
Radiation Balance Between Parallel Plates

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The radiation balance is covered by equations 1-3.

$$\begin{aligned} u(x) &= \varepsilon_1 \sigma T_1^4 + (1 - \varepsilon_1) \int_{-\frac{1}{2}w}^{\frac{1}{2}w} F(x, y, d) \cdot v(y) dy \\ v(y) &= \varepsilon_2 \sigma T_2^4 + (1 - \varepsilon_2) \int_{-\frac{1}{2}w}^{\frac{1}{2}w} F(x, y, d) \cdot u(x) dx \end{aligned} \tag{1}$$

$$F(x, y, d) = \frac{1}{2} \frac{d^2}{\left(d^2 + (x - y)^2\right)^{\frac{3}{2}}} \tag{2}$$

$$\begin{aligned} I_1(x) &= \int_{-\frac{1}{2}w}^{\frac{1}{2}w} F(x, y, d) \cdot v(y) dy \\ I_2(y) &= \int_{-\frac{1}{2}w}^{\frac{1}{2}w} F(x, y, d) \cdot u(x) dx \end{aligned} \tag{3}$$

DATA: $T_1 = 1000$, $T_2 = 500$, $\varepsilon_1 = 0.80$, $\varepsilon_2 = 0.60$, $\sigma = 1.7212 \cdot 10^{-9}$, $d = 1.0$, $w = 1.0$

Question 1

Equation 1 is rearranged to give equation 4.

$$\begin{aligned} u(x) - (1 - \varepsilon_1) \int_{-\frac{1}{2}w}^{\frac{1}{2}w} F(x, y, d) \cdot v(y) dy &= \varepsilon_1 \sigma T_1^4 \\ v(y) - (1 - \varepsilon_2) \int_{-\frac{1}{2}w}^{\frac{1}{2}w} F(x, y, d) \cdot u(x) dx &= \varepsilon_2 \sigma T_2^4 \end{aligned} \tag{4}$$

The integral needs to be discretized in order to make it into a system of linear equations (SLE). The trapez-method is used to obtain equation 5.

$$\begin{aligned} u(x) - (1 - \varepsilon_1) + h \cdot \left(\frac{1}{2} \cdot F(x, y_0, d) \cdot v(y_0) + \frac{1}{2} \cdot F(x, y_N, d) \cdot v(y_N) + \sum_{i=1}^{N-1} F(x, y_i, d) \cdot v(y_i) \right) &= \varepsilon_1 \sigma T_1^4 \\ v(y) - (1 - \varepsilon_2) + h \cdot \left(\frac{1}{2} \cdot F(x_0, y, d) \cdot u(x_0) + \frac{1}{2} \cdot F(x_N, y, d) \cdot u(x_N) + \sum_{i=1}^{N-1} F(x_i, y, d) \cdot u(x_i) \right) &= \varepsilon_2 \sigma T_2^4 \end{aligned} \tag{5}$$

From equation 5 a SLE in the form $A \cdot z = b$ can be made.

Equations 6-8 illustrates a SLE for $N = 4$. The C++ implementation of equation 6 and equation 8 is scaleable (see enclosed main.cpp, line 39-84, line 93-104).

Equation 7 display the returned `VecDoub` by SVD solve function.

