

A New Efficient Algorithm for Validating Chebyshev Approximations of LODE Solutions

Nicolas BRISEBARRE, Mioara JOLDES and Florent BREHARD

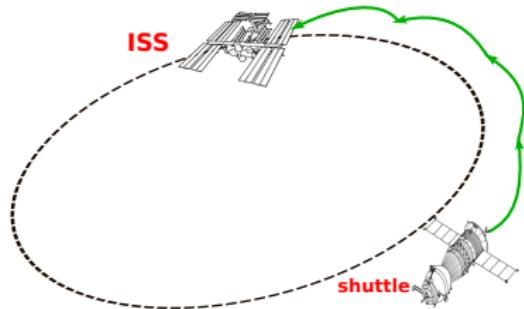
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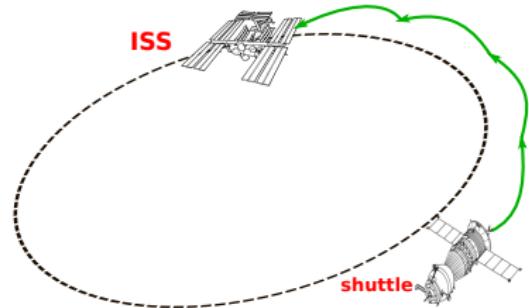
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The Spacecraft Rendez-Vous Problem



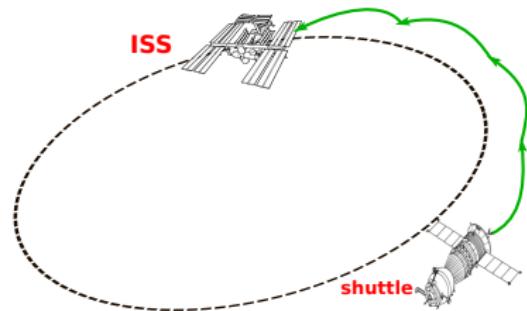
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Linearized Equation of the In-Plane Motion

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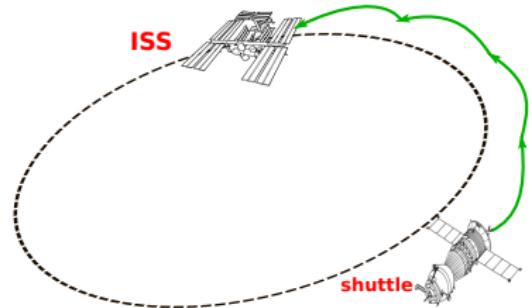


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- Approximating solutions with polynomials.
- Validating approximate solutions with certified error bounds.

1 Some Background in Linear Differential Equations

2 Approximating Functions with Chebyshev Series

3 The Validation Algorithm

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Linear Ordinary Differential Equations

- Linear Ordinary Differential Equation over compact interval I :

$$f^{(r)}(t) + a_{r-1}(t)f^{(r-1)}(t) + \cdots + a_1(t)f'(t) + a_0(t)f(t) = g(t).$$

The functions a_i and g are supposed to be at least continuous.

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- This is an infinite-dimensional linear problem:

$$\mathbf{L} = \partial^r + a_{r-1}\partial^{r-1} + \cdots + a_1\partial + a_0 : \mathcal{C}^r(I) \rightarrow \mathcal{C}^0(I),$$

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Example

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$$\mathbf{L} = \partial^2 + 4 - \frac{3}{1 + e \cos t}$$

$$\mathbf{B} \cdot z = (z(t_0), z'(t_0))$$

Existence and Uniqueness of the Solution

Theorem (Picard-Lindelöf – linear case)

The linear operator:

$$(\mathbf{L}, \mathbf{B}_{t_0}) : \mathcal{C}^r(I) \rightarrow \mathcal{C}^0(I) \times \mathbb{R}^r,$$

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Some problems:

- How to find approximate solutions?
- How to bound the error of an approximate solution?

Previous Works

- Spectral methods:
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- A posteriori validation methods for differential equations:
 - Quasi-Newton fixed-point methods (Yamamoto, Lessard)
 - D-finite approach and iteration method (Benoit, Joldes, Mezzarobba)

Reformulation with Integral Operator

Let $\varphi = f^{(r)} \in \mathcal{C}^0(I)$ with $f(t_0) = v_0 \dots f^{(r-1)}(t_0) = v_{r-1}$.
Then for $i \in [0, r - 1]$:

$$f^{(i)}(t) = \sum_{j=i}^{r-1} \frac{(t - t_0)^{j-i}}{(j - i)!} v_j + \int_{t_0}^t \int_{t_0}^{s_1} \dots \int_{t_0}^{s_{r-1-i}} \varphi ds_1 \dots ds_{r-i}.$$

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$$\varphi + \mathbf{K} \cdot \varphi = \psi,$$

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- $\psi(t) = g(t) + (\text{some function depending on the } v_j \text{'s}).$

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- $\psi(t) = c - (z(t_0) + (t-t_0) z'(t_0)) \left(4 - \frac{3}{1+e \cos t} \right).$

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- The Chebyshev family of polynomials:

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$$T_{n+2}(X) = 2XT_{n+1}(X) - T_n(X).$$

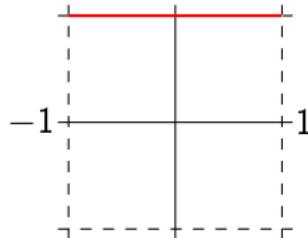
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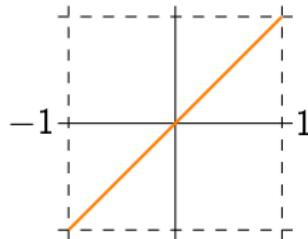
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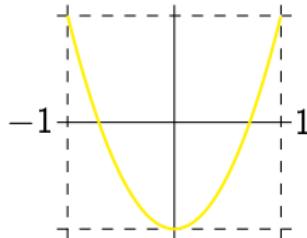
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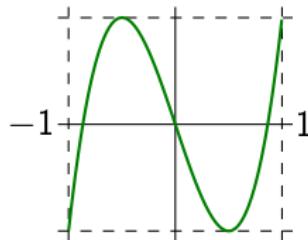
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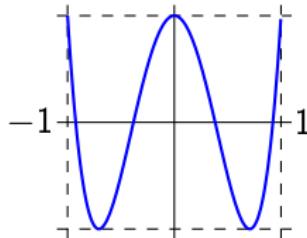
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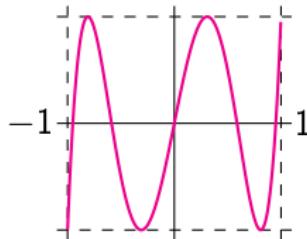
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$$T_5(X) = 16X^5 - 20X^3 + 5X$$

