```
In[*]:= ClearAll["Global`*"]
In[*]:= (* Set the notebook directory to the local directory *)
    SetDirectory[NotebookDirectory[]];
    (* Check that this is the directory *)
    NotebookDirectory[]
```

Overview of the program

In[@]:= (*

This notebook computes the "quasinormal modes" associated with perturbations of holographic superconductors. This simple example illustrates a few numerical techniques including pseudospectral methods to solve differential equations as well as finding eigenvalues. This is a version of the numerics in https:// arxiv.org/pdf/2212.10410 which rescales the U(1) gauge field $A_M \rightarrow$ $A_{\rm M}/Q$ and the complex scalar field $\psi \rightarrow \psi_{\rm M}/Q$ with $Q \rightarrow \infty$. In this limit, the spacetime is fixed and only the matter fields have nontrivial behavior. This is equivalent to suppressing energy and momentum fluctuations in the superconductor and only allowing charge fluctuations. The hydrodynamic modes $\omega(k)$ can be matched exactly including both real and imaginary parts in the limit k << 1, however for the sake of readability, I will only match the velocity of the second sound mode: $\lim_{k\to 0} \omega(k)$. To obtain the imaginary components, one needs the low frequency response functions and then uses Kubo formulae-the numerical techniques for these are very similar to what is shown here, so I have omitted this step. If you would like to see how that works, let me know and I will append it to the program.

*)

(*

There are five main parts to the program which can each be run independently or the whole notebook can be run from the top.

1.) In the first part we derive the equations

of motion describing holographic superconductors in equilibrium and then subject to a linearized perturbation.

- 2.) In the second part we solve the equations of motion for the background solution. The asymptotic behavior of these solutions defines sources and expectation values for the electric current < $\mathtt{J}^{\mu}\mathsf{>}$ and the order parameter $<\!\psi^*\psi\!>$. This is a boundary value problem for a coupled set of ordinary differential equations.
- 3.) In the third part we solve the linearized equations of subject to the boundary condition that the sources vanish. Since response functions are formally given by $G_{00}^{R}(\omega,k)$ = $\frac{\delta < 0>}{\delta s_0}$ where s_0 is the source for the operators < 0> , solutions to these equations give the poles of the response functions. They exist at limited values of $\omega(k)$ which define the dispersion relations of hydrodynamic fluctuations in a superconductor. Numerically, this means that we must find the $\omega(k)$, i.e. we solve an eigenvalue problem. The matrices are dense so there is no way to make diagonalization faster and Mathematica' s built in eigensolver is sufficiently efficient. However, in step 4, we illustrate a way to isolate an individual eigenvalue and increase precision and accuracy.
 - 4. We show how to efficiently increase accuracy for a single eigenvalue.
- 5.) Finally, we compare to the predictions of the hydrodynamic theory and show that the results agree.

*)

```
In[*]:= (* More details on part 2.
```

```
To solve the ODEs, we will use a version of Newton's method. The
 essential idea is that we have an equation we want to solve,
which we may write E_{i}[S^{i}] = 0 for some configuration of the fields S^{i},
but we are at a different value X^{i}(t=0). As with Newton's method of root finding,
from step t to step t+1,
we choose the next X^{i}(t+1) such that E_{i}'[X^{i}(t)](X^{i}(t+1)-X^{i}(t)) =
 -E_{j}[X^{i}(t)]. The fields X^{i} are a vector so this
  can be solved with built in linear algebra tools.
   Mathematica is competitive with other
  programming languages for linear algebra, see https://
 reference.wolfram.com/language/tutorial/LinearAlgebraInMathematicaOverview.html.
      At each step, then, we solve M^{ij}(t) \delta X^{j}(t) = -E^{i}(t) with M^{ij}(t) =
  \frac{\delta E^i}{\delta X^j} (t) until we reach a point where \delta X^i is below some threshold.
```

In[⊕]:= (* More details on part 3.

The linearized equations of motion have the form $\sum_{n} \omega^{n} C^{i}{}_{j}(n) \delta X^{j} = 0$, where $C^{i}{}_{j}$ are differential operators (that include the boundary conditions) acting on the linearized fields δX^{j} and ω is the frequency of the plane wave perturbation. A simple example which has only one power of ω can be written,

$$C_{i}^{i}(0) \delta X^{j} = -\omega C_{i}^{i}(1) \delta X^{j}$$

which has the form of a generalized eigenvalue problem with matrices $C^{i}_{j}(0)$ and $C^{i}_{j}(1)$. Generalized eigenvalue equations can be solved efficiently with Mathematica. When there are more powers of ω , the equations still have the form of a generalized eigenvalue problem, but in terms of a modified eigenvector. For instance, if the largest power of ω is 2, then,

$$\begin{pmatrix} C^{i}_{j}\left(1\right) & C^{i}_{j}\left(0\right) \\ -\mathbf{I} & 0 \end{pmatrix} \begin{pmatrix} \omega\delta\mathbf{X} \\ \delta\mathbf{X} \end{pmatrix} = -\omega \begin{pmatrix} C^{i}_{j}\left(2\right) & 0 \\ 0 & \mathbf{I} \end{pmatrix} \begin{pmatrix} \omega\delta\mathbf{X} \\ \delta\mathbf{X} \end{pmatrix} ,$$

returns the same linearized equation but increases the size of the matrices by a power of 4. If higher powers of ω appeared we would continue this process,

$$\begin{pmatrix} C^{i}{}_{j} \left(nmax-1 \right) & C^{i}{}_{j} \left(nmax-2 \right) & \dots & C^{i}{}_{j} \left(0 \right) \\ -I & 0 & \dots & 0 \\ - & -I & \dots & 0 \\ 0 & 0 & \dots & 0 \end{pmatrix} \begin{pmatrix} \omega^{n-1} \delta X \\ \omega^{n-2} \delta X \\ \dots \\ \delta X \end{pmatrix} = -\omega \begin{pmatrix} C^{i}{}_{j} \left(nmax \right) & 0 & \dots & 0 \\ 0 & I & \dots & 0 \\ 0 & 0 & \dots & 0 \\ 0 & 0 & \dots & I \end{pmatrix} \begin{pmatrix} \omega^{n-1} \delta X \\ \omega^{n-2} \delta X \\ \dots \\ \delta X \end{pmatrix}.$$

