

# Comparison of One-Sided and Diffusion-Limited Evaporation Models for Thin Liquid Droplets (Supplementary Material)

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Here we describe the procedure for solving the axisymmetric Laplace's equation with the hybrid spectral-finite-difference method described in §3. For ease of use, we present this method for a more general problem than that considered in §3.

We wish to numerically solve for the concentration  $c_g$  governed by

$$0 = \nabla^2 c_g = \frac{\partial^2 c_g}{\partial z^{*2}} + \frac{1}{r} \frac{\partial}{\partial r} \left( r \frac{\partial c_g}{\partial r} \right) \quad (1)$$

on the the semi-infinite domain  $(r, z^*) \in (0, d) \times (0, \infty)$ . The concentration  $c_g$  is subject to the symmetric boundary condition  $c_g(r) = c_g(-r)$  as well as  $c_g(r = d) = 0$  and  $c_g \rightarrow 0$  as  $z^* \rightarrow \infty$ . At  $z^* = 0$ , it takes on the value  $c_g(r, z^* = 0) = f(r)$ . Since equation (1) must be solved at each time step, the goal of this numerical method is to reduce equation (1) to a matrix equation that can be quickly and efficiently solved. Note that we do not explicitly include time-dependence in this derivation because it can be encapsulated in the boundary data  $f$ .

We begin by writing  $c_g$  as an expansion of Laguerre functions  $L_n$ ,

$$c_g(r, z^*) \approx \sum_{n=0}^{M-1} a_n(r) L_n(z^*), \quad (2)$$

where the coefficients  $a_n(r)$  vary radially. For now, we assume that there is some set of coefficients  $a_n''(r)$  such that

$$\frac{\partial^2 c_g}{\partial z^{*2}} = \sum_{n=0}^{M-1} a_n(r) \frac{\partial^2 L_n}{\partial z^{*2}} = \sum_{n=0}^{M-1} a_n''(r) L_n(z^*). \quad (3)$$

Then, substituting equation (2) into equation (1) and applying orthogonality of the Laguerre functions gives, for each  $n$ ,

$$a_n''(r) + \frac{1}{r} \frac{\partial}{\partial r} \left( r \frac{\partial a_n}{\partial r} \right) = 0. \quad (4)$$

We first determine  $a_n''$ , and then show how we discretize in  $r$ .

The Laguerre functions are defined as

$$L_n(x) = e^{-x/2} \ell_n(x), \quad (5)$$

where  $\ell_n(x)$  is the  $n$ -th Laguerre polynomial which satisfies

$$\frac{\partial \ell_n}{\partial x} = - \sum_{k=0}^{n-1} \ell_k(x). \quad (6)$$

Thus, we have

$$\frac{dL_n}{dx} = \frac{d}{dx} (e^{-x/2} \ell_n(x)) \quad (7)$$

$$= -\frac{1}{2} e^{-x/2} \ell_n(x) - e^{-x/2} \sum_{k=0}^{n-1} \ell_k(x) \quad (8)$$

$$= -\frac{1}{2} L_n(x) - \sum_{k=0}^{n-1} L_k(x). \quad (9)$$

Defining the lower triangular matrix  $\mathbf{D}$  and vector of Laguerre functions  $\mathbf{L}$  as

$$(\mathbf{D})_{i,j} = \begin{cases} 0 & i \geq j \\ 1 & i < j \end{cases}, \quad (\mathbf{L})_i = L_i(x), \quad (10)$$

we can use equation (9) to write

$$\frac{d\mathbf{L}}{dx} = \left( -\frac{1}{2} \mathbf{I} - \mathbf{D} \right) \mathbf{L}, \quad (11)$$

where  $\mathbf{I}$  is the identity matrix. Thus,

$$\frac{d^2 \mathbf{L}}{dx^2} = \left( -\frac{1}{2} \mathbf{I} - \mathbf{D} \right)^2 \mathbf{L} = \underbrace{\left( \frac{1}{4} \mathbf{I} + \mathbf{D} + \mathbf{D}^2 \right)}_{\mathbf{D}_z} \mathbf{L}. \quad (12)$$

With a vector of coefficients  $\mathbf{a}(r)$ , we have  $c_g = \mathbf{a}^T \mathbf{L}$  and thus

$$\frac{\partial^2 c_g}{\partial z^{*2}} = \mathbf{a}^T \mathbf{D}_z \mathbf{L}, \quad (13)$$

so the coefficients  $a_n''$  are given in vector form by  $\mathbf{a}'' = (\mathbf{a}^T \mathbf{D}_z)^T$ .