$$A = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 & -\frac{1}{2} \cdot \beta_1 \cdot f(x_1, y_0, d) & -\beta_1 \cdot f(x_1, y_1, d) & -\beta_1 \cdot f(x_1, y_2, d) & -\beta_1 \cdot f(x_1, y_3, d) & -\frac{1}{2} \cdot \beta_1 \cdot f(x_1, y_4, d) \\ 0 & 0 & 1 & 0 & 0 & -\frac{1}{2} \cdot \beta_1 \cdot f(x_2, y_0, d) & -\beta_1 \cdot f(x_2, y_1, d) & -\beta_1 \cdot f(x_2, y_2, d) & -\beta_1 \cdot f(x_2, y_3, d) & -\frac{1}{2} \cdot \beta_1 \cdot f(x_2, y_4, d) \\ 0 & 0 & 0 & 1 & 0 & -\frac{1}{2} \cdot \beta_1 \cdot f(x_3, y_0, d) & -\beta_1 \cdot f(x_3, y_1, d) & -\beta_1 \cdot f(x_3, y_2, d) & -\beta_1 \cdot f(x_3, y_3, d) & -\frac{1}{2} \cdot \beta_1 \cdot f(x_3, y_4, d) \\ 0 & 0 & 0 & 0 & 1 & -\frac{1}{2} \cdot \beta_1 \cdot f(x_4, y_0, d) & -\beta_1 \cdot f(x_4, y_1, d) & -\beta_1 \cdot f(x_4, y_2, d) & -\beta_1 \cdot f(x_4, y_3, d) & -\frac{1}{2} \cdot \beta_1 \cdot f(x_4, y_4, d) \\ -\frac{1}{2} \cdot \beta_2 \cdot f(x_0, y_0, d) & -\beta_2 \cdot f(x_1, y_0, d) & -\beta_2 \cdot f(x_2, y_0, d) & -\beta_2 \cdot f(x_4, y_0, d) & -\frac{1}{2} \cdot \beta_2 \cdot f(x_4, y_0, d) & 1 & 0 & 0 & 0 & 0 \\ -\frac{1}{2} \cdot \beta_2 \cdot f(x_0, y_1, d) & -\beta_2 \cdot f(x_1, y_1, d) & -\beta_2 \cdot f(x_2, y_1, d) & -\beta_2 \cdot f(x_4, y_1, d) & -\frac{1}{2} \cdot \beta_2 \cdot f(x_4, y_1, d) & 0 & 1 & 0 & 0 & 0 \\ -\frac{1}{2} \cdot \beta_2 \cdot f(x_0, y_2, d) & -\beta_2 \cdot f(x_1, y_2, d) & -\beta_2 \cdot f(x_2, y_2, d) & -\beta_2 \cdot f(x_4, y_2, d) & -\frac{1}{2} \cdot \beta_2 \cdot f(x_4, y_2, d) & 0 & 0 & 1 & 0 & 0 \\ -\frac{1}{2} \cdot \beta_2 \cdot f(x_0, y_3, d) & -\beta_2 \cdot f(x_1, y_3, d) & -\beta_2 \cdot f(x_2, y_3, d) & -\beta_2 \cdot f(x_4, y_3, d) & -\frac{1}{2} \cdot \beta_2 \cdot f(x_4, y_3, d) & 0 & 0 & 0 & 1 & 0 \\ -\frac{1}{2} \cdot \beta_2 \cdot f(x_0, y_4, d) & -\beta_2 \cdot f(x_1, y_4, d) & -\beta_2 \cdot f(x_2, y_4, d) & -\beta_2 \cdot f(x_4, y_4, d) & -\frac{1}{2} \cdot \beta_2 \cdot f(x_4, y_4, d) & 0 & 0 & 0 & 0 & 1 \end{bmatrix} \quad (6)$$

where: $\beta_1 = (1 - \varepsilon_1) \cdot h$, $\beta_2 = (1 - \varepsilon_2) \cdot h$ and $h = \frac{(\frac{1}{2}w - (-\frac{1}{2}w))}{N}$

$$z = [u_0 \quad u_1 \quad u_2 \quad u_3 \quad u_4 \quad v_0 \quad v_1 \quad v_2 \quad v_3 \quad v_4]^T \quad (7)$$

$$b = [\varepsilon_1 \sigma T_1^4 \quad \varepsilon_1 \sigma T_1^4 \quad \varepsilon_1 \sigma T_1^4 \quad \varepsilon_1 \sigma T_1^4 \quad \varepsilon_1 \sigma T_1^4 \quad \varepsilon_2 \sigma T_2^4 \quad \varepsilon_2 \sigma T_2^4 \quad \varepsilon_2 \sigma T_2^4 \quad \varepsilon_2 \sigma T_2^4 \quad \varepsilon_2 \sigma T_2^4]^T \quad (8)$$

Question 2

SVD is used to solve the SLE from question 1. Results are shown in table 1. Q_1 and Q_2 needs to be calculated. A rearrangement of the equation for Q_1 is made in equation 9. A similar rearrangement is made for the equation for Q_2 .