*)

```
In[*]:= (* More details on part 4.
     In part 4, we increase the grid size to get better accuracy. However,
    this drastically increases the size of the matrices we would need to
     diagonalize. In general, we care about only the lowest lying eigenvalues,
    so diagonalizing the full matrix is overkill. The most efficient route is to
     promote \omega to a function of z, \omega \rightarrow \omega[z], with an equation of motion \omega'[z] = 0,
    and then use Newton-Raphson. When we do this,
    it is important to fix a normalization for the eigenvector,
    though the choice is arbitrary. We will choose
     that the linearized fields sum to one on the horizon.
    *)
log_{ij} = (\star \text{ Other notes: There are ways to optimize these types of programs via compiled})
       functions which can include arbitrary precision computations. However,
    these tend to slightly increase overhead for the problem presented
     here so instead I just construct the numerical versions of the
     differential equations using replacement rules which resulted
     in a more efficient and readable code for this problem. *)
```

1.) Deriving the equations of motion

2.) Background field configurations via Newton-Raphson

3.) Quasinormal modes or hydrodynamic spectrum

<code>In[*]:= (* Make sure variables don't have any values stored in memory*)</code>

```
In[*]:= Clear[kk, zmin, zmax]
 Numerical Parameters
In[*]:= (* If we start here,
     useful to redefine parameters for the numerics. Start with precision *)
ln[\bullet]:= \{zmin, zmax\} = \{0, 1\};
ln[-]:= mp = 60;
    $MinPrecision = mp;
ln[\bullet]:= (* Set threshold for "0" < 10^{-mp} *)
ln(-):= chopmin = 10^{-100};
```

```
nie]:= (* Define number of grid points and min and max values. Since we
       will need a background solution, here we import the solution. *)
I_{n[\cdot]} = \{Q, \lambda data, D0\lambda, D1\lambda, D2\lambda, \mu tab, \mu stab, \phi vectab, Axvectab, \eta vectab\} = I_{n[\cdot]} = \{Q, \lambda data, D0\lambda, D1\lambda, D2\lambda, \mu tab, \mu stab, \phi vectab, Axvectab, \eta vectab\}
         Import[NotebookDirectory[] <> "Data/probe_background_4d_sols.mx"];
In[*]:= NumGridPoints = Length[λdata] - 1;
In[\bullet]:= N\mu = Length[\mu tab] - 1;
     N\mu s = Length[\mu stab] - 1;
      \{\mu \min, \mu \max\} = \{\mu tab[[1]], \mu tab[[N\mu + 1]]\};
      \{\mu \text{smin}, \mu \text{smax}\} = \{\mu \text{stab}[1], \mu \text{stab}[N\mu \text{s} + 1]\};
_{ln[*]:=} (* We need to discretize the wavevectors. Can do a Chebyshev grid again. It is
         good to solve for at least 2 different k values so that you can check \omega ~ vk +
       O(k^2) for the second sound mode. More k vectors is even better,
      but time consuming. Below, we will do this using a faster method. *)
     Numk = 1;
     {kmin, kmax} = \left\{\frac{1}{100}, 1/10\right\};
     ktab = N \left[ Table \left[ \frac{(kmax + kmin)}{2} - \frac{(kmax - kmin)}{2} * Cos[\pi * j / Numk], \{j, 0, Numk\} \right], mp \right];
     eigs = Table[0, {i, 1, Numk + 1}];
```

Initialize EQs

Linearize EOs

Find eigenvalues

```
\mathit{In[*]}:= (* Can choose a particular solution, or could loop over solutions,
    though would need to change the structure of the eigs array. *)
     {\muStart, \muFinish} = {8, 8};
     {\musStart, \musFinish} = {7, 7};
    \muscount = \musStart;
    \mucount = \muStart;
    kcount = 1;
In[*]:= nb = CreateDocument["", WindowSize → {Scaled[1/5], Scaled[1/2]}];
    While [\mu scount \leq \mu sFinish,
       \mucount = \muStart;
       While [\mu count \leq \mu Finish,
        kcount = 1;
```

```
While[kcount ≤ Numk + 1,
 NotebookWrite[nb,
  Cell[BoxData@RowBox[{"kcount =", kcount, "timing = ", Timing[
           \mu = \mu tab [\mu count];
           \mus = \mustab[[\muscount]];
           \eta \text{vec} = \eta \text{vectab}[\mu \text{scount}, \mu \text{count}];
           \phivec = \phivectab[[\muscount, \mucount]];
           Axvec = Axvectab[[μscount, μcount]];
            kk = ktab[kcount];
           Do [coeffargs[i] = \{z \rightarrow \lambda data[i]\},
                \phi''[z] \rightarrow D2\lambda[i].\phi vec, \phi'[z] \rightarrow D1\lambda[i].\phi vec, \phi[z] \rightarrow \phi vec[i],
                Ax''[z] \rightarrow D2\lambda[i].Axvec, Ax'[z] \rightarrow D1\lambda[i].Axvec, Ax[z] \rightarrow Axvec[i],
                \eta''[z] \rightarrow D2\lambda[[i]].\eta \text{vec}, \eta'[z] \rightarrow D1\lambda[[i]].\eta \text{vec}, \eta[z] \rightarrow \eta \text{vec}[[i]]
               }, {i, 1, NumGridPoints + 1}];
           M = ArrayFlatten[Table[Sum[Flatten[Table[Piecewise[{{linBCcoeffs[m,
                              1, n, dd] /. (coeffargs[1] /. \{z \rightarrow zmin\}), i = 1\},
                         {linBCcoeffs[m, 2, n, dd] /. (coeffargs[NumGridPoints + 1] /.
                               \{z \rightarrow zmax\}), i == NumGridPoints + 1}, {linEQcoeffs[m, n,
                              dd] /. coeffargs[i], i # 1 && i # NumGridPoints + 1}}],
                      {i, 1, NumGridPoints + 1}]] x Drmatrices[dd], {dd, 1,
                   Length[Drmatrices]}], {m, 1, Length[EQs]}, {n, 1, Length[EQs]}]];
           MQNMO = M /. \{\omega \rightarrow 0\};
           Mdim = Length[MQNM0[1]];
           MQNM1 = D[M, \omega] /. {\omega \rightarrow \Theta};
           MQNM2 = (1/2) * D[M, \{\omega, 2\}] /. \{\omega \rightarrow 0\};
           Amat = ArrayFlatten[{{MQNM2, 0 * IdentityMatrix[Mdim]}},
                {0 * IdentityMatrix[Mdim], IdentityMatrix[Mdim]}}];
            Bmat = ArrayFlatten[
               {{MQNM1, MQNM0}, {-IdentityMatrix[Mdim], 0 * IdentityMatrix[Mdim]}}];
            Clear[M, MQNM0, MQNM1, MQNM2, Mdim];
            eigs[kcount] = Chop[Eigensystem[{Bmat, -Amat}], chopmin];
          ][1]]}], "Output"]];
```

```
kcount++;
       ];
       \mucount++;
      ];
      \muscount++;
     ];
    NotebookClose[nb];
In[**]:= Export[NotebookDirectory[] <> "Data/probe_eigenvalues_and_eigenvectors_4d.mx",
       {μsStart, μsFinish, μStart, μFinish, ktab, eigs}];
```