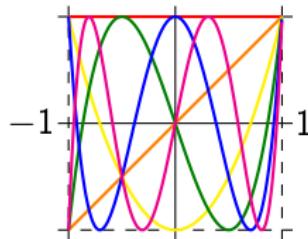
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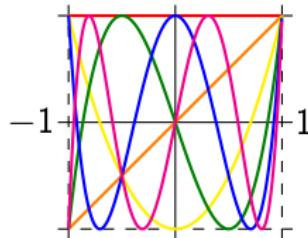
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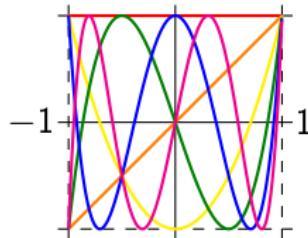
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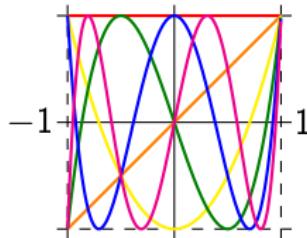
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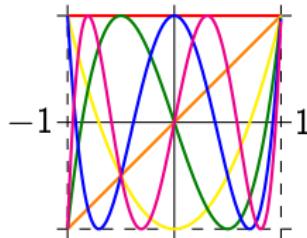
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- Integration:

$$\int T_n = \frac{1}{2} \left(\frac{T_{n+1}}{n+1} - \frac{T_{n-1}}{n-1} \right).$$

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- Chebyshev coefficients:

$$a_n = \frac{1}{\pi} \int_0^\pi f(\cos \vartheta) \cos n\vartheta d\vartheta, \quad n \in \mathbb{Z}.$$

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$$\langle T_n, T_m \rangle = \begin{cases} 0 & \text{if } n \neq \pm m, \\ \pi & \text{if } n = m = 0, \\ \frac{\pi}{2} & \text{if } n = \pm m \neq 0. \end{cases}$$

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$$a_n = \frac{1}{\pi} \int_0^\pi f(\cos \vartheta) \cos n\vartheta d\vartheta, \quad n \in \mathbb{Z}.$$

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$$\hat{f}(t) = \sum_{n \in \mathbb{Z}} a_n T_n(t), \quad t \in [-1, 1].$$

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Convergence Theorems for Chebyshev Series

- Let $\hat{f}^{[N]} = \sum_{|n| \leq N} a_n T_n$.

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$$\mathcal{C}^1 \subset \Psi^1 \subset \mathcal{C}^0$$

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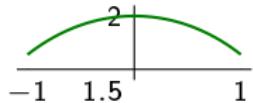
$$\mathcal{C}^1 \subset \Psi^1 \subset \mathcal{C}^0$$

Theorem

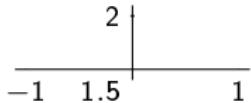
If f is \mathcal{C}^r ($r \geq 1$), then $a_n = O(n^{-r})$.

Approximating our Example

- Approximation of $t \mapsto 4 - \frac{3}{1 + e \cos t}$ over $[-1, 1]$ ($e = 0.5$):

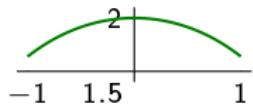


$\alpha(t)$

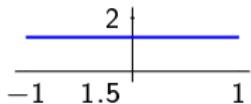


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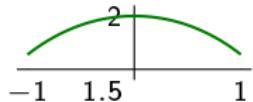


$$1.82$$

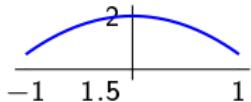
$$|\alpha(t) - 1.82| \leq 0.2$$

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$$\alpha(t)$$



$$1.82 - 0.18 T_2(t)$$

$$|\alpha(t) - (1.82 - 0.18 T_2(t))| \leq 0.007$$

3 The Validation Algorithm

The Almost-Banded Structure of the Operator \mathbf{K}

$$\mathbf{K} \cdot \varphi(t) = \sum_{j=0}^{r-1} a_j(t) \int_{t_0}^t \frac{(t-s)^{r-1-j}}{(r-1-j)!} \varphi(s) ds$$

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T_i



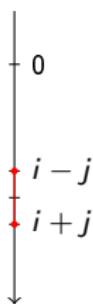
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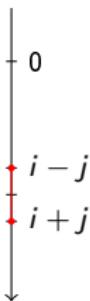
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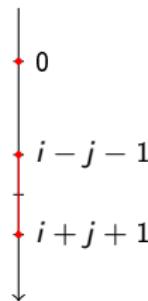
T_i



$T_j T_i$



$\int_{t_0}^t T_j T_i$



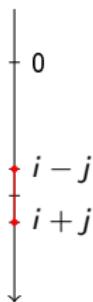
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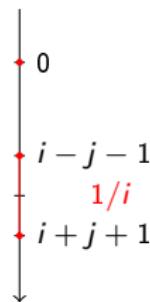
T_i



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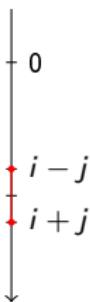
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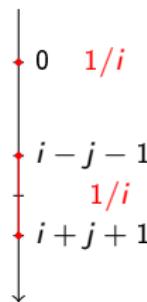
T_i



$T_j T_i$

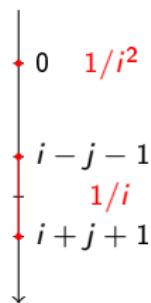
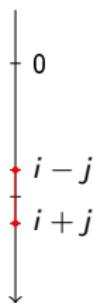


$\int_{t_0}^t T_j T_i$



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 T_i $T_j T_i$ $\int_{-1}^t T_j T_i$ 

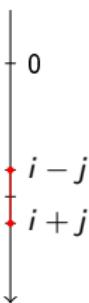
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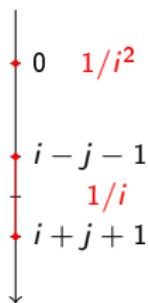
T_i



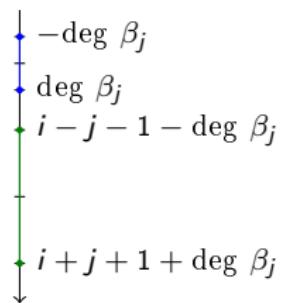
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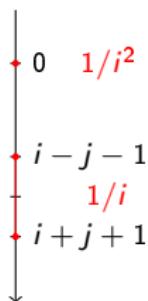
T_i



