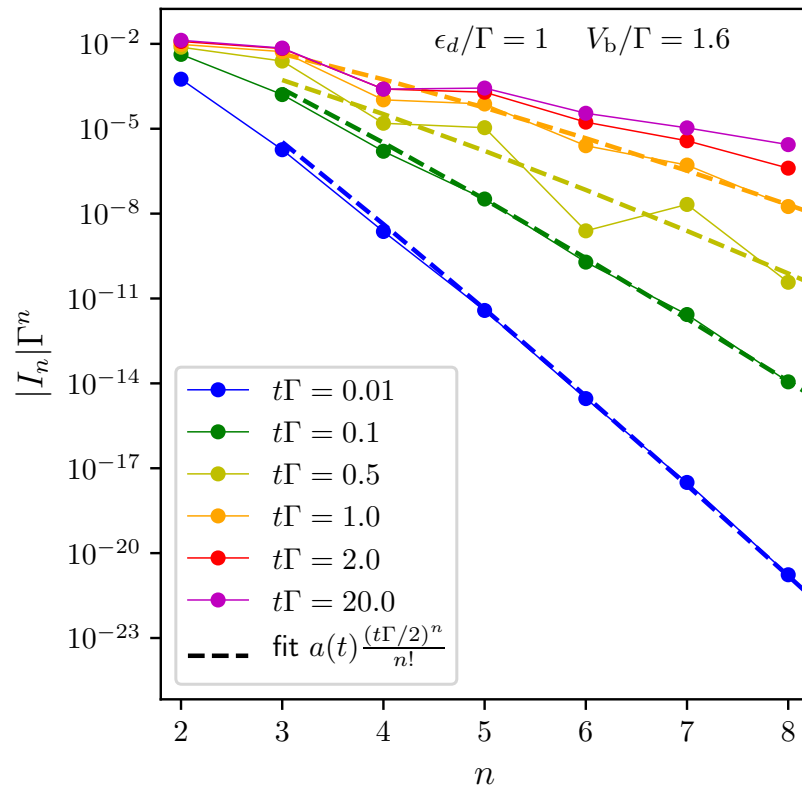
PHYSICAL REVIEW X **9**, 041008 (2019)

PHYSICAL REVIEW B **100**, 125129 (2019)

PHYSICAL REVIEW B **91**, 245154 (2015)



# A WARNING ON CONVERGENCE RADIUS



We work directly in the stationary limit

$$\lim_{n \rightarrow \infty} \lim_{t \rightarrow \infty} O_n(t) \neq \lim_{t \rightarrow \infty} \lim_{n \rightarrow \infty} O_n(t)$$