4.) Optional: Increasing accuracy

```
n_{[lpha]:=} (* If we like, we can increase the accuracy of a qnm solution. The simplest way
      to do this is to promote \omega to a function of z and use Newton-Raphson,
    since this is much more efficient than diagonalizing a matrix. We
     will increase the size of the grid using our previous
     solutions as seeds. First we will do the background*)
```

Background field configurations via Newton-Raphson

Choose an index for μ and μ s

```
m_{\ell}:= (* Since we only looked at one choice of \mu and \mus for \omega above,
      we will just use those*)
ln[\bullet]:= \{\mu \text{sind}, \mu \text{ind}\} = \{\mu \text{sStart}, \mu \text{Start}\};
```

Define Chebyshev grid and derivative matrices on the larger grid

```
<code>ln[∗]:= (* Increase grid size. It is good to do this little</code>
     by little since the eigenfunctions are highly oscillatory,
    interpolation will introduce substantial error *)
In[*]:= NumGridPoints = 60;
    \{zmin, zmax\} = \{0, 1\};
```

```
In[*]:= (* grid points λ *)
      \lambda data = N \left[ Table \left[ \frac{(zmax + zmin)}{2} - \frac{(zmax - zmin)}{2} * Cos[\pi * j / NumGridPoints] \right],
             {j, 0, NumGridPoints}], mp];(*Chebyshev Table*)
      a\lambda = N[Table[Product[If[j == k, 1, (\lambda data[j] - \lambda data[k])], \{k, 1, NumGridPoints + 1\}],
             {j, NumGridPoints + 1}], mp];
      D1\lambda = Table \Big[ If \Big[ i = j, Sum \Big[ If \Big[ k = j, 0, \frac{1}{\lambda data \llbracket j \rrbracket - \lambda data \llbracket k \rrbracket} \Big], \{k, 1, NumGridPoints + 1\} \Big], \{k, 1, NumGridPoints + 1\} \Big], \{k, 1, NumGridPoints + 1\} \Big]
             aλ[i]
aλ[j] (λdata[i] - λdata[j])
],
            {i, 1, NumGridPoints + 1}, {j, 1, NumGridPoints + 1} ;
      Clear[aλ];
      D2\lambda = D1\lambda.D1\lambda;
      D0λ = IdentityMatrix[NumGridPoints + 1];
      Drmatrices = \{D2\lambda, D1\lambda, D0\lambda\};
       (* test to check derivatives *)
      Max[Table[3 \lambda data[i]]^2, \{i, 1, NumGridPoints + 1\}] -
            D1λ.Table[λdata[i]]<sup>3</sup>, {i, 1, NumGridPoints + 1}] // Chop]
      Max[Table[6 λdata[i]], {i, 1, NumGridPoints + 1}] -
            D2λ.Table[λdata[i]]<sup>3</sup>, {i, 1, NumGridPoints + 1}] // Chop]
Out[ • ]= 0
Out[ • ]= 0
       Numerical Parameters
ln[.] = mp = 60;
       $MinPrecision = mp;
```

```
ln[-]:= chopmin = 10^{-100};
<code>/n[•]:= (* Import the lower accuracy backgrounds*)</code>
In[•]:= {Q, λdataLowPrec, D0λLowPrec, D1λLowPrec, D2λLowPrec,
                                             \mutab, \mustab, \phivectabLowPrec, AxvectabLowPrec, \etavectabLowPrec} =
                                        Import[NotebookDirectory[] <> "Data/probe_background_4d_sols.mx"];
<code>ln[∗]:= (* Import the lower accuracy eigenvectors and eigenvalues*)</code>
m_{\ell} = \{\mu \text{SStart}, \mu \text{SFinish}, \mu \text{Start}, \mu \text{Finish}, \text{ktabLowPrec}, \text{eigsLowPrec}\} = \{\mu \text{SSTart}, \mu \text{SFinish}, \mu \text{STart}, \mu
                                       Import[NotebookDirectory[] <> "Data/probe_eigenvalues_and_eigenvectors_4d.mx"];
l_{n/\pi}|_{x=1} (* Make sure there is an eigenvector and eigenvalue at the desired index,
                          else choose the first value for which there is one *)
```

```
m[\bullet]:=\mu \text{sind} = \text{If}[\mu \text{sStart} \leq \mu \text{sind} \leq \mu \text{sFinish}, \mu \text{sind}, \mu \text{sStart}];