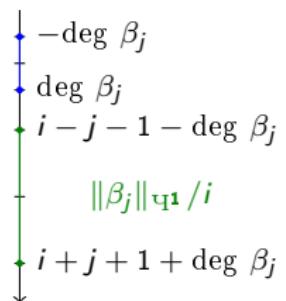
$T_j T_i$



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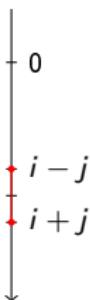
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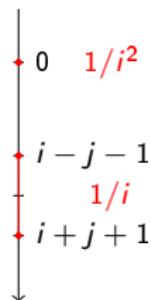
T_i



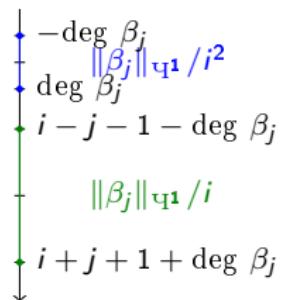
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$\int_{-1}^t T_j T_i$

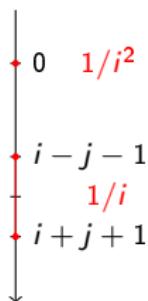
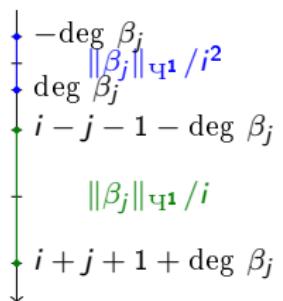


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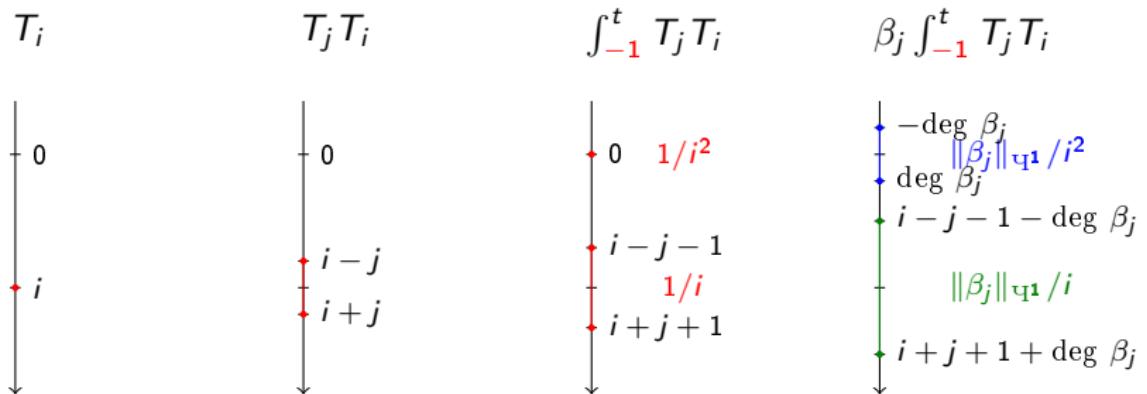
 T_i  $T_j T_i$  $\int_{-1}^t T_j T_i$  $\beta_j \int_{-1}^t T_j T_i$ 

$$h = \max_{0 \leq j < r} \deg \beta_j,$$

$$d = \max_{0 \leq j < r} (j+1 + \deg \beta_j).$$

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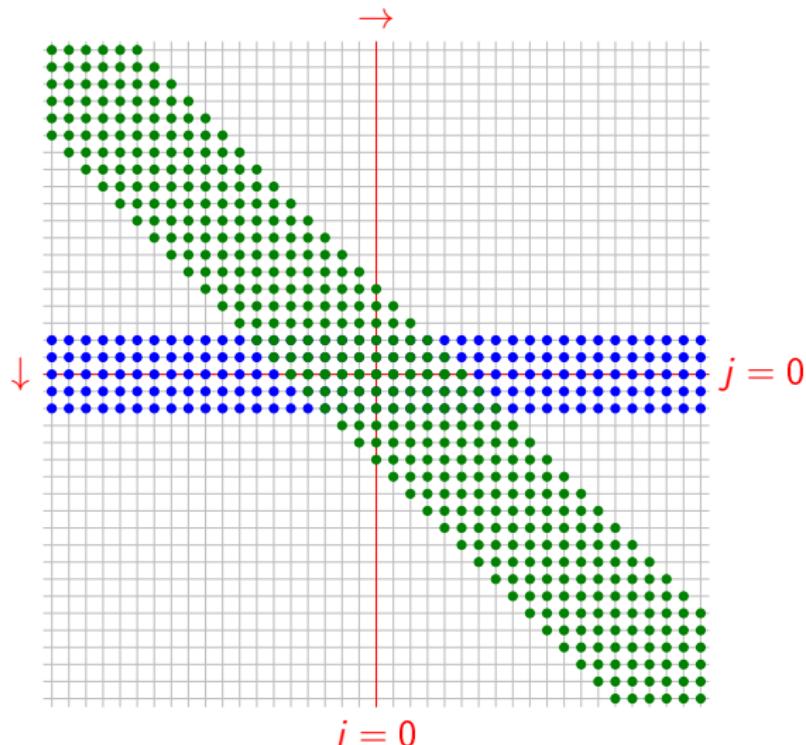


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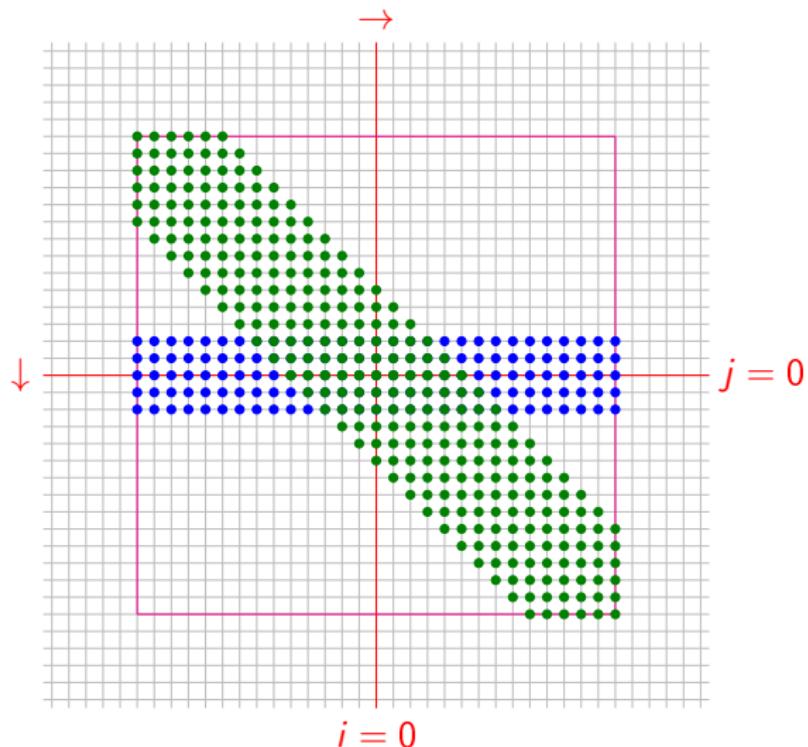
$\Rightarrow \mathbf{K}$ is an (h, d) -almost banded operator.

