

Critical analysis of the advantages and limitations of Gaussian elimination method

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Matrix inverse calculation consumes more time for mathematicians and can bring errors, whereas using the "Gaussian elimination method" the inversion problems can help in transferring the set of equation. This way of calculation becomes a fast, accurate, and easy process to obtain the exact results. As suggested by Liu et al. (2022), Gaussian elimination can be indicated as an immediate process for executing linear equations with authenticity and bringing error-free consequences. Hence, conducting inversion with three variables can be a difficult task, and using the method of Gaussian elimination can be advantageous. On the contrary, Regev et al. (2022) opined that Gaussian elimination can be effectively conducted for computing the "factor of a square matrix", "rank of the matrix", and "inversion of the matrix". Thus, this method of elimination can be beneficial in reducing "row echelon form". In addition, this method gives a choice for computing from left to right for ordering the variables which makes it easy for deriving the actual result without any errors.

Gaussian elimination has been represented as an easy algorithm to calculate the huge matrix and that cannot be solved intuitively. This process of elimination can be highly helpful for computing determinants and inverse matrices. As depicted by Bryant et al. (2022), the method of Gaussian elimination can be useful for reducing the values of coefficients in the upper-triangular matrix to get the result. Therefore, this elimination method has been often regarded as a direct method for obtaining unknown values using "back substitution". Computing the process of inversion and multiplication in solving linear equations using Gaussian elimination has been interesting and results can be obtained easily without failure. As argued by Notarfrancesco (2022), though the Gaussian elimination process is reliable and dependable, the main limitation of this method is that the process is slow for conducting linear equations. Hence, this method can solve more than 2 linear equations concurrently, yet this process consumes more time when the matrix is huge.

The study of the existing literature has disclosed that the process of Gaussian can be denoted as a statistical model or stochastic process. This process can extend to get multivariate for gathering random variables in order to create a distribution dataset. As specified by Paredes et al. (2022), the Gaussian procedure can allow the collection of infinite variables and can be useful for executing machine learning since it is a flexible tool for computation. Thus, multivariate effectively maintains the functioning of covariance and mean values in order to observe the variables to execute the analytical process. On the contrary, Esser et al. (2022) suggested that this elimination process of Gaussian can efficiently encode any assumptions of mathematicians and can solve complex problems easily. Hence, the procedure of row transformation and reduction can be highly useful and advantageous in the method of Gaussian for evaluating coefficient matrices.

When any "pivot elements" become zero or small values, then the equation has to be rewritten in a different order to avoid the derivation of zero. This has been indicated as another limitation of the Gaussian elimination method for getting results for linear equations. On the other hand, a back-solving process in the Gaussian method makes the system solution of the linear equation the same and the outcomes can be derived easily. The reducing system through matrix algebra has presented this elimination method unique in the linear equation system. Therefore, this method has been recommended as the most efficient and performs in a better way to solve linear equations.

Discussion on features of the Gaussian elimination method

The method of Gaussian elimination has been denoted as an extremely prevalent process in computing linear equations to achieve accurate results. Computational efficiency can be achieved by performing row reduction on the matrix for measuring the algorithm in the correct way. Moreover, this elimination process restricts mathematicians to calculate both column and row calculations. While solving linear equations, mathematicians have been allowed to conduct only row operations since column operations require a change of variables. As recommended by Nikolopoulos et al. (2022),

Gaussian elimination can be expressed as "matrix multiplications" since the determinants have been multiplicative in nature for commuting linear equations. Hence, the framework of Gaussian has been represented as smooth, flexible, and reliable for executing the function of co-variance. As argued by Guo et al. (2022), reaching the point of conclusive normalization, mathematician conduct row elimination and row reduction using the Gaussian method for resolving linear systems of equations. Thus, the constants in the equations required to be multiplied by "a constant" to make the coefficients of one variable the same for both equations. At the final stage of calculation, only one variable remains which can be solved easily.