                   \muind = If[\muStart \leq \muind \leq \muFinish, \muind, \muStart];
                   Initialize EQs
  ln[*]:= fieldredefa = \{\eta[z] \rightarrow z^2 * \eta[z], \phi[z] \rightarrow \phi[z] (1-z), f[z] \rightarrow 1-z^3\};
                    fieldredef = Flatten[{fieldredefa, D[fieldredefa, z], D[fieldredefa, {z, 2}]}];
                    Clear[fieldredefa]
  In[*]:= backgroundEQs =
                             Import[NotebookDirectory[] <> "EQs/probe_4d_backgroundEQs.mx"] /. fieldredef //
                                       Expand // Simplify;
                    Clear[fieldredef]
  In[•]:= Series[backgroundEQs, {z, 0, -1}] // Simplify
                    Series[backgroundEQs /. \{z \rightarrow 1-z\}, \{z, 0, -1\}] // Simplify
Out[\circ]= \{0[z]^3, 0[z]^4, 0[z]^4\}
Out[\circ]= \{0[z]^1, 0[z]^1, 0[z]^0\}
 In[*]:= EQs =
                              \left(\left\{\frac{\mathsf{backgroundEQs}[\![1]\!]}{\mathsf{z}^3\ (1-\mathsf{z})}\,,\,\,\frac{\mathsf{backgroundEQs}[\![2]\!]}{\mathsf{z}^4\ (1-\mathsf{z})}\,,\,\,\frac{\mathsf{backgroundEQs}[\![3]\!]}{\mathsf{z}^4}\right\}\right) \ //\ \mathsf{Expand}\ //\ \mathsf{Simplify}\ //\ \mathsf{Expand}\ //\ \mathsf{Expand
                                      PowerExpand // Simplify;
  ln[\cdot]:= \{bc1, bc2, bc3\} = EQs /. \{z \rightarrow 1\} // Simplify;
  log_{in[\pi]} = BCs = \{ \{ \eta'[zmin], bc1 \}, \{ \phi[zmin] - \mu, bc2 \}, \{ Ax[zmin] - \mu s, bc3 \} \} // Simplify;
  In[*]:= Clear[bc1, bc2, bc3, backgroundEQs]
                   Linearize EQs
```

```
ln[\sigma]:= fieldRep = \{\eta \to \eta + \varepsilon s * \delta \eta, \phi \to \phi + \varepsilon s * \delta \phi, Ax \to Ax + \varepsilon s * \delta Ax\};
     linBackEQ = \partial_{\epsilon s} (EQs /. fieldRep) /. \epsilon s \rightarrow 0;
     linBackBC = \partial_{\epsilon s} (BCs /. fieldRep) /. \epsilon s \rightarrow 0;
     Clear[fieldRep];
     linFieldList = \{\delta\eta[z], \delta\phi[z], \delta Ax[z]\};
     (*dlinFieldList[choose a function][choose a derivative]*)
     dlinFieldList =
        Table [\{\partial_{z,z} \text{linFieldList}[fun], \partial_{z} \text{linFieldList}[fun], \text{linFieldList}[fun]\},
         {fun, 1, Length[linFieldList]}];
     Clear[linFieldList]
     (* Note: must change argument for boundary conditions from z to zmin or zmax *)
     boundrules = \{z \rightarrow zmin, z \rightarrow zmax\};
     Do[linEQcoeffs[eq, fun, der] = D[linBackEQ[eq], dlinFieldList[fun][der]]],
        {eq, 1, Length[EQs]}, {fun, 1, Length[EQs]}, {der, 1, Length[dlinFieldList[eq]]}];
     (*BCc[BC_i][min or max][choose a function][derivative]*)
     Do[linBCcoeffs[eq, bound, fun, der] =
          \partial_{(dlinFieldList[[fun][[der]]/.boundrules[[bound]])} linBackBC[[eq]][[bound]], \\ \{eq, 1, Length[EQs]\}, \\
        {bound, 1, 2}, {fun, 1, Length[EQs]}, {der, 1, Length[dlinFieldList[eq]]}];
In[*]:= Clear[linBackBC, linBackEQ, dlinFieldList, boundrules]
     Seeds
     Newton-Raphson loop
n[[*]:= (* Define thresholds which will halt the evaluation if exceeded*)
ln[-]:= Error = 10^{-(mp/2)};
     ErrorMax = 1000000;
In[*]:= nb = CreateDocument["", WindowSize → {Scaled[1/5], Scaled[1/2]}];
     \mu = \mu tab[\mu ind];
     \mu s = \mu stab[\mu sind];
     \epsilon p = (ErrorMax - Error) / 2;
     count = 0;
     NotebookWrite[nb,
        Cell[BoxData@RowBox[{"\mu_s # ", \muscount, ".) \mu # ", \mucount, ".) ", Timing[
                While[ep > Error && ep < ErrorMax,
```

```
(* rules for replacing functions in the equations *)