The Almost-Banded Structure of the Operator \mathbf{K}



The infinite-dimensional operator \mathbf{K} .

The Almost-Banded Structure of the Operator \mathbf{K}



The final-dimensional truncation $\mathbf{K}^{[N]}$.

The Almost-Banded Structure of the Operator \mathbf{K} - Example

$$\mathbf{K} \cdot \varphi = t \left(4 - \frac{3}{1 + e \cos t} \right) \int_{t_0}^t \varphi(s) ds + \left(-4 + \frac{3}{1 + e \cos t} \right) \int_{t_0}^t s \varphi(s) ds$$

The Almost-Banded Structure of the Operator \mathbf{K} - Example

$$\mathbf{K} \cdot \varphi \approx t(1.82 - 0.18 T_2(t)) \int_{t_0}^t \varphi(s) ds + (-1.82 + 0.18 T_2(t)) \int_{t_0}^t s \varphi(s) ds$$

The Almost-Banded Structure of the Operator \mathbf{K} - Example

$$\mathbf{K} \cdot \varphi \approx \underbrace{(1.73 T_1(t) - 0.09 T_3(t))}_{\beta_0(t)} \int_{t_0}^t \varphi(s) ds + \underbrace{(-1.82 + 0.18 T_2(t))}_{\beta_1(t)} \int_{t_0}^t s \varphi(s) ds$$

The Almost-Banded Structure of the Operator K - Example

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1.342500	-0.606667	-0.326250	0.364000	-0.155417	0.086667	-0.056875	0.040444	-0.030333	0.023636	-0.018958	0.015556	-0.013090	0.010398	-0.005479	0.008235	-0.007222	0.006386	-0.005687	0.005098	-0.004596
1.730000	-0.652500	-0.576667	0.438125	-0.115333	0.070208	-0.045429	0.036042	-0.027469	0.021625	-0.017475	0.014417	-0.012098	0.010298	-0.008872	0.007723	-0.006784	0.006097	-0.005356	0.004806	-0.004336
0.320000	0.069000	-0.271158	-0.036000	0.093833	-0.060571	0.004580	-0.004800	0.003900	-0.002338	0.001875	-0.001538	0.001286	-0.001091	0.000937	-0.000814	0.000714	-0.000632	0.000562	-0.000504	0.000455
-0.090000	0.109583	0.030000	-0.137375	0.006000	0.036042	0.002571	-0.002625	0.001429	-0.001125	0.000669	-0.000750	0.000629	-0.000565	0.000462	-0.000462	0.000353	-0.000313	0.000279	-0.000259	0.000226
-0.022500	0	0.052917	0	-0.065167	0	0.024036	0	-0.000536	0	0	0	0	0	0	0	0	0	0	0	0
0	-0.003750	0	0.028375	0	-0.040327	0	0.016184	0	-0.009402	0	0	0	0	0	0	0	0	0	0	0
0	0	-0.018750	0	0.018167	0	-0.027527	0	0.011548	0	-0.009312	0	0	0	0	0	0	0	0	0	0
0	0	0	-0.011125	0	0.012708	0	-0.020921	0	0.008680	0	-0.009259	0	0	0	0	0	0	0	0	0
0	0	0	0	-0.007050	0	0.009411	0	-0.015239	0	0.006774	0	-0.005205	0	0	0	0	0	0	0	0
0	0	0	0	0	-0.000536	0	0.007257	0	-0.011981	0	0.005431	0	-0.000170	0	0	0	0	0	0	0
0	0	0	0	0	0	-0.000402	0	0.005770	0	-0.003675	0	0.004451	0	-0.001444	0	0	0	0	0	0
0	0	0	0	0	0	0	-0.000312	0	0.004659	0	-0.007978	0	0.003715	0	-0.000124	0	0	0	0	0
0	0	0	0	0	0	0	0	-0.000250	0	0.003302	0	-0.006632	0	0.003147	0	-0.000107	0	0	0	0
0	0	0	0	0	0	0	0	-0.000205	0	0.003292	0	-0.005634	0	0.002701	0	-0.000994	0	0	0	0
0	0	0	0	0	0	0	0	0	-0.000170	0	0.002815	0	-0.004905	0	0.002343	0	-0.000983	0	0	0
0	0	0	0	0	0	0	0	0	0	-0.000144	0	0.002435	0	-0.004269	0	0.002052	0	-0.000974	0	0
0	0	0	0	0	0	0	0	0	0	-0.000124	0	0.002127	0	-0.003749	0	0.001812	0	-0.000965	0	0
0	0	0	0	0	0	0	0	0	0	-0.000107	0	0.001874	0	-0.003319	0	0.001612	0	-0.000955	0	0
0	0	0	0	0	0	0	0	0	0	-0.000094	0	0.001663	0	-0.002959	0	0.001443	0	-0.000944	0	0
0	0	0	0	0	0	0	0	0	0	-0.000083	0	0.001487	0	-0.002655	0	-0.002395	0	-0.002337	0	-0.002329

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$$\begin{aligned}\tilde{\varphi} = & -0.6 T_0 - 1.19 T_1 + 0.62 T_2 + 0.17 T_3 - 0.05 T_4 - 0.01 T_5 \\ & + 2.1 \cdot 10^{-3} T_6 + 3.2 \cdot 10^{-3} T_7 - 5.8 \cdot 10^{-5} T_8 - 7.6 \cdot 10^{-6} T_9 + 1.2 \cdot 10^{-6} T_{10} \\ & + 1.4 \cdot 10^{-7} T_{11} - 1.9 \cdot 10^{-8} T_{12} - 2.0 \cdot 10^{-9} T_{13} + 2.6 \cdot 10^{-10} T_{14} + 2.5 \cdot 10^{-11} T_{15} \\ & - 3.0 \cdot 10^{-12} T_{16} - 2.6 \cdot 10^{-13} T_{17} + 3.0 \cdot 10^{-14} T_{18} + 2.5 \cdot 10^{-15} T_{19} - 2.6 \cdot 10^{-16} T_{20}\end{aligned}$$

