

# **Oncogenic Role of KLF8 Variants in Breast Cancer**



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A thesis submitted to the National University of Sciences and Technology, Islamabad,

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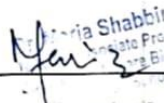
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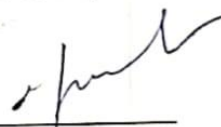
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## **DEDICATION**

This thesis is dedicated to my beloved family and my teachers who supported me throughout this process and believed in my capabilities.

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## **ABSTRACT**

Breast cancer is the most prevalent cancer among females. In Pakistan, late diagnosis of this disease leads to a higher death rate. This type of cancer involves uncontrolled division of epithelial cells surrounding ducts and lobules of breast. This uncontrolled division is mediated by transcription factors. One of the family of transcription factors is the Kruppel-like factors family. These are the zinc-finger proteins that function in cell cycle progression. However, these are also involved in various malignancies where they directly affect cell cycle regulation, apoptosis, metastasis and signaling pathways. Therefore, there is a need to study these proteins. One of the members of this family of proteins is KLF8. This factor is found to be associated with different types of cancers and other disorders. There is not enough evidence of single nucleotide polymorphism in KLF8. This study is aimed at studying the association between SNPs of KLF8 and breast cancer. The SNPs in KLF8 affect the structure and function of the KLF8 protein. In-silico analysis of the SNP rs868781835 revealed that its affect might probably be damaging, while SNP rs1196572444 was found to be damaging. This SNP was rendered to genotype analysis where the genotype GT was associated with metastasis of breast cancer.

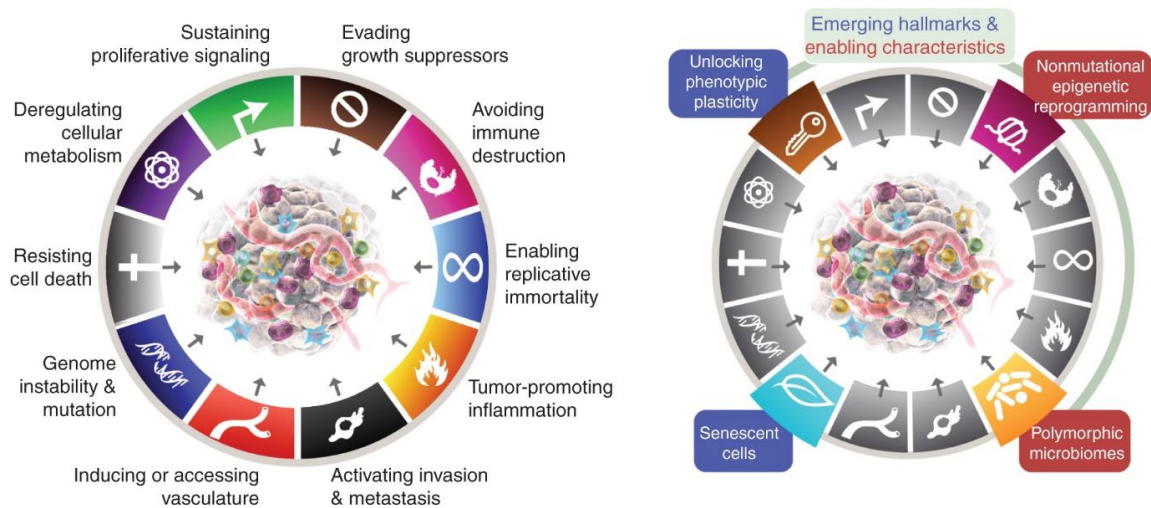
**Keywords:** Breast cancer, KLF8, SNPs, zinc-finger proteins, In-silico

# CHAPTER 1: INTRODUCTION

## 1.1 Cancer Overview

Cancer is a disease in which certain cells of the body undergo uncontrolled divisions leading to altered signaling and metabolism. The abnormal mass of cells formed in this way is termed as ‘tumor’. This mass establishes its own microenvironment. It is established through research that genetic mutations cause normal cells to transform. A lot of data is available to chalk out its root cause but the quest to find a cure continues (Upadhyay, 2021).