Do [coeffargs[i] = \{z \rightarrow \lambda data[i]\},
    \phi''[z] \rightarrow D2\lambda[[i]].\phi \text{vec}, \phi'[z] \rightarrow D1\lambda[[i]].\phi \text{vec}, \phi[z] \rightarrow \phi \text{vec}[[i]],
    Ax''[z] \rightarrow D2\lambda[i]. Axvec, Ax'[z] \rightarrow D1\lambda[i]. Axvec, Ax[z] \rightarrow Axvec[i],
    \eta''[z] \rightarrow D2\lambda[[i]].\eta \text{vec}, \eta'[z] \rightarrow D1\lambda[[i]].\eta \text{vec}, \eta[z] \rightarrow \eta \text{vec}[[i]]
  }, {i, 1, NumGridPoints + 1}];
(* Construct the derivative matrix *)
M = ArrayFlatten[Table[Sum[Flatten[Table[Piecewise[{{linBCcoeffs[m,
                1, n, dd] /. (coeffargs[1] /. \{z \rightarrow zmin\}), i = 1\},
            {linBCcoeffs[m, 2, n, dd] /. (coeffargs[NumGridPoints + 1] /.
                  \{z \rightarrow zmax\}), i = NumGridPoints + 1}, \{linEQcoeffs[m, n, dd] / .
               coeffargs[i], i # 1 && i # NumGridPoints + 1}}],
          {i, 1, NumGridPoints + 1}]] x Drmatrices[dd], {dd, 1, Length[
        Drmatrices]}], {m, 1, Length[EQs]}, {n, 1, Length[EQs]}]];
(* Construct the value of the
 eoms given the current configuration of the fields. *)
Etot = Chop[Flatten[Parallelize[Table[Piecewise[
        \{\{BCs[m, 1]\} /. (coeffargs[1] /. \{z \rightarrow zmin\}), i == 1\}, \{BCs[m, 2]\} /.
            (coeffargs[NumGridPoints + 1] /. \{z \rightarrow zmax\}), i = NumGridPoints +
              1}, \{EQs[m] / . coeffargs[i], i \neq 1 \& \& i \neq NumGridPoints + 1\}\}],
       {m, 1, Length[EQs]}, {i, 1, NumGridPoints + 1}]]], chopmin];
(* Solve for the change in the fields *)
deltaFields = Chop[LinearSolve[M, -Etot], chopmin];
(* Compute the norm of deltaFields to keep track of convergence *)
ep = Norm[deltaFields, ∞];
If[NumberQ[ep], NotebookWrite[nb,
  Cell[BoxData@RowBox[{"ep= ", ep}], "Output"]], Break[]];
(* If the norm is too large, introduce friction. *)
friction = If [\epsilon p < 1, 1, 1/10];
combinedFields = combinedFields + friction * deltaFields;
(* partition the solution
 for the combined fields into individual fields *)
\{\eta \text{vec}, \phi \text{vec}, Ax \text{vec}\} = Partition[combinedFields, NumGridPoints + 1];
Clear[friction, deltaFields];
count++;
```

```
];
              ][1], ", # Iteration = ", count}], "Output"]];
     NotebookClose[nb];
     (* Example plot of a background field *)
In[n] := ListLinePlot[Table[{\lambda data[i], \eta vec[i]}}, {i, 1, NumGridPoints + 1}]]
     6
     5
Out[ • ]= 3 |
     2
                0.2
In[*]:= (* Export solutions,
     include Chebyshev grid parameters so we don't duplicate efforts *)
In[**]:= Export[NotebookDirectory[] <> "Data/probe_background_4d_sols_increasedAccuracy.mx",
        {\musind, \muind, Q, \lambdadata, D0\lambda, D1\lambda, D2\lambda, \phivec, Axvec, \etavec}];
  Quasinormal mode via Newton-Raphson. Want to track superfluid second
  sound.
n_{[*]:=} (* We will now scan a larger set of wavevectors zooming in on a single mode. \star)
     Numerical Parameters
In[*]:= (* If we start here,
     useful to redefine parameters for the numerics. Start with precision *)
ln[.] = mp = 60;
     $MinPrecision = mp;
ln[\cdot]:= (* Set threshold for "0" < 10^{-mp} *)
In[*]:= chopmin = 10^{-100};
ln[*]:= (* Import higher accuracy background *)
In[e]:= \{\mu \text{sind}, \mu \text{ind}, Q, \lambda \text{data}, D0\lambda, D1\lambda, D2\lambda, \phi \text{vec}, Axvec}, \eta \text{vec}\} = Import[e]
         NotebookDirectory[] <> "Data/probe_background_4d_sols_increasedAccuracy.mx"];
```