Simultaneous linear equations can be solved by making the values of co-efficient the same in both equations. This process can be applied to either adding or subtracting the equations in order to form a new comparison that contains one flexible only. Thereby, the process of eliminating the variables can be easily and flexibly executed by mathematicians without any error to obtain the final results. As opined by Bellini et al. (2022), the formation of numerical codes for executing the Gaussian elimination method for tracking row operations. Hence, "elementary row operation" can be conducted from left to right which helps mathematicians to stay focused on the computation of the matrix. Gaussian elimination has been indicated as an ingenious and famous method for solving linear equations by swapping, adding, and multiplying. In addition, this method allows mathematicians to make focus on unknown variables.

Coefficients and constant terms have been indicated as the main components of the matrix, and the elementary operations have been the main purpose of executing elimination to get the final outcome of the linear equation. As suggested by Antoniou et al. (2022), Gaussian elimination has been denoted as a prevalent technique that shows the path of conducting rank matrices. Thus, understanding the proper and accurate way to eliminate the variables has been considered the main feature of the Gaussian elimination process to acquire answers to linear equations. The operational activities of linear equations can be wholly converted to row transformation by interchanging, adding, and multiplying values of rows. On the contrary, Ullah et al. (2022) mentioned that by performing row deduction in the matrix, mathematicians can be able to fill the matrix with zeros and the left part can be denoted as a "row echelon matrix". Hence, the elimination method helps to do back substitution for deriving the solution of the bottom at first, then preparing the process to reach the top to get the final solutions.

Gaussian elimination helps the linear calculation to achieve the exact result if the application of transformation is correct for the augmented matrix. This method has been regarded as effective, has low complexity, is convenient, and requires less computation for significant problems. These features of this elimination method have been making used to solve linear equations widely to obtain authentic results.

Effects of the "Gaussian Elimination Method"

Based on the research work of Yang, Tang & Sinanoglu (2019), the "Gaussian Elimination Method" is one of the most well-known fundamental methods in linear algebra that has been used by Mathematicians for solving linear equations. In addition to that, this method has also been used by mathematicians for matrix operations. Effect of this can be far-reaching. This method can have a significant impact in various fields such as mathematics, physics as well as computer science. There are some advantages of the "Gaussian Elimination Method" that has been discussed here. Based on the research work of Yun, Cui & Ma (2019), the primary effect of the "Gaussian Elimination Method" has been its ability to solve various linear equations. In this method a given set of linear equations with some unknown variables the method applies some row operations in instruction to transform the whole scheme into an equivalent arrangement. This whole system has been done in a row-echelon system.

The concept of Row operations

Row operations are been all the calculations that have been used row matrix in order to solve the system and equations. Based on the words of Fu, Liu & Wang (2019), this row matrix has been used by mathematicians to simply row reduce the matrix for others. Generally, there are three different rows that are being used in the completion of both equations. Each and every of which will yield a row equivalent matrix. This means when a mathematician works on an augmented matrix, the solution set to the underlying system of equations will stay the same. In other words, it can be said a matrix

row operation can be referred to as an arithmetic operation or combination of such operations that can be applied to the row of the matrix in order to solve a system of linear equations. Based on the research conclusions of Doha, Hafez &Youssri (2019), matrix row operations are ones that have been used after the user converted a linear equation into a matrix equation. This method helps the user to simplify the whole issue as much as possible in order to find out proper solutions to all the variables present in the system. The system used in this method for solving the whole issue of the equation with the help of row operation is called Gaussian elimination. In total there are three matrix elementary row operations such as interchanging two rows and multiplying a row by any constant. In addition to that, adding a row to another row is another part of the row operation. From these three, the user can have multiple combinations such as multiplying any row by any constant which is not zero to add it or minus it from another row. In addition to that, in some cases, the user can exchange all these rows. As per the research work of Hustrulid, Qianyuan&Krauland (2021), “elementary row operations” on an augmented matrix never change the whole solution or set of the whole equation. In other words, it can also be said, even though the user operating on the matrix, the matrix obtained after every calculation is similar to the one the user used before. There are in total row operations that have been done by researchers all these methods have been discussed below. In addition to that, it can also add the “matrix row operations” are the basis for any row reduction. This type of operation matrix is also named a rectangular matrix. This method has also been known as the Gaussian elimination method. Therefore, the proof process will play a major role in this type of method.

