

Spacing Ratio Distribution

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- Or is it $t \rightarrow -t, \theta \rightarrow -\theta, p \rightarrow p$?
- The issue is the second definition is the one used by Lemarie et al in the “Universality of the Anderson Transition” paper (Lemarié, Grémaud, and Delande 2009).

The Problem

$$H = \frac{p^2}{2} + K \cos(\theta) \sum_n \delta(t - n\tau)$$

This hamiltonian is symmetric wrt $t \rightarrow -t$, $\theta \rightarrow -\theta$ i.e time reversal and wrt $\theta \rightarrow -\theta$, $p \rightarrow -p$ i.e. parity. The time reversal only holds if we consider only $\Delta t = N\tau$ but that is fine I guess.

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- ~~But even then, the distribution should be e^{-s} for the spacing. which is not the case.~~
- The Izrailev and Atas papers I was talking about last time were indeed on kicked rotors on a torus as you suspected.

Parity is a headache (cont.)

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- Maybe I'm missing something but apparently, eigenvector solving doesn't allow me to specify accuracy/tolerance anywhere.
- ~~All in all, I have no clue what is going wrong.~~

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- I found a few places where it was detailed and I managed to implement GUE, GOE, GSE, CUE, COE, CSE generators. (Mezzadri 2007; Edelman and Rao 2005; Sirca and Horvat 2012)
- It turned out that my program worked correctly for the unitary and orthogonal ensembles but failed for the symplectic ensembles because they show degeneracy.

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- The symplectic ensembles show Kramer's degeneracy due to the half integral spin. Because of that, each energy level has at least a double degeneracy (up and down spin for spin $\frac{1}{2}$).

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- And this worked for the earlier case of the quantum kicked rotor as well. Using the parity eigenvalue would be more accurate in some sense because we don't remove other symmetries that may exist that we don't know of, but it is not good numerically.

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- This method removes the degeneracies arising from all the symmetries, but if we are sure that there is only one symmetry this will work. By taking the ratio of number of unique eigenvalues to the original number of eigenvalues, we can find the mean degeneracy.

Results

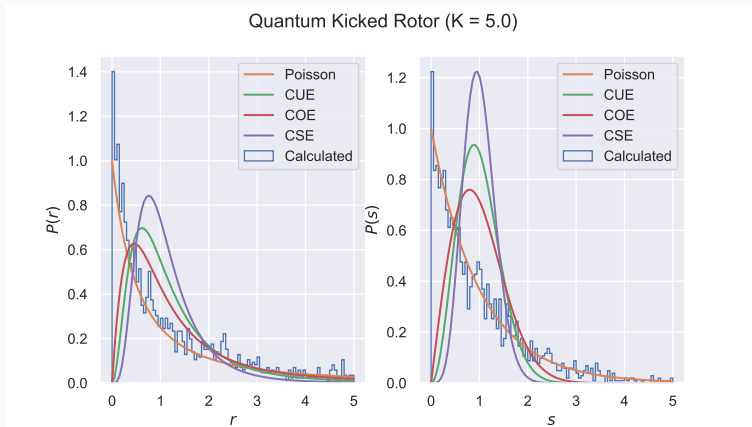


Figure 1: Standard kicked rotor with $K = 5$, p -basis $[-2000, 2000]$, 3991 eigenphases used and 10 discarded (tol = 0.1). Uniqueness ratio is 0.52 with tol = 10^{-8}

Quasiperiodic Kicked Rotor Systems

$$H = \frac{p_1^2}{2} + p_2\omega_2 + p_3\omega_3 + K\cos(\theta_1)(1 + \alpha\cos(\theta_2)\cos(\theta_3)) \sum_n \delta(t - n)$$

We use $\hbar = 2.85$, $\omega_2 = 2\pi\sqrt{5}$, $\omega_3 = 2\pi\sqrt{13}$ and the momentum basis is from -10 to 10. We consider a tolerance of 0.1 for discarding eigenvalues based on their modulus. We consider a tolerance of 10^{-8} for degeneracy.

Results (cont.)

For large K values, the system nicely fits into the poisson distribution.