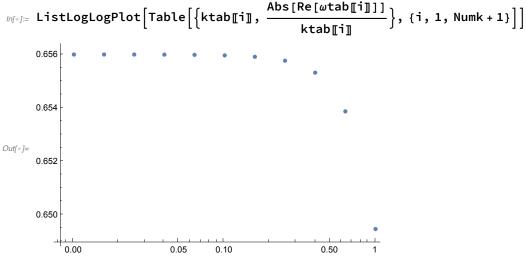
```
In[*]:= kmin = ktabLowPrec[[1]];
     kmax = 10^2 kmin;
     Numk = 10;
     ktab = N[Table[kmin * 10^2 \frac{j}{Numk}, {j, 0, Numk}], mp];
ln[\bullet] = \omega tab = Table[0, \{i, 1, Numk + 1\}];
     Initialize EOs
     Linearize EQs
     Seeds for eigenvalue problem
ln[*]:= (* The velocity of the superfluid sound mode is bounded above
      and below by 1 so let's try and find this eigenvalue. Furthermore,
     if we are in a stable regime, Im[\omega] <
      0 but \text{Im}[\omega/k^2] is not too negative for the hydrodynamic modes. In Eigensystem,
     eigenvalues are listed first and eigenvectors are listed second. For us,
     eigs[wavevectorindex, 1 for eigenvalue or 2 for eigenvector, which eigenvalue].
       We expect a pair of modes. *)
In[a]:= hydroindex = Table Position
         Table \left[ \frac{Abs[Re[eigsLowPrec[[j, 1, i]]]]}{ktabLowPrec[[j]]} < 1 \&\& -10 < \frac{Im[eigsLowPrec[[j, 1, i]]]}{ktabLowPrec[[j]]^2} < 0, \right]
           {i, 1, Length[eigsLowPrec[j, 1]]}, True, {j, 1, Length[ktabLowPrec]}];
In[•]:= hydroindex
Out[\circ] = \{ \{ \{324\}, \{325\}, \{327\} \}, \{ \{324\}, \{325\}, \{327\} \} \} \}
In[*]:= (* For the coarse grid with N=40,
     we see a mode at index 324 or 325 which satisfies the hydrodynamic
      criteria for our values of k. We use one of these as our seed. *)
log_{ij} = \omega seed = Table[eigsLowPrec[1, 1, hydroindex[1, 1, 1]]], {i, 1, NumGridPoints + 1}];
     Export[NotebookDirectory[] <> "Data/probe_omegaseed_4d.mx", {ωseed[1], ktab[1]}}];
ln[*]:= (* Now, recall the eigenvectors have length 2*(number of fields)*
        (number of low precision grid points). The 2 is from \omega\delta X and
       δX. We want just the fields over the grids so must partition. The
       number of fluctuating fields is the number of equations -
      1 because we have included \omega'[z] as an equation. *)
Info = != eigenfields = Partition[Partition[eigsLowPrec[1, 2, hydroindex[1, 1, 1]]],
           (Length[EQs] - 1) * (Length[λdataLowPrec])][2]], Length[λdataLowPrec]];
     eigenfields2 = Partition[Partition[eigsLowPrec[1, 2, hydroindex[1, 1, 1]]],
           (Length[EQs] - 1) * (Length[λdataLowPrec])][1], Length[λdataLowPrec]];
```

```
ln[\cdot]:= (* Check that eigenfields2 = \omega*eigenfields1 *)
log_{\sigma} = Table[Max[Chop[eigenfields2[j]] - \omega seed[1]] * eigenfields[j]]], {j, 1, 4}]
Out[\circ] = \{0, 0, 0, 0\}
<code>ln[•]:= (* The order of the fields follows from the above</code>
     linearization step with linFieldList. It is at, ax, \sigma r, then \sigma i. *)
In[*]:= (* First interpolate *)
    atint = Interpolation[
        Table[{λdataLowPrec[i], eigenfields[1, i]}, {i, 1, Length[λdataLowPrec]}]];
    axint = Interpolation[
        Table[{λdataLowPrec[i], eigenfields[2, i]}, {i, 1, Length[λdataLowPrec]}]];
    σrint = Interpolation[
        Table[{λdataLowPrec[i], eigenfields[3, i]}, {i, 1, Length[λdataLowPrec]}]];
    σiint = Interpolation[
        Table[{λdataLowPrec[i], eigenfields[4, i]}, {i, 1, Length[λdataLowPrec]}]];
In[*]:= (* Now make the seeds. Importantly,
    we have to rescale to impose our boundary condition at the horizon. *)
In[*]:= horcon = Sum[eigenfields[j, Length[λdataLowPrec]]], {j, 1, 4}];
In[•]:= horcon
0.00379969064776502399501435581479130617880182712406576007484155 i
In[*]:= atseed = Table[atint[λdata[i]]], {i, 1, NumGridPoints + 1}];
    axseed = Table[axint[λdata[i]], {i, 1, NumGridPoints + 1}];
    orseed = Table[orint[\lambdadata[i]], {i, 1, NumGridPoints + 1}];
    σiseed = Table[σiint[λdata[i]], {i, 1, NumGridPoints + 1}];
```

```
In[*]:= {Show[ListLinePlot[
                                   Table[\{\lambda data[i], Re[atseed[i]]\}, \{i, 1, NumGridPoints + 1\}]], ListPlot[
                                    Table[{λdataLowPrec[i], Re[eigenfields[1, i]]}, {i, 1, Length[λdataLowPrec]}]]],
                          Show[ListLinePlot[
                                   Table[\{\lambda data[i]\}, Re[axseed[i]]\}, \{i, 1, NumGridPoints + 1\}]], ListPlot[
                                   Table \cite{AdataLowPrec[i]}, Re[eigenfields[2, i]]\}, \{i, 1, Length[\lambda dataLowPrec]\}]]], and the proof of th
                          Show[ListLinePlot[
                                    Table[{λdata[i], Re[σrseed[i]]}, {i, 1, NumGridPoints + 1}]], ListPlot[
                                    Table \cite{AdataLowPrec[i]}, Re \cite{Beigenfields[3, i]]}, \cite{AdataLowPrec[i]}, Length \cite{AdataLowPrec[i]}, Table \cite{Beigenfields[3, i]]}, Table \cite{Beigenfields
                          Show[ListLinePlot[
                                   Table[{λdata[i], Re[σiseed[i]]]}, {i, 1, NumGridPoints + 1}]], ListPlot[
                                   Table[\{\lambda dataLowPrec[[i]], Re[eigenfields[[4, i]]]\}, \{i, 1, Length[\lambda dataLowPrec]\}]]]\}
                            0.4
                                                                                                                                                                                       0.4
                                                                                                                                                                                                        0.6
                                                                                                                                                                                                                          0.8
                                                                                                                                                                                                                                             1.0
                            0.2
                                                                                                                                          -0.2
                                                                                                                                           -0.4
                                                                                       0.6
                                                                                                         8.0
                                                                                                                           1.0,
Out[ • ]=
                       -0.2