We now discretize each coefficient  $a_n(r)$  into its values at a discrete set of  $N$  nodes  $\{r_j\}$  placed in the center of cells. We then apply a positivity-preserving second-order centered finite-difference method that gives

$$\frac{1}{r} \frac{\partial}{\partial r} \left( r \frac{\partial a_n}{\partial r} \right) \Big|_{r_j} = \frac{2}{r_j \Delta r_j} [\alpha_j a_n(r_{j-1}) - (\alpha_j + \alpha_{j+1}) a_n(r_j) + \alpha_{j+1} a_n(r_{j+1})], \quad (14)$$

where

$$\alpha_j = \frac{r_{j-1/2}}{\Delta r_j + \Delta r_{j-1}}, \quad (15)$$

$r_{j-1/2}$  is the position of the boundary between cells  $j$  and  $j-1$ , and  $\Delta r_j$  is the width of cell  $j$ . Note that the boundary conditions in  $r$  define the values of ghost nodes needed to evaluate the above expressions near the boundaries of the domain. Defining the  $N \times M$  coefficient matrix  $(\mathbf{A})_{i,j} = a_j(r_i)$  and  $r$ -derivative matrix  $\mathbf{D}_r$  as

$$(\mathbf{D}_r)_{i,j} = \frac{2}{r_i \Delta r_i} \begin{cases} \alpha_i & j = i - 1 \\ -(\alpha_i + \alpha_{i+1}) & j = i \\ \alpha_{i+1} & j = i + 1 \\ 0 & \text{otherwise} \end{cases}, \quad (16)$$

the discrete form of the left-hand side of equation (4) is

$$\mathbf{A} \mathbf{D}_z + \mathbf{D}_r \mathbf{A}. \quad (17)$$

If a higher order-of-accuracy is desired, one can replace the second-order method here for one of higher order. Doing so simply changes the entries of  $\mathbf{D}_r$  and the rest of this method is unchanged.

Unfortunately the form of equation (17) is not computationally useful since the coefficient matrix  $\mathbf{A}$  is both left and right multiplied by other matrices. However, we can remedy this by stacking the matrix  $\mathbf{A}$  into a vector and expressing the right multiplication as a left multiplication. Define the mapping vectors  $\mathbf{I}$  and  $\mathbf{J}$  and coefficient vector  $\mathbf{u}$  (all with lengths  $NM$ ) such that

$$(\mathbf{u})_i = a_{J_i}(r_{I_i}) = (\mathbf{A})_{I_i, J_i}. \quad (18)$$

Note that  $J_i \in \{1, \dots, M\}$  and  $I_i \in \{1, \dots, N\}$ . Now, we can use the stacked  $z$ -derivative matrix

$$(\mathbf{D}_z^*)_{i,j} = \begin{cases} (\mathbf{D}_z)_{J_j, J_i} & I_i = I_j \\ 0 & \text{otherwise} \end{cases}. \quad (19)$$

To see this matrix gives the desired result (left multiplication instead of right), note that

$$(\mathbf{D}_z^* \mathbf{u})_i = \sum_{j=1}^{NM} (\mathbf{D}_z^*)_{i,j} (\mathbf{u})_j = \sum_{\substack{j \\ I_i = I_j}} (\mathbf{D}_z^*)_{i,j} a_{J_j}(r_{I_j}) = \sum_{\substack{j \\ I_i = I_j}} (\mathbf{D}_z)_{J_j, J_i} a_{J_j}(r_{I_i}). \quad (20)$$