General Ideas for Validation of Linear Problems

- Recall: For the integral equation of unknown φ

$$(\mathbf{I} + \mathbf{K}) \cdot \varphi = \psi,$$

we want to validate an approximate solution $\tilde{\varphi}$:

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- If $\|\mathbf{D}\mathbf{T}\|_{\Psi^1} = \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} = k < 1$, \mathbf{T} is contractive and we get a tight enclosure of the approximation error:

$$\frac{\|\mathbf{T} \cdot \tilde{\varphi} - \tilde{\varphi}\|_{\Psi^1}}{1+k} \leq \|\tilde{\varphi} - \varphi^*\|_{\Psi^1} \leq \frac{\|\mathbf{T} \cdot \tilde{\varphi} - \tilde{\varphi}\|_{\Psi^1}}{1-k}.$$

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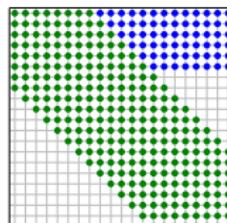
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- Computing an (h', d') almost-banded approximate inverse: $O(n(h' + d')(h + d))$.



Approximate Inverse for our Example

0.161442	0.292063	0.389506	-0.202732	0.032680	-0.025353	0.022468	-0.016198	0.012170	-0.009532	0.007748	-0.006186	0.005355	-0.004555	0.003922	-0.003413	0.002957	-0.002652	0.002364	-0.002123	0.001915
-0.951737	1.182179	0.521477	-0.177420	-0.063659	0.010819	0.004834	-0.002709	0.002226	-0.002039	0.001797	-0.001567	0.001369	-0.001201	0.001059	-0.000933	0.000836	-0.000749	0.000675	-0.000611	0.000555
0.013779	-0.233695	1.197801	0.162217	-0.147569	0.030291	-0.013394	0.012951	-0.009866	0.007675	-0.006197	0.005110	-0.004285	0.003645	-0.003138	0.002731	-0.002398	0.002122	-0.001892	0.001659	-0.001532
0.137573	-0.112650	-0.075379	1.156558	0.009202	-0.048363	-0.000583	0.002626	-0.000322	0.000261	-0.000260	0.000227	-0.000198	0.000174	-0.000153	0.000136	-0.000121	0.000108	-0.000098	0.000088	-0.000080
0.093187	0.020279	-0.069441	-0.014077	1.079375	-0.001761	-0.025165	-0.001125	0.001745	-0.000666	0.000523	-0.000443	0.000372	-0.000315	0.000272	-0.000237	0.000208	-0.000184	0.000164	-0.000147	0.000133
-0.007791	0.067936	0.004269	-0.034520	-0.009521	1.043723	0.000933	-0.017202	0.000918	0.000559	0.000015	-0.000020	0.000011	-0.000010	0.000009	-0.000008	0.000007	-0.000005	0.000005	-0.000005	0.000005
-0.000931	-0.000630	0.003429	0.000576	-0.020468	0.000072	1.028668	0.000046	-0.012114	0.000027	0.000035	0.000018	-0.000020	0.000013	-0.000011	0.000010	-0.000005	0.000008	-0.000007	0.000006	-0.000005
0.000259	-0.000232	-0.000142	0.001781	0.000017	-0.013596	-0.000001	1.026722	-0.000001	-0.000081	-0.000000	0.000000	-0.000007	-0.000000	-0.000002	-0.000000	0.000000	-0.000000	0.000000	-0.000000	
0.000093	0.000023	-0.000079	-0.000016	0.001818	-0.000002	-0.009855	-0.000001	1.015624	-0.000001	-0.006950	-0.000001	0.000241	-0.000000	-0.000001	-0.000000	0.000000	-0.000000	0.000000	-0.000000	
-0.000006	0.000006	0.000003	-0.000032	-0.000008	0.000066	0.000000	-0.007508	0.000000	1.012219	0.000000	-0.005544	0.000000	0.000024	0.000000	-0.000001	0.000000	0.000000	0.000000	-0.000000	
-0.000000	-0.000000	0.000002	0.000000	-0.000014	0.000000	0.000475	0.000000	-0.005523	0.000000	1.005628	0.000000	-0.004527	0.000000	0.000161	0.000000	-0.000001	0.000000	0.000000	0.000000	-0.000000
0.000000	-0.000000	-0.000000	0.000001	0.000000	-0.000007	-0.000000	0.000357	-0.000000	-0.004798	-0.000000	1.000081	-0.000000	-0.003767	-0.000000	0.000135	-0.000000	-0.000001	-0.000000	0.000000	-0.000000
0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000000	-0.000004	-0.000000	0.000279	-0.000000	-0.003969	-0.000000	1.006764	-0.000000	-0.003185	-0.000000	0.000116	-0.000000	-0.000000	-0.000000	0.000000
-0.000000	0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000003	0.000000	0.000224	0.000000	-0.003339	0.000000	1.005746	0.000000	-0.002729	0.000000	0.000100	0.000000	-0.000000	0.000000	-0.000000
-0.000000	-0.000000	0.000000	0.000000	-0.000000	0.000000	0.000000	-0.000000	-0.000002	0.000000	0.000184	0.000000	-0.002849	0.000000	1.004943	0.000000	-0.002364	0.000000	0.000000	0.000000	-0.000000
0.000000	-0.000000	0.000000	0.000000	-0.000000	0.000000	0.000000	-0.000000	-0.000001	0.000000	-0.000000	0.000154	-0.000000	-0.002460	0.000000	1.004298	-0.000000	-0.002068	-0.000000	0.000077	-0.000000
0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000001	0.000000	-0.000001	0.000131	-0.000000	-0.002146	0.000000	1.003772	-0.000000	-0.001825	-0.000000	0.000069	-0.000000
-0.000000	0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000000	-0.000000	-0.000001	0.000000	-0.000001	0.000000	-0.000000	0.000000	1.003337	0.000000	-0.001622	0.000000	-0.000000	0.000000	-0.000000
-0.000000	-0.000000	0.000000	0.000000	-0.000000	0.000000	-0.000000	-0.000000	-0.000000	0.000000	-0.000000	0.000000	-0.000000	-0.000000	1.002573	0.000000	-0.001451	0.000000	-0.000000	0.000000	-0.000000
0.000000	-0.000000	0.000000	0.000000	-0.000000	0.000000	-0.000000	-0.000000	-0.000000	0.000000	-0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000000	1.002654	-0.000000	-0.001600	0.000000	-0.000000
0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000000	-0.000000	-0.000000	-0.000000	0.000000	-0.000000	0.000000	-0.000000	-0.000000	0.000000	-0.000000	0.000000	-0.000000	0.000000	-0.000000	1.002403