                                                                                                                                           -0.6
                         -0.4
                                                                                                                                          -0.8
                         -0.6
                                                                       0.4
                                                                                                                                                                       0.2
                                                                                                                                                                                        0.4
                                                                                                                                                                                                                           0.8
                                                                                                                                                                                                          0.6
                                                                                                                                                                                                                                             1.0
                          -0.05
                                                                                                                                           -0.05
                         -0.10
                          -0.15
                                                                                                                                           -0.10
                          -0.20
                          -0.25
                                                                                                                                          -0.15
                          -0.30 F
  ln[∗]:= {atvec, axvec, σrvec, σivec, ωvec} = {atseed, axseed, σrseed, σiseed, ωseed};
  ln[∗]:= combinedFields = Flatten[{atvec, axvec, σrvec, σivec, ωvec}];
                    Find eigenvalues
  log_{*}:= nb = CreateDocument["", WindowSize \rightarrow {Scaled[1/5], Scaled[1/2]}];
                     kcount = 1;
                    While[kcount ≤ Numk + 1,
                              \epsilon p = (ErrorMax - Error) / 2;
                              count = 0;
                              NotebookWrite[nb,
                                   Cell[BoxData@RowBox[{"kcount = ", kcount, " timing = ", Timing[
                                                                 While [ep > Error && ep < ErrorMax,
```

```
\mu = \mu \mathsf{tab}[\mu \mathsf{ind}];
\mu s = \mu stab[\mu sind];
kk = ktab[kcount];
Do[coeffargs[i] = \{z \rightarrow \lambda data[i]\},
     \phi''[z] \rightarrow D2\lambda[i].\phi vec, \phi'[z] \rightarrow D1\lambda[i].\phi vec, \phi[z] \rightarrow \phi vec[i],
     Ax''[z] \rightarrow D2\lambda[[i]].Axvec, Ax'[z] \rightarrow D1\lambda[[i]].Axvec, Ax[z] \rightarrow Axvec[[i]],
     \eta''[z] \rightarrow D2\lambda[i].\eta \text{vec}, \eta'[z] \rightarrow D1\lambda[i].\eta \text{vec}, \eta[z] \rightarrow \eta \text{vec}[i],
     at''[z] \rightarrow D2\lambda[i]].atvec, at'[z] \rightarrow D1\lambda[i]].atvec, at[z] \rightarrow atvec[i],
     ax''[z] \rightarrow D2\lambda[i].axvec, ax'[z] \rightarrow D1\lambda[i].axvec, ax[z] \rightarrow axvec[i],
     \sigma r''[z] \rightarrow D2\lambda[i].\sigma rvec, \sigma r'[z] \rightarrow D1\lambda[i].\sigma rvec, \sigma r[z] \rightarrow \sigma rvec[i],
     \sigma \texttt{i''}[\texttt{z}] \rightarrow \texttt{D2}\lambda[\![\texttt{i}]\!].\sigma \texttt{ivec}, \ \sigma \texttt{i'}[\texttt{z}] \rightarrow \texttt{D1}\lambda[\![\texttt{i}]\!].\sigma \texttt{ivec}, \ \sigma \texttt{i}[\texttt{z}] \rightarrow \sigma \texttt{ivec}[\![\texttt{i}]\!],
     \omega''[z] \to \mathsf{D}2\lambda[\![\mathsf{i}]\!].\omega\mathsf{vec},\,\omega'[z] \to \mathsf{D}1\lambda[\![\mathsf{i}]\!].\omega\mathsf{vec},\,\omega[z] \to \omega\mathsf{vec}[\![\mathsf{i}]\!]
   }, {i, 1, NumGridPoints + 1}];
(* Construct the derivative matrix *)
M = ArrayFlatten[Table[Sum[Flatten[Table[Piecewise[{{linBCcoeffs[m,
                     1, n, dd] /. (coeffargs[1] /. \{z \rightarrow zmin\}), i = 1\},
               {linBCcoeffs[m, 2, n, dd] /. (coeffargs[NumGridPoints + 1] /.
                      \{z \rightarrow zmax\}), i = NumGridPoints + 1}, {linEQcoeffs[m, n,
                    dd] /. coeffargs[i], i # 1 && i # NumGridPoints + 1}}],
            {i, 1, NumGridPoints + 1}]] x Drmatrices[[dd]], {dd, 1,
        Length[Drmatrices]}], {m, 1, Length[EQs]}, {n, 1, Length[EQs]}]];
(* Construct the value of the eoms
  given the current configuration of the fields. *)
Etot = Chop[Flatten[Parallelize[Table[Piecewise[
          \{\{BCs[m, 1] /. (coeffargs[1] /. \{z \rightarrow zmin\}), i = 1\}, \{BCs[m, 2] /. \}
               (coeffargs[NumGridPoints + 1] /. {z → zmax}), i == NumGridPoints +
                 1}, {EQs[m] /. coeffargs[i], i ≠ 1 && i ≠ NumGridPoints + 1}}],
         {m, 1, Length[EQs]}, {i, 1, NumGridPoints + 1}]]], chopmin];
(* Solve for the change in the fields *)
deltaFields = Chop[LinearSolve[M, -Etot], chopmin];
(* Compute the norm of deltaFields to keep track of convergence *)
ep = Norm[deltaFields, ∞];
If [NumberQ[€p], NotebookWrite[nb,
   Cell[BoxData@RowBox[{"ep= ", ep}], "Output"]], Break[]];
```

```
(* If the norm is too large, introduce friction. *)
                   friction = If [\epsilon p < 1, 1, 1/10];
                   combinedFields = combinedFields + friction * deltaFields;
                   (* partition the solution
                    for the combined fields into individual fields *)
                   {atvec, axvec, \sigmarvec, \sigmaivec, \omegavec} =
                    Partition[combinedFields, NumGridPoints + 1];
                   Clear[friction, deltaFields];
                   count++;
                 ];
               [1], ", # Iteration = ", count}], "Output"]];
       \omegatab[kcount] = \omegavec[1];
       kcount++;
      ];
     NotebookClose[nb];
In[*]:= ListLogLogPlot[Table[{ktab[i], Abs[Re[ωtab[i]]]}, {i, 1, Numk + 1}]]
     0.500
     0.100
Out[•]= 0.050
     0.010
         0.00
                          0.05
                                 0.10
ln[*]:= (* Show that this is linear, i.e. a sound mode \omega(k) \sim v*k*)
```