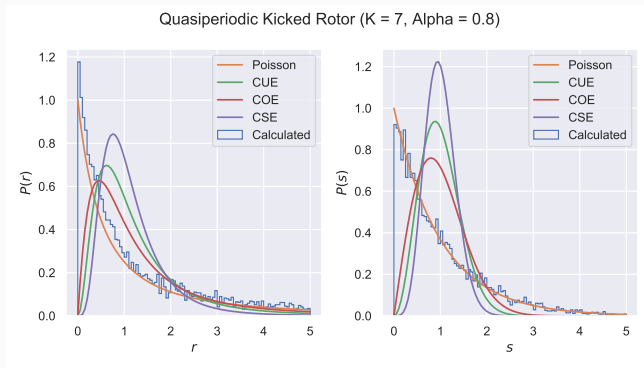


Figure 2: Post-critical (Metallic): $K = 7$, $\alpha = 0.8$ with 6934 eigenphases used. No degeneracy, but 2327 phases were discarded as they were out of the $[0.9, 1.1]$ range

Results (cont.)

At critical point, things seem okay, but there is a noticeable difference at zero.

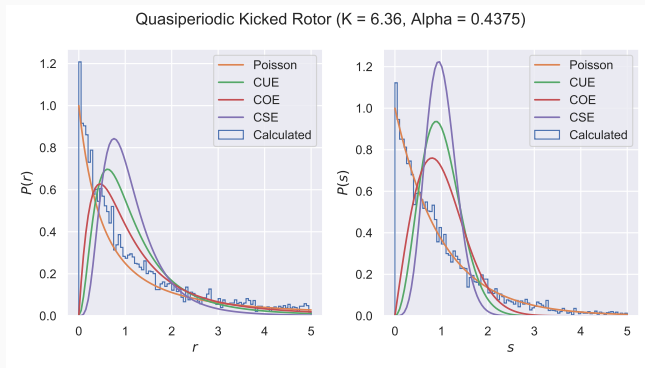


Figure 3: Critical: $K = 6.36$, $\alpha = 0.4375$ with 7874 eigenphases. No degeneracy, but 1387 phases were discarded for being out of range.

Results (cont.)

For small K values, things get weird...

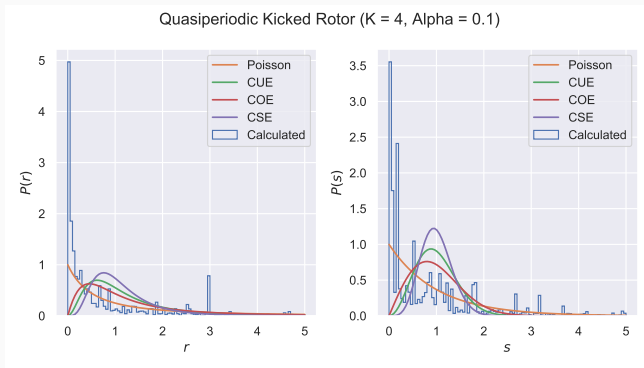


Figure 4: Pre-critical (Insulator): $K = 4$, $\alpha = 0.1$ with 8379 eigenphases. No degeneracy, but 882 phases were discarded for being out of range.

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Verification

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- I tried plotting the energy, entropy, p_1 -distribution plots for the matrix. But the dimensionality is too low for me to actually see anything. It does look pretty similar to the corresponding plots made by the direct method, but there are slight differences.

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- I tried plotting the energy, entropy, p_1 -distribution plots for the matrix. But the dimensionality is too low for me to actually see anything. It does look pretty similar to the corresponding plots made by the direct method, but there are slight differences.
- Since I am trusting the direct method, I thought of generating the matrix using that and comparing the two. Now what I'm essentially doing in this is calculating $F\mathbf{e}_j$ and comparing it with the j th column of the matrix generated.

- The two are not as close as one would hope it to be but the error seems to decrease with the size of the matrix. The maximum size I could do on my laptop was $[-15, 15]$. Unsurprisingly, the difference is maximum near the edges both row-wise and col-wise.

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- I also tried generating random states and comparing the difference between the generated outcomes, but I thought the direct matrix comparison was a better metric.
- Matching tolerances: $\text{atol} = 10^{-5}$, $\text{rtol} = 10^{-8}$