Swap two rows

When researchers work with the help of the equation system that helps in the process of solving the problem., the order they write the question does not affect the whole solution. Based on the words of Chen et al. (2021), in this method, helps to showcase the system of the equation. On top of that, in this method, every row is an equation itself. In this method, the researcher can swap two rows and it will not impact the whole equation as a whole. In other words, it can be said, this is a method of row that allows the user to switch any row that they need in the solution.

In this above figure, it can be seen that the double arrow is pointing at the row sweep. When the mathematics will perform row operation, use notations such as this help to track what has been done in the mathematics. It is easy for the researcher to have some mathematics mistakes, as well as if this type of mistake happens, based on the research conclusions of Ai et al. (2019), this notation helps the researcher to go back and easily. In this image and the formal notation for this row, operations can be R1- R3. In this method, the user interchanges the row 1 and row 3. In Another way of describing this is R1- R3, where R1 shows information about Row one and R3 is showing information about R three. Exchanging the rows of a matrix produces an “Equivalent matrix” This means that the algebraic relationship remains the same for the whole calculation.

In this case, if someone looks at it in-depth, interchanging two rows in a representing the matrix that helps the methods of “linear equations”. Based on the research work of Kissinger & de Griend (2019), it can be said, the user just exchanges the whole order in which the equations have been listed in the system. This means it did not have any effect on the whole solution or the variable in the equation.

On the other hand, Ramesh et al. (2022) have argued that the rows in any augmented matrix come from any “linear system of equations”. This means each of the equations represents the whole system. Hence, the order in which the user lists a system or equation makes nothing on the proportional relationship expressed through their variables. It can also be expressed with the help of a system.

In this system, there are two equations, each has been present as “row matrix notation”. In the above figure, this does not make any difference in the second row in the x and y not present. Hence, it does not matter how the user describes them in the matrix equation as long as the user keeps them in rows. In addition to that, the user can exchange a complete row with another complete row. In other words, it can be said, based on the research conclusions of Deuflhard&Hohmann (2021), the user can exchange the order of the columns as these might change the coefficient, which has been deeply associated with any variable in the equation. Accordingly, it will change the equation.

Multiplication of the row with the help of non-zero constant

Sometimes it will be advantageous to multiply each pair by a number like 2 or $1/3$. The fundamental system of equations' whole solution could be altered by doing this because multiplying any formula by a nonzero result frequently produces a nonzero result. In a comparable mathematical issue (provided that the researcher is able to multiply every component of the solution). According to the findings of Amy & Gheorghiu (2021), this kind of row operation enables the user to multiply an entire row by any constant that is not zero. Negatives are not properly employed in these scenarios.

In this above figure, each of the entries in row 2 has been multiplied by the constant. Based on the words of East & Ripley (2021), a fair equation in this situation might arise why any mathematician does row operation. However, in this figure, the row 2 has been multiplied by 3 just for the sake of describing how it can work.

In this "Elementary row operations matrix" process the user has multiplied the non-zero constant 3 by the initial row of the original matrix. This operation can be represented by $3R_1 - R_1$. Influenced by the research conclusions of Noeiaghdam & Araghi (2020), As it can be seen, this multiplication will impact each of the coefficients in row one in a similar way one can apply such a factor to all of the terms. In equations that have been multiplied by any constant.

If in any equation the user multiplies the non-zero constant 2 by the initial row of the matrix. This can be meant the resulting equation for the row might be $2(6w-2x+8y=4) = 12w-4x+16y=8$. There it can be seen how this multiplication for non-zero constants can make changes in the value of all the variables when solving for them as the proportionality relationship remains.

Based on the research conclusions of Yen, Verteletskyi & Izmaylov (2020), this is one of the most valuable factors things to notice. Since it brings to the attention of the user how this row operation takes a deep focus on the constant multiplied in the matrix as well as different from zero. If instead of a non-zero constant, the user multiplies the initial row by zero, this can be meant the user will end up with an equation for that row equal to $0 = 0$. As a result, the user will be left with only two problems to solve in a system with three unknown variables. Hence, it will become an equation that is not possible for anyone to solve anymore.