In[*]:= Export[

NotebookDirectory[] <> "Data/probe_fluctuation_4d_sols_increasedAccuracy.mx", { μ sind, μ ind, ktab, ω tab}];

5.) Matching to hydrodynamic theory

```
In[*]:= (* In appendix B of https://arxiv.org/pdf/2312.08243,
      I write the expression for the velocity of second sound in the probe limit
       (eq. B.3). Here we will just match the result to our numerics rather than
       deriving this expression. Since we only increased precision for one point,
     we will use the lower precision data for the thermodynamic derivatives. It
       is easy to change this by including more points when we increase precision.*)
In[\bullet]:= Clear[\mu s, \mu]
In [*]:= (* Define \chi \xi \xi = \mu/\partial_{\varepsilon}(\xi \rho_{s}) *)
\lim \left\{ \sum_{n \in \mathbb{N}} \chi \xi \xi \left[ \mu s_n, \mu_n, \rho s_n, \chi n s h_n \right] \right\} := \frac{\mu}{\mu s * \chi n s h + \rho s};
<code>ln[•]:= (* Define the velocity of second sound *)</code>
\log \left[\chi - \chi - \chi - \chi \right] := \frac{\chi - \chi}{\chi - \chi} - \sqrt{\frac{\chi - \chi}{\chi - \chi}} + \frac{1}{\chi - \chi}
<code>/n[•]:= (* Construct thermodynamic derivatives *)</code>
In[*]:= {Q, λdataLowPrec, D0λLowPrec, D1λLowPrec, D2λLowPrec,
          \mutab, \mustab, \phivectabLowPrec, AxvectabLowPrec, \etavectabLowPrec} =
         Import[NotebookDirectory[] <> "Data/probe background 4d sols.mx"];
In[\bullet]:= N\mu = Length[\mu tab] - 1;
     N\mu s = Length[\mu stab] - 1;
```

0.170

0.175

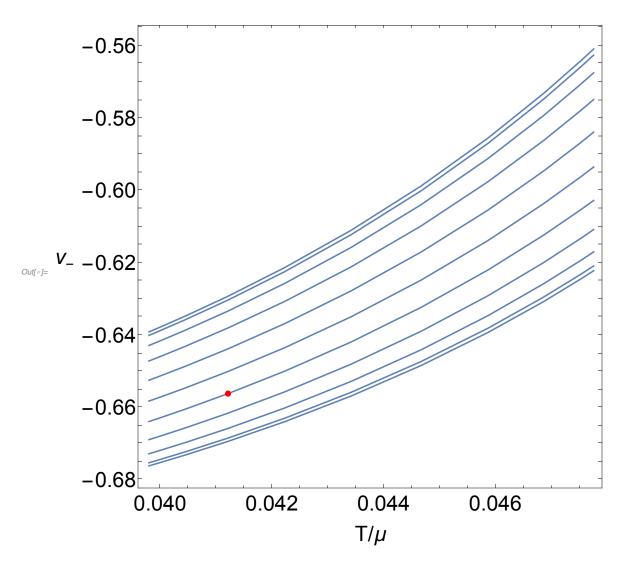
0.180

0.185

0.190

0.195

```
In[•]:= Show
             \mathsf{Table}\Big[\mathsf{ListLinePlot}\Big[\mathsf{Table}\Big[\Big\{\frac{1}{\mu\mathsf{tab}\llbracket\mathtt{j}\rrbracket}\,,\,\frac{\rho\mathsf{tab}\llbracket\mathtt{i}\,,\,\mathtt{j}\rrbracket}{\mu\mathsf{tab}\llbracket\mathtt{j}\rrbracket^2}\Big\},\,\{\mathtt{j}\,,\,\mathtt{1}\,,\,\mathtt{N}\mu\,+\,\mathtt{1}\}\Big]\Big]\,,\,\{\mathtt{i}\,,\,\mathtt{1}\,,\,\mathtt{N}\mu\,\mathrm{s}\,+\,\mathtt{1}\}\Big]\Big]
           0.278
           0.276
           0.274
Out[ • ]=
           0.272
           0.270
                            0.170
                                           0.175
                                                                                                       0.195
                                                          0.180
                                                                         0.185
                                                                                        0.190
 In[*]:= Show
             \mathsf{Table}\Big[\mathsf{ListLinePlot}\Big[\mathsf{Table}\Big[\Big\{\frac{1}{\mu\mathsf{tab}\llbracket\mathtt{j}\rrbracket}\,,\,\frac{\rho\mathsf{stab}\llbracket\mathtt{i}\,,\,\mathtt{j}\rrbracket}{\mu\mathsf{tab}\rrbracket\mathtt{i}\rrbracket^2}\Big\},\,\{\mathtt{j},\,1,\,\mathsf{N}\mu+1\}\Big]\Big]\,,\,\{\mathtt{i},\,1,\,\mathsf{N}\mu\mathsf{s}+1\}\Big]\Big]
           0.24
           0.23
           0.22
Out[ • ]=
           0.21
           0.20
                          0.170
                                          0.175
                                                         0.180
                                                                        0.185
                                                                                        0.190
                                                                                                       0.195
                                                                                                                       0.200
 ln[\bullet]:= d\rho d\mathcal{E} = D1\mu s.\rho tab;
           d\rho d\mu = Table[D1\mu[j].\rho tab[i], \{i, 1, N\mu s + 1\}, \{j, 1, N\mu + 1\}];
           d\rho sdg = D1\mu s.\rho stab;
 log[\cdot]:= velocityTab = Table[velocityMinus[d\rhodg[i, j], d<math>\rhod\mu[i, j],
                     \chi\xi\xi[\mu\mathsf{stab}[\![i]\!],\mu\mathsf{tab}[\![j]\!],\rho\mathsf{stab}[\![i,j]\!],\mathsf{d}\rho\mathsf{sd}\xi[\![i,j]\!]]],\{\mathsf{i},1,\mathsf{N}\mu\mathsf{s}+1\},\{\mathsf{j},1,\mathsf{N}\mu+1\}];
 _{ln[*]:=} (* Plot velocities. The red dot is where we look at the velocity*)
 In[*]:= {ωcheck, kcheck} = Import[NotebookDirectory[] <> "Data/probe_omegaseed_4d.mx"];
```



In[*]:= velocityTab[[μsind, μind]]

Out[e] = -0.655998956186536502845012096669952750391312256977564128187941

In[•]:= Re [ωcheck / kcheck]

Out [*] = -0.655998634310152763995544284255483129790086710832858677329209]

```
In[*]:= Im[\omega check/kcheck^2]
\textit{Out}[\ ^{\circ}] = \ -0.0284500844664581646848706393651982598002923915234086331998859
ln[\cdot]:= (* matches to 6 significant figures. The
       discrepancy is both numerical and from finite k effects. *)
```