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-0.951737	1.182179	0.521477	-0.177420	-0.063659	0.010819	0.004034	-0.002709	0.002226	-0.002039	0.001797	-0.001567	0.001369	-0.001281	0.001059	-0.000939	0.000836	-0.000749	0.000675	-0.000611	0.000555	
0.013779	-0.238395	1.137801	0.162217	-0.147560	0.020291	-0.013304	0.012961	-0.009866	0.007675	-0.006137	0.005110	-0.004285	0.003645	-0.003138	0.002731	-0.002398	0.002122	-0.001892	0.001639	-0.001532	
0.137573	-0.112050	-0.075379	1.156558	0.009202	-0.048363	-0.000583	0.002026	0	0	0	0	0	0	0	0	0	0	0	0	0	
0.003107	0.020279	-0.060441	-0.014077	1.079575	-0.001761	-0.025165	-0.001125	0.001745	-0.000666	0.000523	0	0	0	0	0	0	0	0	0	0	
-0.007791	0.007936	0.004269	-0.034920	0	1.043723	0	-0.017202	0	0.000553	0	0	0	0	0	0	0	0	0	0	0	
0	0	0.003439	0	-0.020460	0	1.026868	0	-0.012114	0	0	0	0	0	0	0	0	0	0	0	0	
0	0	0	0.001781	0	-0.013596	0	1.020722	0	-0.008981	0	0	0	0	0	0	0	0	0	0	0	
0	0	0	0	0.001018	0	-0.009855	0	1.015624	0	-0.006950	0	0	0	0	0	0	0	0	0	0	
0	0	0	0	0	0.009666	0	-0.007568	0	1.012219	0	-0.005544	0	0	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	-0.005923	0	1.009828	0	-0.004527	0	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.004798	0	1.008081	0	-0.003767	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.003969	0	1.006764	0	-0.003185	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.003339	0	1.005746	0	-0.002729	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.002849	0	1.004943	0	-0.002364	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.002460	0	1.004298	0	-0.002968	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.002146	0	1.003772	0	-0.001825	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.001888	0	1.003337	0	-0.001522	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.001675	0	1.002973	0	-0.001451	0	0	0	0	0	0	0	
0	0	0	0	0	0	0	0	0	-0.001496	0	1.002654	0	-0.001344	0	1.002483	0	0	0	0	0	0

Computing the Operator Norm (1/2)

- Decomposition of the operator norm:

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$

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- Addition and Multiplication are trivially handled.
- Computing Ψ^1 -norm = maximum of 1-norms of the columns.
- With (h', d') -almost-banded A : $O(n(h' + d')(h + d))$.

Computing the Operator Norm (1/2)

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Example

In our case, the approximation error is:

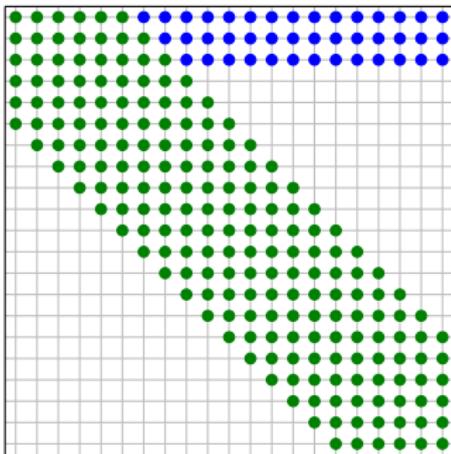
$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} \leq 1.5 \cdot 10^{-3}$$

Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\text{Op}} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\text{Op}} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\text{Op}}.$$

Computing the Operator Norm (2/2)

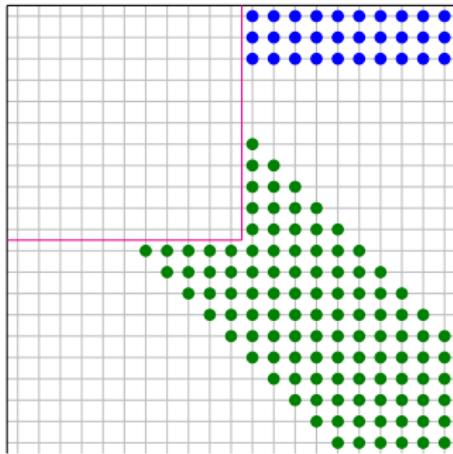
$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\text{Op}} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\text{Op}} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\text{Op}}.$$



\mathbf{K}

Computing the Operator Norm (2/2)

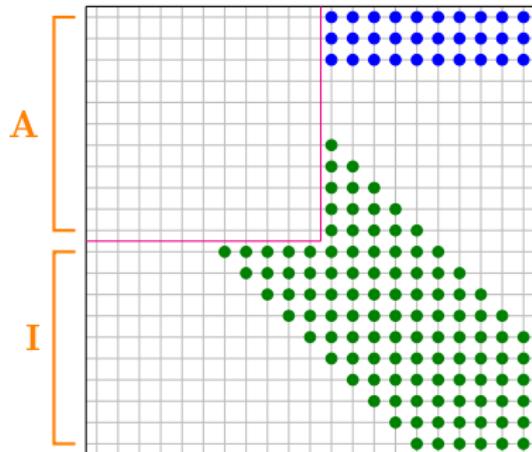
$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\text{Op}} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\text{Op}} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\text{Op}}.$$



$$\mathbf{K} - \mathbf{K}^{[N]}$$

Computing the Operator Norm (2/2)