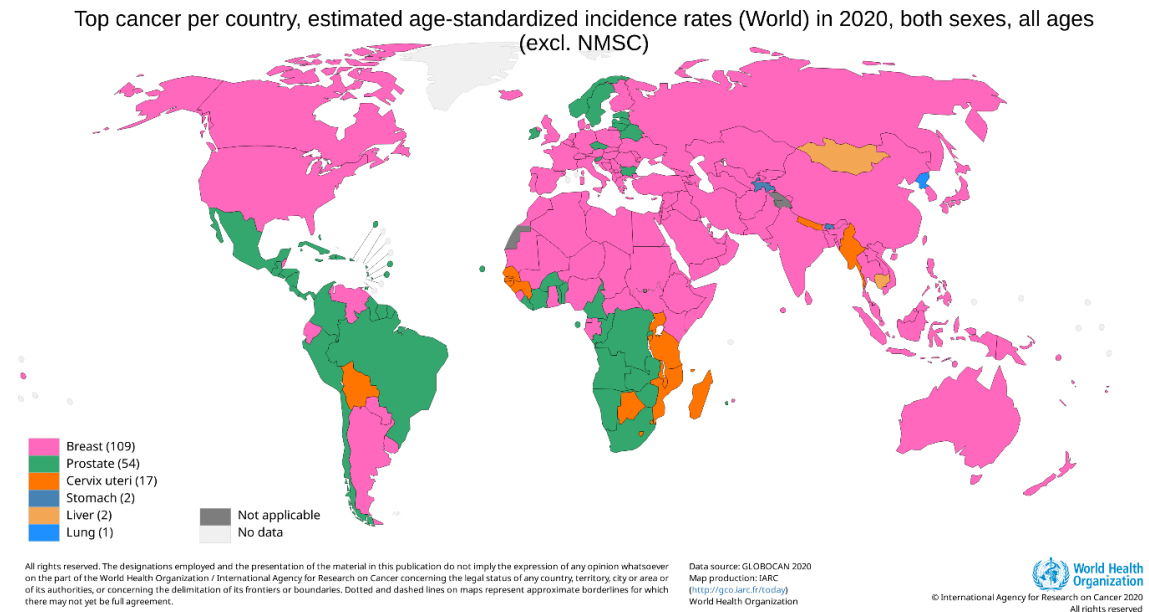
The functional capabilities of cells to grow rapidly are referred to as the hallmarks of cancer. To date, there are eight hallmarks of cancer. These include maintaining signaling pathways for proliferation, deceiving growth suppressors, opposition to apoptosis, unlimited replicative potential, establishment of a vasculature, ability to invade the nearby tissue and metastasis, metabolic reprogramming, and immune evasion (Hanahan et al., 2011).



**Figure 1.1** Hallmarks of cancer (Hanahan et al., 2011)

### 1.1.1 Global Cancer Prevalence

According to the International Agency for Research on Cancer, the disease has been diagnosed in 19.3 million people, and 10 million fell prey to it excluding non-melanoma skin cancer as of the year 2020. Breast cancer held its new place as the most commonly diagnosed cancer type replacing lung cancer and its associated death rate remained high (Sung et al., 2021).



**Figure 1.2 Global Cancer Prevalence (WHO)**

It is evident from the map that breast cancer is the most prevalent one. Our focus will be on this type of cancer.

## 1.2 Breast Cancer

Breast cancer is a multifactorial disease where the epithelial cells lining the ducts and lobules of the breast undergo changes regulated by genes and hormones. Its appearance in an individual involves risk factors that may be modifiable or non-modifiable. The modifiable factors include therapies, alcohol intake, being overweight, exposure to certain chemical and cancer-causing agents (carcinogens), excessive exposure to radiation, lack of physical activity, and insufficient intake of vitamins. The non-modifiable factors include sex, age, ethnicity, family history, previous therapies for cancer that involve radiation, the

hormonal status of the body, density of breast tissue, and genetic mutations (Łukasiewicz et al., 2021a). Of all the factors, we will focus on the genetic basis of breast cancer.

Molecular and histological evidence shows that breast cancer is of three types, one in which hormone receptors (estrogen and progesterone) are expressed, the second that expresses human epidermal receptor (HER2), and the triple-negative breast cancer (TNBC) (Barzaman et al., 2020).

