Notes on Spectral Graph Theory

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The Laplacian matrix is the product of an oriented incident matrix, say \mathbb{M} , and its transpose. Note that this matrix is Hermitian, which is a property we will exploit in the numerical computation of eigenvalues.

$$\mathbb{L} = \mathbb{M}\mathbb{M}^{\mathsf{T}} \tag{1}$$

We can determine if all nodes within a submatrix of an oriented incident matrix is connected from its algebraic connectivity. The second eigenvalue of the Laplacian matrix of a subgraph is the algebraic connectivity, and thus the subgraph is connected if the second eigenvalue is greater than zero.

$$\Delta_{\mathbb{L}}(s) = |s\mathbb{I} - \mathbb{L}| \tag{2}$$

$$\underline{\lambda} = \Delta_{\mathbb{L}}(0) \tag{3}$$

$$\lambda_2 > 0 \Leftrightarrow \text{connected}$$
 (4)

Of course, we may also compute the transitive closure from the adjacency matrix.

$$\mathbb{C}^* = \left(\mathbb{A} + \mathbb{I}_{N \times N}\right)^N \tag{5}$$

We may exploit the fact that our Laplacian matrix is Hermitian, and use Kung's Algorithm for \mathbb{QR} factorization (qrDecomp). We may then use the \mathbb{QR} Algorithm (eigW) to compute the eigenvalues of the Laplacian matrix.

To support Kung's Algorithm, we introduce a notation for transvections, $T_{i,j}^N(x)$. The transvection is an $N \times N$ matrix with ones along the diaginal, the value x at row i and column j, and zeros elsewhere.

$$\mathbb{T}_{i,j}^{N}(x) = \mathbb{I}_{N} + x\mathbb{E}_{i,j} \tag{6}$$

Kung's \mathbb{QR} Algorithm then computes a strictly positive diagonalization matrix, \mathbb{E} , and resultant diagonal matrix \mathbb{D} . We define $\mathbb{C} = \sqrt{\mathbb{D}}$, and can write the factorization of a matrix $\mathbb{M}_{N \times N}$ as follows.

$$\mathbb{E} = \prod_{i=1}^{N} \prod_{j \neq i} T_{i,j}^{N} \left(\mathbb{M}_{i,j} \right)^{*}$$
 (7)

$$\mathbb{D} = (\mathbb{E}\mathbb{M})^* (\mathbb{M}\mathbb{E}) = \mathbb{E}^* \mathbb{M}^* \mathbb{M}\mathbb{E}$$
 (8)

$$\mathbb{C} = \sqrt{\mathbb{D}} \tag{9}$$

$$\mathbb{Q} = \mathbb{AEC}^{-1} \tag{10}$$

$$\mathbb{R} = \mathbb{C}\mathbb{E}^{-1} \tag{11}$$

Algorithm 1 Transvection

 $\overline{\mathbf{function}} \; \mathbf{TRANSVECTION}(\mathbb{M}_{N\times N}, i, j)$

▷ Initialize an identity matrix

allocate $\mathbb{T}_{N \times N} \leftarrow 0$

for $k \leftarrow 1 \dots N$ do

 $\mathbb{T}_{k,k} \leftarrow 1$

end for

▷ Conjugate and copy at i,j

 $\mathbb{T}_{i,j} \leftarrow \mathbb{M}_{i,j}^*$

 \triangleright Return the transvection

return \mathbb{T}

end function

The QR Algorithm iteratively computes the eigenvalues of a matrix using a QR decomposition. The algorithm converges when the elements in the subdiagonal approach zero.

Algorithm 2 Kung's Algorithm

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function QRDECOMP(\mathbb{M}_{N \ times N})
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 \triangleright Initialize an identity matrix

```
allocate \mathbb{E}_{N\times N}\leftarrow 0
for k\leftarrow 1\dots N do
\mathbb{E}_{k,k}\leftarrow 1
end for
```

 \triangleright Create the diagonalizer by applying conjugate pairs of transvections to remove off-diagonal elements

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\begin{aligned} & \textbf{for} \ i \leftarrow 1 \dots N \ \textbf{do} \\ & \textbf{for} \ j \leftarrow 1 \dots N \ \textbf{do} \\ & \textbf{if} \ i \neq j \ \textbf{then} \\ & \mathbb{E} = \mathbb{E} \cdot \text{Transvection}(\mathbb{M}, i, j) \\ & \textbf{end if} \\ & \textbf{end for} \end{aligned}
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 \triangleright Apply the diagonalizer to create the diagonalization

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\mathbb{D} \leftarrow \mathbb{E}^*\mathbb{M}^*\mathbb{M}\mathbb{E}
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 $\mathbb{C} \leftarrow \sqrt{\mathbb{D}}$

 \triangleright Compute the $\mathbb Q$ and $\mathbb R$ matrices

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\mathbb{Q} \leftarrow \mathbb{A}\mathbb{E}\mathbb{C}^{-1}\mathbb{R} \leftarrow \mathbb{C}\mathbb{E}^{-1}
```

 \triangleright Return the orthogonal and upper triangular decomposition

return \mathbb{Q}, \mathbb{R}

end function

Algorithm 3 QR Algorithm

 $\frac{\mathbb{I}_{\mathrm{function}} \mathbb{E} \mathbb{I}_{\mathrm{GW}}(\mathbb{A}_{0}^{N \times N})}{\mathbb{A}_{0}^{N \times N}}$

 \triangleright Set the initial conditions

$$k \leftarrow 0$$

$$Q_0, \mathbb{R}_0 \leftarrow qrDecomp(A_0)$$

 \triangleright Iterate until the subdiagonal converges to zero

while
$$\mathbb{A}_{k_{i,j}} < \epsilon, \forall j = i - 1 \forall i \in 1 \dots N$$
 do

$$k \leftarrow k+1$$

$$\mathbb{Q}_{k-1}, \mathbb{R}_{k-1} \leftarrow \text{QRDECOMP}(\mathbb{A}_{k-1})$$

$$\mathbb{A}_k \leftarrow \mathbb{R}_{k-1} \mathbb{Q}_{k-1}$$

end while

 \triangleright Create a vector from the diagonal elements of \mathbb{A}_k

$$\underline{\lambda} \leftarrow \{a_{i,j} | \forall a_{i,j} \in \mathbb{A}_k \land i = j\}$$

 \triangleright Return eigenvalues

return $\underline{\lambda}$

end function