Now, for a given index  $i$ , there are precisely  $M$  indices  $j$  such that  $I_i = I_j$  (i.e.,  $\mathbf{I}$  contains each value in  $\{1, \dots, N\}$  exactly  $M$  times). These indices must be distinct, and furthermore, the corresponding  $J_j$  cover  $\{1, \dots, M\}$ . Thus, the above sum can be replaced by

$$(\mathbf{D}_z^* \mathbf{u})_i = \sum_{j=1}^M (\mathbf{D}_z)_{j, J_i} a_j(r_{I_i}) = (\mathbf{A} \mathbf{D}_z)_{I_i, J_i}. \quad (21)$$

Thus, we have turned the right multiplication by a derivative matrix into a left multiplication.

We can similarly form the stacked  $r$ -derivative matrix

$$(\mathbf{D}_r^*)_{i,j} = \begin{cases} (\mathbf{D}_r)_{I_i, I_j} & J_i = J_j \\ 0 & \text{otherwise} \end{cases}. \quad (22)$$

By arguments similar to those above, one can show that

$$(\mathbf{D}_r^* \mathbf{u})_i = (\mathbf{D}_r \mathbf{A})_{I_i, J_i}. \quad (23)$$

Therefore, equation (4) can also be written

$$\mathbf{D}_z^* \mathbf{u} + \mathbf{D}_r^* \mathbf{u} = \underbrace{(\mathbf{D}_z^* + \mathbf{D}_r^*)}_{\mathbf{K}} \mathbf{u}. \quad (24)$$

The boundary conditions in  $r$  are encoded in the  $r$ -derivative matrix  $\mathbf{D}_r$ . The boundary condition  $c_g \rightarrow 0$  as  $z^* \rightarrow \infty$  is naturally satisfied by the Laguerre function expansion, but the boundary condition at  $z^* = 0$  is not. For this, we use boundary bordering. Noting that  $L_n(0) = 1$ , we have

$$c_g(r_k, z^* = 0) = \sum_{n=0}^{M-1} a_n(r_k) = f(r_k) \quad (25)$$

for each  $k$ . Thus, we redefine  $\mathbf{K}$  and define the boundary data vector  $\mathbf{c}$  as

$$(\mathbf{K})_{i,j} = \begin{cases} 1 & J_i = M, I_i = I_j \\ 0 & J_i = M, I_i \neq I_j \\ (\mathbf{D}_z^* + \mathbf{D}_r^*)_{i,j} & \text{otherwise} \end{cases}, \quad (\mathbf{c})_i = \begin{cases} f(r_{I_i}) & J_i = M \\ 0 & \text{otherwise} \end{cases}, \quad (26)$$

giving the linear system

$$\mathbf{K} \mathbf{u} = \mathbf{c}. \quad (27)$$

This enforces that the quantity in equation (24) vanishes except for a set of  $N$  equations that are replaced with equation (25) for each  $k$  (the choice of  $J_i = M$  is heuristic; formally, we can replace any set of  $N$  equations). While  $\mathbf{K}$  has  $(NM)^2$  components, it is strongly banded with proper choice of the mapping vectors  $\mathbf{I}$  and  $\mathbf{J}$ , so equation (27) can be quickly solved by a banded-system solver. While this must be done every time the boundary data  $f(r)$  changes,  $\mathbf{K}$  is constant and can be precomputed (and factorized using banded LU factorization).

With equation (2), the evaporative flux  $J$  is given by

$$J = - \left. \frac{\partial c_g}{\partial z^*} \right|_{z^*=0} \approx - \sum_{n=0}^{M-1} a'_n(r). \quad (28)$$

In discrete matrix form, the coefficients  $a'_n$  are (from equation (11))

$$\mathbf{A}' = \mathbf{A} \left( -\frac{1}{2} \mathbf{I} - \mathbf{D} \right). \quad (29)$$

One then computes the sum  $-\sum_j (\mathbf{A}')_{i,j}$  to obtain the evaporative flux  $J(r_i)$  at each node position  $r_i$ .