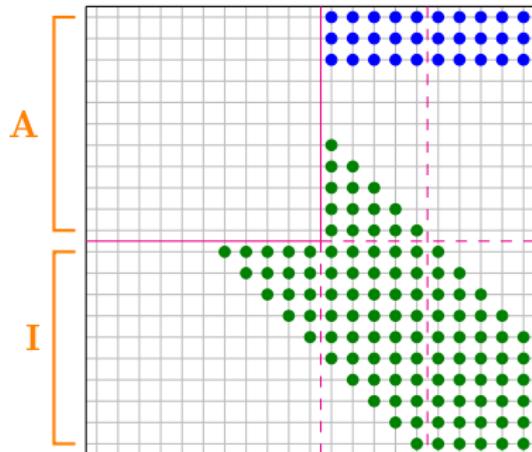
$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\text{Op}} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\text{Op}} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\text{Op}}.$$



$$\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})$$

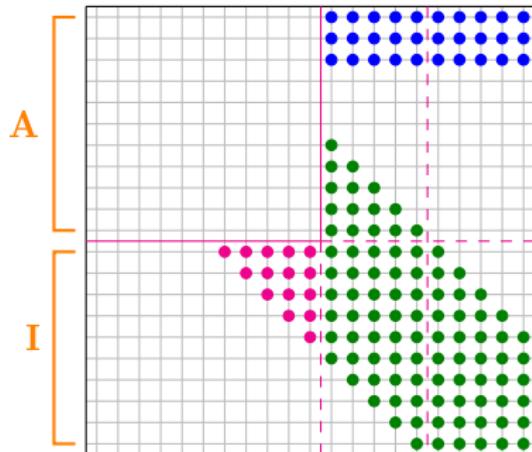
Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\text{Op}} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\text{Op}} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\text{Op}}.$$



Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$

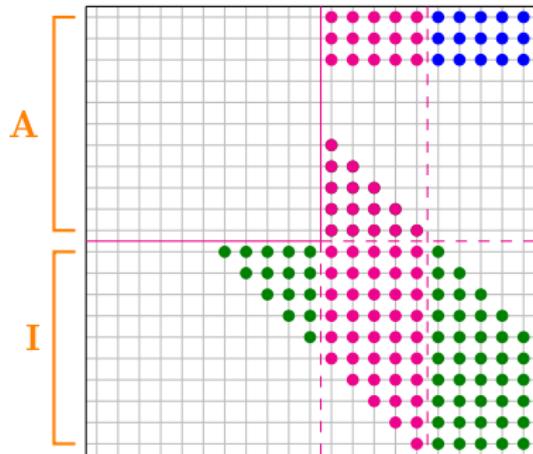


- Direct computation.

$$\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})$$

Computing the Operator Norm (2/2)

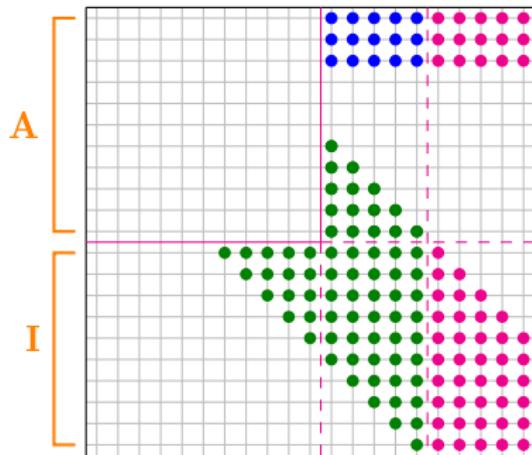
$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$



- Direct computation
- Apply \mathbf{A} and direct computation.

Computing the Operator Norm (2/2)

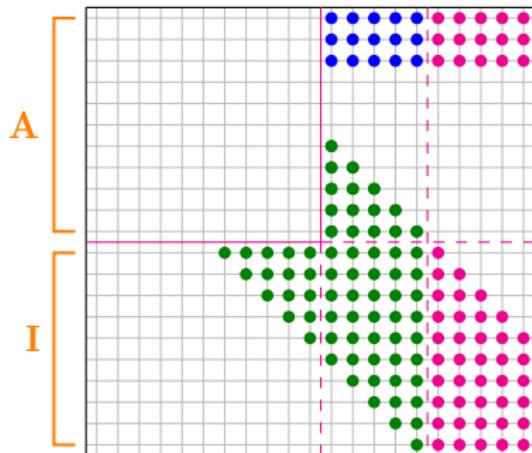
$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$



- Direct computation
- Apply \mathbf{A} and direct computation.
- Bound the remaining *infinite* number of columns:

Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$

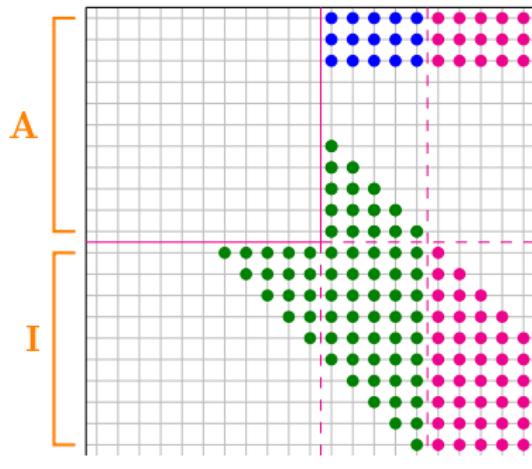


$$\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})$$

- Direct computation
- Apply \mathbf{A} and direct computation.
- Bound the remaining *infinite* number of columns:
 - Using the bounds in $1/i$ and $1/i^2$: possibly big overestimations.

Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$

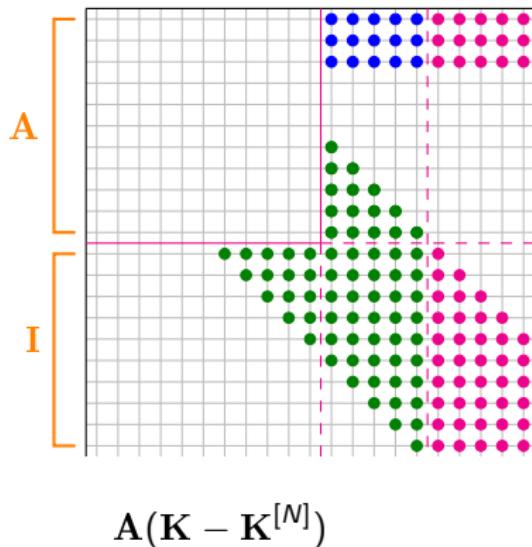


$\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})$

- Direct computation
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Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$

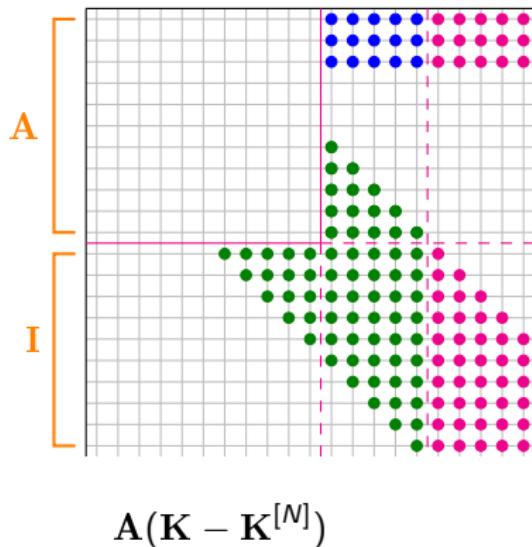


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Truncation error of the example

Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$



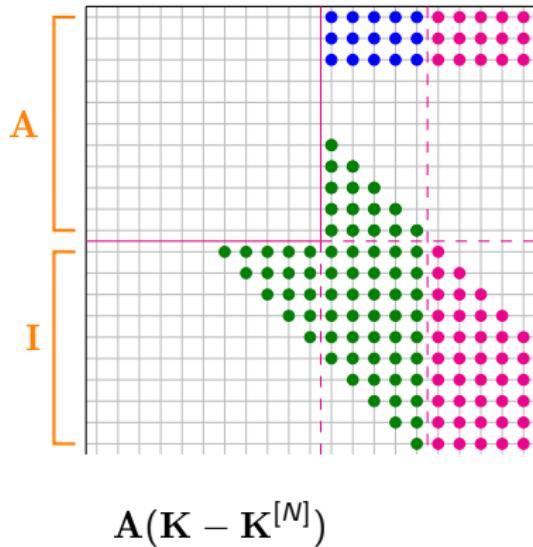
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Truncation error of the example

$1.3 \cdot 10^{-3}$

Computing the Operator Norm (2/2)

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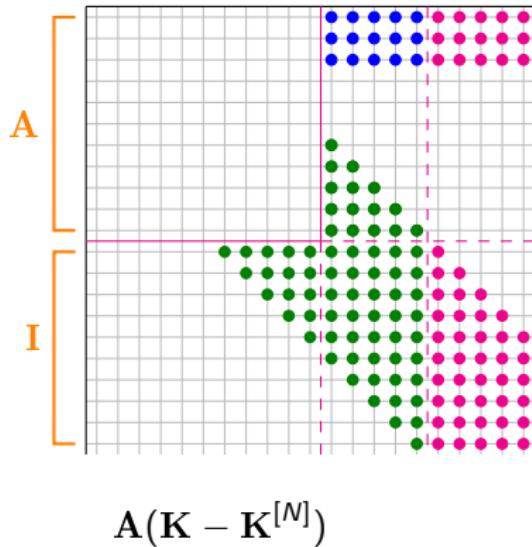
Truncation error of the example

$1.3 \cdot 10^{-3}$

$5.2 \cdot 10^{-3}$

Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$



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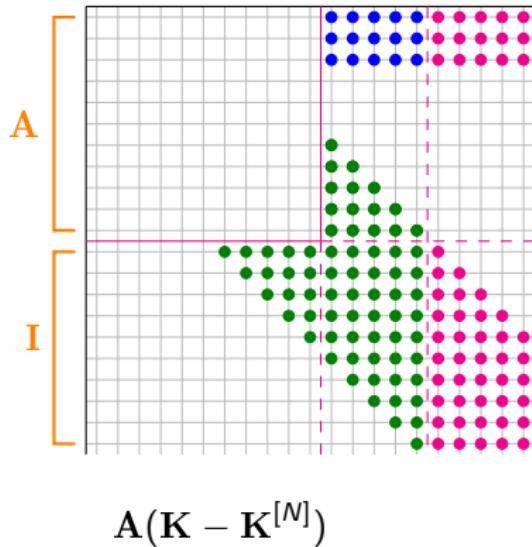
$$1.3 \cdot 10^{-3}$$

$$5.2 \cdot 10^{-3}$$

$$9.4 \cdot 10^{-3} + 2.7 \cdot 10^{-3}$$

Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$



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Truncation error of the example

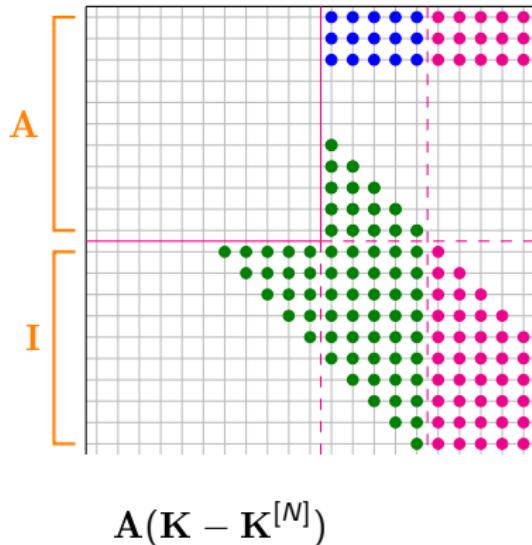
$$1.3 \cdot 10^{-3}$$

$$5.2 \cdot 10^{-3}$$

$$1.21 \cdot 10^{-2}$$

Computing the Operator Norm (2/2)

$$\|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K})\|_{\Psi^1} \leq \|\mathbf{I} - \mathbf{A}(\mathbf{I} + \mathbf{K}^{[N]})\|_{\Psi^1} + \|\mathbf{A}(\mathbf{K} - \mathbf{K}^{[N]})\|_{\Psi^1}.$$



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Truncation error of the example

$$1.3 \cdot 10^{-3}$$

$$5.2 \cdot 10^{-3}$$

$$1.21 \cdot 10^{-2}$$

$$\Rightarrow 1.21 \cdot 10^{-2}$$

Bring our Example to the End

- $k \leq 1.5 \cdot 10^{-3} + 1.21 \cdot 10^{-2}$.

Bring our Example to the End

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Bring our Example to the End

- $k \leq 1.36 \cdot 10^{-2}$.
- $\|\mathbf{T} \cdot \tilde{\varphi} - \tilde{\varphi}\|_{\Psi^1} = \|\mathbf{A}(\tilde{\varphi} + \mathbf{K} \cdot \tilde{\varphi} - \psi)\|_{\Psi^1} = 6.48 \cdot 10^{-16}$.

Bring our Example to the End

- $k \leq 1.36 \cdot 10^{-2}$.
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- Hence:

$$\frac{6.48 \cdot 10^{-16}}{1 + k} \leq \|\tilde{\varphi} - \varphi^*\|_{\Psi^1} \leq \frac{6.48 \cdot 10^{-16}}{1 - k}$$

Bring our Example to the End

- $k \leq 1.36 \cdot 10^{-2}$.
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- Hence:

$$6.39 \cdot 10^{-16} \leq \|\tilde{\varphi} - \varphi^*\|_{\Psi^1} \leq 6.57 \cdot 10^{-16}$$

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