Homework #1

"Optimal orthogonal transformation"

Mathematical Modelling

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Task 1

Equation:
$$\mathbf{Q} * \mathbf{x}_i + \mathbf{b} = \mathbf{y}_i$$
, where $i = 1, ..., n$

Given matrices X and Y:

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$$X = \begin{bmatrix} 1 & 0 \\ 0 & 1 \\ -1 & 0 \\ 0 & -1 \\ 1 & 1 \\ -1 & 1 \\ -1 & -1 \\ 1 & -1 \end{bmatrix} \quad Y = \begin{bmatrix} 1.007 & 0.207 \\ -0.407 & 0.207 \\ -0.407 & -1.207 \\ 1.007 & -1.207 \\ 0.3 & 0.914 \\ -1.114 & -0.5 \\ 0.3 & -1.914 \\ 1.714 & -0.5 \end{bmatrix}$$

Matrix Q and vector b after naïve approach:

$$Q = \begin{bmatrix} 0.70711 & 0.70711 \\ -0.70711 & 0.70711 \end{bmatrix} \quad b = \begin{bmatrix} 0.3 & -0.5 \end{bmatrix}$$

Task 2

My function naive(X, Y) takes matrices X, Y of the same dimension $n \times k$ and returns square matrix Q and vector b, as it is specified in a specification.

Overview of my solution

My idea is to simplify the work and to bring the expression

$$Qx_i + b = y_i$$
, where $i = 1, ..., n$

into the Ax = b format to improve flexibility and increase the number of possibilities I can do to get the matrix Q and vector b from the new "unknowns" matrix B.

After converting the system to the form I am comfortable with and finding the unknowns matrix B, I can easily get the rotation matrix Q and displacement vector b out of it.

In the last step, I will apply QR decomposition to make sure that the resulting Q matrix is square and orthogonal (which is what the rotation matrix should be).

Detailed Principle of action

- 1. Matrix Preparation. Let's move a little bit backwards in solving the problem. Let's imagine that we already have our convenient form X * B = Y. Now if we look more carefully, we see that in order for the left side of the expression to also represent a sum where the second operand (vector b) is not multiplied by a vector from matrix X, we need to add a column of ones to matrix X.
- **2.** Obtaining the matrix of unknowns. Then, to solve our equation and get the unknowns matrix B from the equation X * B = Y, we need to multiply both parts by the Moore Penrose inverse matrix $X \cdot B = X^{\dagger} * Y$.
- 3. Getting the values, we are looking for. Cut the matrix B. The last row of matrix B will be the vector b to be searched, and the rest of it will be matrix O.
- 4. Ensure that Q is orthogonal. Use QR decomposition to obtain a new matrix Q which is exactly orthogonal.

Task 3

$$C = U * S * V^T$$

From the Theorem Multiplication Theorem: det(AB) = det(A) * det(B). We can use the same principle in our equation: $det(C) = det(U) * det(S) * det(V^T)$.

As U and V matrices are orthogonal, then their determinants equal to +/-1

S matrix is a diagonal matrix of singular values. It means that det(S) equals to the product of singular value that are greater or equals 0, because singular value is a square root of eigenvalue.

Now we can see that as det(S) is always positive as the product of non-negative numbers, then the sign of determinant of matrix C relies on a sign of a determinant of the product of U and V^T matrices.

Task 4

The function kabsch(X, Y) takes matrices X, Y of the same dimension $\mathbf{n} \times \mathbf{k}$ and returns square matrix Q and vector b, as it is specified in a specification.

Overview of my solution

In implementing the function, I followed the tasks in the function specification behind this reference step by step.

Detailed Principle of action

- 1. Creation of matrices of centroids X and Y, where each matrix element is equal to the difference of mean of the corresponding matrix, corresponding measurement and matrix value at the specified position matrix X.
- 2. Obtaining the covariance C of a k x k matrix. C = ...
- 3. Applying Singular Value Decomposition on matrix C to partition it into 3 matrices U, S and V^T
- 4. Creating a diagonal matrix D with dimensions k x k, where the rightmost lower element is 1 or -1 depending on the sign of the matrix determinant of C or $U * V^T$
- 5. Obtaining the sought matrix Q by multiplication of matrices U, D and V^T
- 6. Obtaining the sought vector b from the formula: $\mathbf{b} = \overline{\mathbf{y}} \mathbf{Q} * \overline{\mathbf{x}}$

References

- 1. Wikipedia. Linear Least Squares: https://en.wikipedia.org/wiki/Linear_least_squares
- 2. Wikipedia. QR Decomposition: https://en.wikipedia.org/wiki/QR_decomposition
- 3. Wikipedia. Kabsch Algorithm: https://en.wikipedia.org/wiki/Kabsch algorithm