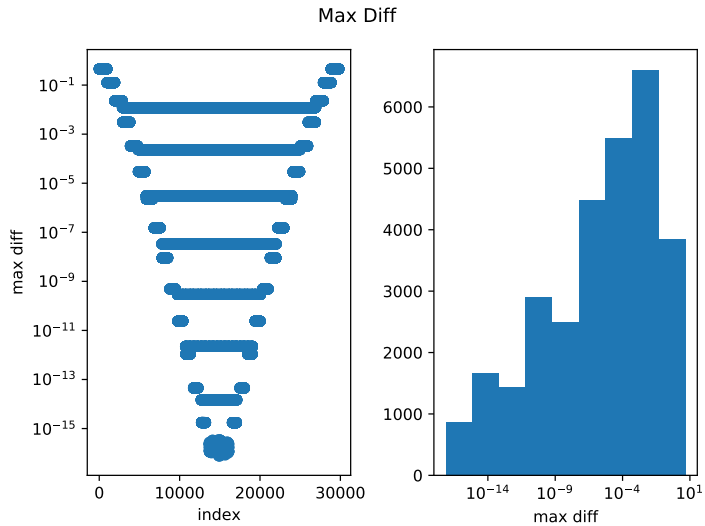


Figure 5: $K = 3$, $\alpha = 0.1$

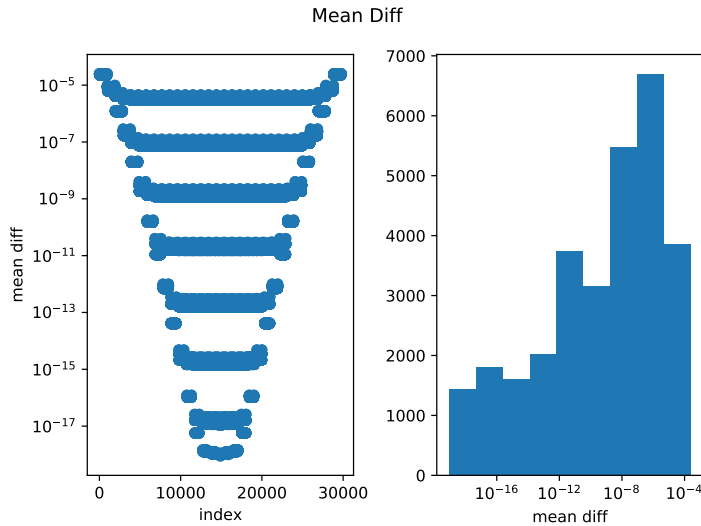


Figure 6: $K = 3$, $\alpha = 0.1$

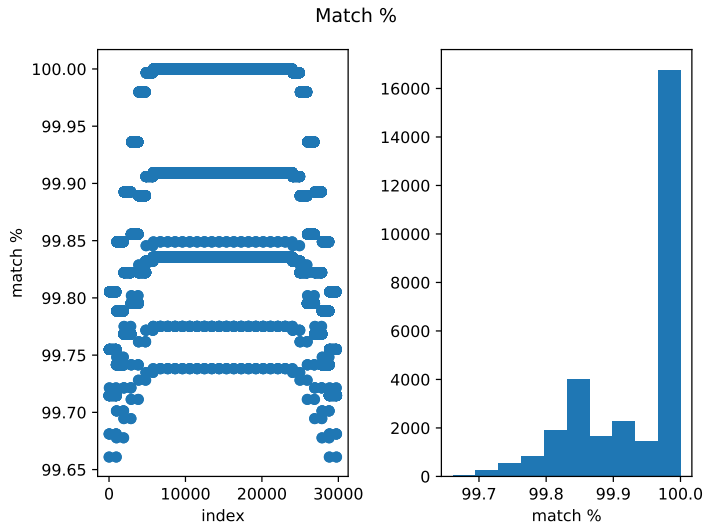


Figure 7: $K = 3$, $\alpha = 0.1$

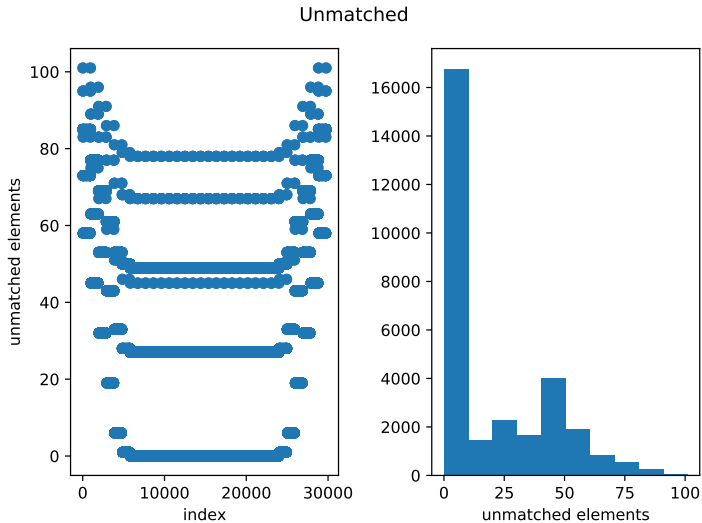


Figure 8: $K = 3$, $\alpha = 0.1$

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