The constants c_{11} and c_{12} belongs to the equation for Q_1 and c_{21} and c_{22} belongs to the equation for Q_2 .

$$\begin{aligned}
 Q_1 &= \int_{-\frac{1}{2}w}^{\frac{1}{2}w} (u(x) - I(x)) dx \\
 u(x) &= c_{11} + c_{12} \cdot I_1 \\
 I_1 &= \frac{u(x) - c_{11}}{c_{12}} \\
 Q_1 &= \int_{-\frac{1}{2}w}^{\frac{1}{2}w} \left(u(x) - \frac{u(x) - c_{11}}{c_{12}} \right) dx
 \end{aligned} \tag{9}$$

$$\begin{aligned}
 c_{11} &= \varepsilon_1 \cdot \sigma \cdot T_1^4 \\
 c_{12} &= (1 - \varepsilon_1) \\
 c_{21} &= \varepsilon_2 \cdot \sigma \cdot T_2^4 \\
 c_{22} &= (1 - \varepsilon_2)
 \end{aligned} \tag{10}$$

The integrals in equation 9 need to be discretized. This is done with the trapez-method and the resulting equations can be seen in equation 11. The results for Q_1 and Q_2 can be seen in table 2.

$$\begin{aligned}
 Q_1 &= \frac{1}{2} \cdot h \cdot \left(u(x_0) - \left(\frac{u(x_0) - c_{11}}{c_{12}} \right) + u(x_N) - \left(\frac{u(x_N) - c_{11}}{c_{12}} \right) \right) + h \sum_{i=1}^{N-1} u(x_i) - \left(\frac{u(x_i) - c_{11}}{c_{12}} \right) \\
 Q_2 &= \frac{1}{2} \cdot h \cdot \left(v(y_0) - \left(\frac{v(y_0) - c_{21}}{c_{22}} \right) + v(y_N) - \left(\frac{v(y_N) - c_{21}}{c_{22}} \right) \right) + h \sum_{i=1}^{N-1} v(y_i) - \left(\frac{v(y_i) - c_{21}}{c_{22}} \right)
 \end{aligned} \tag{11}$$

The values for $u(x)$ and $v(y)$, $x = y = \pm 0.5$, $x = y = \pm 0.25$ and $x = y = 0$ for a given N is showed in table 1.

N	$u(-0.50)$	$u(-0.25)$	$u(0.0)$	$u(0.25)$	$u(0.50)$	$v(-0.50)$	$v(-0.25)$	$v(0.0)$	$v(0.25)$	$v(0.50)$
4	1.39015295e+03	1.39407857e+03	1.39407857e+03	1.39407857e+03	1.39015295e+03	2.60504838e+02	2.97020995e+02	3.11005259e+02	2.97020995e+02	2.60504838e+02
8	1.39040598e+03	1.39560589e+03	1.39444651e+03	1.39444651e+03	1.39040598e+03	2.61150040e+02	2.82020745e+02	3.12971204e+02	3.09290227e+02	2.61150040e+02
16	1.39046959e+03	1.39600813e+03	1.39453871e+03	1.39453871e+03	1.39046959e+03	2.61311714e+02	2.72227782e+02	3.13460057e+02	3.12528101e+02	2.61311714e+02
32	1.39048552e+03	1.39610999e+03	1.39456177e+03	1.39456177e+03	1.39048552e+03	2.61352155e+02	2.66906623e+02	3.13582105e+02	3.13348382e+02	2.61352155e+02
64	1.39048950e+03	1.39613554e+03	1.39456754e+03	1.39456754e+03	1.39048950e+03	2.61362267e+02	2.64160578e+02	3.13612607e+02	3.13554131e+02	2.61362267e+02
128	1.39049050e+03	1.39614193e+03	1.39456898e+03	1.39456898e+03	1.39049050e+03	2.61364795e+02	2.62768839e+02	3.13620232e+02	3.13605610e+02	2.61364795e+02
256	1.39049075e+03	1.39614353e+03	1.39456934e+03	1.39456934e+03	1.39049075e+03	2.61365427e+02	2.62068623e+02	3.13622138e+02	3.13618483e+02	2.61365427e+02
512	1.39049081e+03	1.39614393e+03	1.39456943e+03	1.39456943e+03	1.39049081e+03	2.61365585e+02	2.61717471e+02	3.13622615e+02	3.13621701e+02	2.61365585e+02
1024	1.39049083e+03	1.39614403e+03	1.39456945e+03	1.39456945e+03	1.39049083e+03	2.61365624e+02	2.61541638e+02	3.13622734e+02	3.13622505e+02	2.61365624e+02

Table 1: Values for $u(x)$ and $v(y)$, $x = y = \pm 0.5$, $x = y = \pm 0.25$ and $x = y = 0$ for a given N .