Addition a row to another

As per the research work of Song et al. (2021), there is a third type of "matrix row operation", that consists of allowing two rows to substrates or add from each other. This figure has been presenting an addition row, where the user replaced row 2 for the result of the addition of row 1 and row 2. This type of operation has been represented by the researcher with the help of $R_1 + R_2$ to R_2 . Based on the research conclusions of Nash, Gheorghiu, and Mosca's (2020), the user can do this with any two rows present in the research method. It can be said, adding them or subtracting them from one another as well as replace on them with the result can be effective. Having seen all three Rudimentary levels of operations that is well-intentioned of noting the combination of the argument of the "matrix row operations" meaning, whenever the user has to apply a combination of the Rudimentary operations seen above into an augmented matrix.

According to the research conclusions of Hustrulid, Qianyuan & Krauland (2021), the user should always remember that one of the primaries aims of the row operations on augmented matrices is to simplify the matrices as high as possible in order to get a solution of the variables in the system equation they have been representing. Such simple matrix forms have been called Echelon forms. The reduced Echelon form and they are the key factor in finding the ultimate solutions.

Concept of Echelon form

Nash, Gheorghiu, and Mosca's (2020) research findings state that when a matrix is in Echelon form in a linear equation, this means that a "Gaussian elimination" has taken place and the matrix has taken on this shape. The matrix's row echelon form may indicate that the "Gaussian elimination process" has been applied to the matrix's row, and the matrix's pillar echelon form may show that the "Gaussian elimination" has been applied to the matrix's column. In other words, if this transpose is in "row echelon form," it can be claimed that the matrix has been preserved in column echelon form.

As a result, the remainder of the paper solely takes into account the "row echelon form". According to the research conclusions of Soos, Gocht & Meel (2020), the Parallel properties of the "column

echelon form" can be deduced with ease with the help of transposing all the matrices. "Row echelon form" refers specifically to a matrix. Regardless of who is involved, there are several requirements that must be met in this scenario, such as the need that the bottom rows of the matrix include only zeros. In addition, every row above has a leading entry that is to the right of every other row's leading entry, even if it is a non-zero entry in the leftmost area of the row. Many transcripts also include the requirement that the constant be 1. Others view this as a scaled-down "row echelon form," on the other hand.

Based on the research work of Soos, Gocht & Meel (2020), the matrix is said to be in "row echelon form" when all its non-zero commotions have a hinge that has been a non-zero entry such that all the Entrance to the left and below it is equal to zero. While the linear system's "coefficient matrix" is in "row echelon form." It is one of the systems were employing an algorithm to compute the entire solution is easier for the user. The back-substitution algorithm is another name for this technique.

According to the circumstances indicated above, which the research has illuminated, every entrance in the table below will be resulting in the value of the coefficient being zero. The following is an illustration of a "row echelon form" 4*5 matrix. No "row echelon form" has been used for the aforementioned scenario. The "row echelon form" makes it simple to infer many matrix properties, including the kernel and rank.

Reduced "row echelon form"

Based on the research discussions of Nash, Gheorghiu & Mosca (2020), any matrix is which can be referred to as the reduced "row echelon form" is based on some factors. This type of Echelon form is also recognized as the "row canonical form". There are some conditions this type of method has to follow for example this row canonical form is in "row echelon form". In addition to that, the leading entrance in the row which is not zero should be a 1, this is also known as the leading 1. On top of that, each of the columns that have contained the leading 1, should have zeros in all the other entries on the matrix.

Conclusion:

As stated in the study work of Yen, Verteletskyi, and Izmaylov (2020), Gauss-Jordan elimination can be used to quickly capture the reduced "row echelon form" of a matrix. The row form of a matrix is not the same as this kind. This Echelon matrix approach is quite distinctive. This is independent of the computation algorithm. Despite the fact that the "row echelon form" is not a novel concept, all variants of the column Echelons and the condensed "row echelon form" have nearly the same proportion of zero rows in any given matrix. Additionally, pivots are situated in identical indices. The matrix in reduced "row echelon form" shown in figure 2.11 demonstrates that the left part of the matrix has not always been an identity matrix. The "Hermite normal form" has also been a row echelon form for matrix with coefficients of integers that may be computed both with and without the use of a rational denominators and numbers. Additionally, the reduced "row echelon form" is typically a matrix with an integer coefficient. This further includes a non-integer coefficient.

References:

- Nikolopoulos, S., Kalogeris, I., Papadopoulos, V., & Stavroulakis, G. (2022). Ai-enhanced iterative solvers for accelerating the solution of large scale parametrized linear systems of equations. arXiv preprint arXiv:2207.02543. Retrieved on: 29th June,2023, Retrieved from: <https://arxiv.org/pdf/2207.02543>
- Noeiaghdam, S., & Araghi, M. A. F. (2020). A Novel Algorithm to Evaluate Definite Integrals by the Gauss-Legendre Integration Rule Based on the Stochastic Arithmetic: Application in the Model of Osmosis System. *Mathematical Modelling of Engineering Problems*, 7(4). Retrieved on: 29th June,2023, Retrieved from: <https://search.ebscohost.com/login.aspx?direct=true&profile=ehost&scope=site&authtype=crawler&jrnl=23690739&AN=147786848&h=CM5WDoZBaGNAeSTKYGc7lLwisIObEMgHZzbd581Zd0TFSRkLg8oOg6F6a7VfMJgIZnx7wqjQAOhso2IgXsZz1g%3D%3D&crl=c>
- Notarfrancesco, L. (2022). Arrows: a Computer Algebra System in Smalltalk. In *FAST Workshop 2022 on Smalltalk Related Technologies*. Retrieved on: 29th June,2023, Retrieved from: <https://openreview.net/pdf?id=oIozVmVool>

- Online.stat.psu.edu, (2023), M.7 Gauss-Jordan Elimination. Retrieved on: 29th June,2023, Retrieved from:<https://online.stat.psu.edu/statprogram/reviews/matrix-algebra/gauss-jordan-elimination>
- Onofri, A., Terzaroli, N., & Russi, L. (2021). Linear models for diallel crosses: a review with R functions. *Theoretical and Applied Genetics*, 134, 585-601. Retrieved on: 29th June,2023, Retrieved from: DOI: <https://doi.org/10.1007/s00122-020-03716-8>
- Pandey, P., & Pandey, M. M. (2021). Research methodology tools and techniques. Bridge Center. . Retrieved on: 30th June, 2023 Retrieved from: <http://dspace.vnbrims.org:13000/jspui/bitstream/123456789/4666/1/RESEARCH%20METHODOLOGY%20TOOLS%20AND%20TECHNIQUES.pdf>
- Paredes, P., Choudhari, M., Carpenter, M. H., & Li, F. (2022). The Harmonic Linearized Navier-Stokes Equations for Transition Prediction in Three-Dimensional Flows (No. ICCFD11-2702). *International Conference on Computational Fluid Dynamics*. Retrieved on: 29th June,2023, Retrieved from: https://ntrs.nasa.gov/api/citations/20220008151/downloads/Paredes_ICCFD11_final.pdf
- Parish, E. J., Wentland, C. R., & Duraisamy, K. (2020). The AdjointPetrov–Galerkin method for non-linear model reduction. *Computer Methods in Applied Mechanics and Engineering*, 365, 112991. Retrieved on: 4th July, 2023. Retrieved from: <https://www.sciencedirect.com/science/article/am/pii/S0045782520301754>
- Pascacio, P., Casteleyn, S., Torres-Sospedra, J., Lohan, E. S., & Nurmi, J. (2021). Collaborative indoor positioning systems: A systematic review. *Sensors*, 21(3), 1002. Retrieved on: 4th July, 2023. Retrieved from: <https://www.mdpi.com/1424-8220/21/3/1002/pdf>