Question 3

The Richardson α^k estimate and Richardson error estimate is shown in 12. The α^k estimate is seen in table 2 to converge to 4. Thus $\alpha^k = 4$ is used in the Richarson error estimate.

$$\begin{aligned}\alpha^k &= \frac{S(h_1) - S(h_2)}{S(h_2) - S(h_3)} \\ EE &= \frac{S(h_2) - S(h_1)}{\alpha^k - 1}\end{aligned}\tag{12}$$

Q_1 , Q_2 , EE and α^k is calculated for a given N , and the results can be seen in table 2. It is clearly seen that the estimated error decreases when N is increased.

N	Q_1	α_1^k	EE_1	Q_2	α_2^k	EE_2
4	1.27409460e+03	-0.00000000e+00	4.24698201e+02	-2.76582033e+02	0.00000000e+00	-9.21940109e+01
8	1.27208978e+03	-6.35513621e+02	-6.68275530e-01	-2.80870375e+02	6.44962508e+01	-1.42944760e+00
16	1.27158320e+03	3.95762143e+00	-1.68857871e-01	-2.81952494e+02	3.96291264e+00	-3.60706311e-01
32	1.27145623e+03	3.98967651e+00	-4.23237000e-02	-2.82223632e+02	3.99103702e+00	-9.03790941e-02
64	1.27142447e+03	3.99743475e+00	-1.05877150e-02	-2.82291454e+02	3.99777714e+00	-2.26073368e-02
128	1.27141653e+03	3.99935965e+00	-2.64735257e-03	-2.82308412e+02	3.99944538e+00	-5.65261796e-03
256	1.27141454e+03	3.99983995e+00	-6.61864625e-04	-2.82312651e+02	3.99986141e+00	-1.41320345e-03
512	1.27141405e+03	3.99996020e+00	-1.65467803e-04	-2.82313711e+02	3.99996540e+00	-3.53303919e-04
1024	1.27141392e+03	3.99999013e+00	-4.13670527e-05	-2.82313976e+02	3.99999125e+00	-8.83261731e-05

Table 2: Q_1 , Q_2 , EE and α^k is calculated for a given N .

The code¹ for solving the SLE, calculating Q_1 and Q_2 and calculating the Richardson α^k and error estimate is shown in code snippet 1.

¹The complete code implementation can by found inside the zipped file.

```

139 //Solve with SVD
140 SVD obj(AA);
141 VecDoub z(NN);
142 obj.solve(bb, z);
143
144 //Calculate Q1 and Q2
145 VecDoub z_u(NN/2), z_v(NN/2);
146 for(int i = 0; i < NN; i++){
147     if(i < NN/2){
148         z_u[i] = z[i];
149     }
150     else{
151         z_v[i - NN/2] = z[i];
152     }
153 }
154 Q1_lastlast = Q1_last;
155 Q2_lastlast = Q2_last;
156 Q1_last = Q1;
157 Q2_last = Q2;
158
159 //Calculate Q1 and Q2
160 Q1 = Q(z_u, const1_1, const1_2, h);
161 Q2 = Q(z_v, const2_1, const2_2, h);
162
163 //Richardson alpha_k estimate
164 alpha_k1 = (Q1_lastlast - Q1_last) / (Q1_last - Q1);
165 alpha_k2 = (Q2_lastlast - Q2_last) / (Q2_last - Q2);
166
167
168 // It is seen that both alpha_k converges to 4!
169 alpha_k = 4;
170
171 //Richardson error estimate
172 error1 = (Q1 - Q1_last) / (alpha_k - 1);
173 error2 = (Q2 - Q2_last) / (alpha_k - 1);

```

Code Snippet 1: Code snippet of main.cpp, line 139-173.