### *1.2.1 Types of Breast Cancer*

According to Hormone Receptors (estrogen and Progesterone)

The hormone receptor-positive breast cancer expresses estrogen and progesterone receptors in approximately 70% of all breast cancers (Brufsky et al., 2018). Estrogen is an important hormone involved in reproduction, development, and growth and is steroidal in nature. They exert their effects through binding with their receptors which belong to the nuclear receptor transcription factor family. These receptors regulate the transition from G0 to the S phase in the breast epithelial cell cycle and so, defects of these receptors lead to breast cancer formation (Zhang et al., 2014). Progesterone is also a steroid hormone with numerous functions in the human body. It specifically functions in the reproductive system in maintaining pregnancy. It is produced by the adrenal cortex (Cable et al., 2023). Progesterone activity is associated with the activation of its receptors. It is a nuclear transcription factor. Its role has been known in estrogen receptor signaling in ER-positive breast cancer (Boland et al., 2020).

### HER2

HER2 is a membrane receptor that effects cell proliferation and survival. Its amplification leads to tumor development and invasion in breast cancer. It is amplified in 15-20% of breast carcinomas (Krishnamurti et al., 2014).

Triple-negative Breast Cancer (TNBC)



This type of breast cancer signifies a group of tumors where there is an absence of estrogen receptor, progesterone receptor, and over-expression of the human epidermal growth factor receptor 2 gene. 10-20% of human breast cancers are of this type (P. Kumar & Aggarwal, 2016).

### **1.3 Single Nucleotide Polymorphism (SNP)**

A SNP is a position in DNA where there is a change among individuals of a species (Nicholson et al., 2002). This polymorphism is most common, which occurs after every 1000 base pairs in the human genome (Karki et al., 2015). These are responsible for producing variations in the population (Cotton et al., 2018). These SNPs might be linked to phenotypes or disease. For this, genome-wide association studies (GWAS) are carried out. SNPs can alter gene function through affecting mRNA splicing, nucleocytoplasmic export, stability, and translation, thus changing the protein's activity (Robert et al., 2018). SNPs can also lead to disease susceptibility. This is indicated by the fact that a person possessing a certain SNP is more likely to have a disease when compared to a healthy person without that particular SNP. Some examples include, Alzheimer's disease, sickle-cell anemia, schizophrenia and various malignancies (Besseneyi et al., 2004).

#### *1.3.1 SNPs in Cancer*

As described earlier, SNPs can cause disease susceptibility. Likewise, these can cause cancer susceptibility if it falls in gene that regulates DNA mismatch repair, cell cycle regulation, metabolic activity, and immunity. These also function as biomarkers in various malignancies as diagnostic and therapeutic agents. If these lie in the promoter region, they modify the promoter activity, binding of transcription factor, DNA methylation and histone modification. If the SNP lies in exonal region, it promotes cancer susceptibility through suppressing gene transcription and translation. In intronic regions, these produce splice variants and either promote or prevent binding of long non-coding RNAs (lncRNAs), while in 5'UTR they affect translation and in 3'UTR, they impact binding of miRNAs (Deng et al., 2017).

### *1.3.2 SNPs in Breast Cancer*

The association between breast cancer and SNPs has been identified through genome-wide association studies. However, not much is known about the disease prognosis (He et al., 2019).

Here, we will study, the impact of single nucleotide polymorphism of a gene belonging to the family of Kruppel-like factors (KLFs). We have chosen to take the factor, KLF8. There is not enough evidence of a link between breast cancer and KLF8 polymorphism. It is yet to be discovered whether there is any genetic link between KLF8 SNPs and an increased risk of developing breast cancer. As a result, the primary goal of this research is to identify the possibly harmful SNPs found in KLF8 and link them with breast cancer. This will help us to explore more about KLF8, which will most likely lead to the development of a new possible therapeutic target as well as a prognostic marker for early detection of breast cancer.

## CHAPTER 2: LITERATURE REVIEW

### 2.1 Overview

The oldest depiction of cancer was of bone cancer from ancient Egypt. The term ‘cancer’ originated from a Greek word ‘karkinos’ coined by Hippocrates which depicted carcinoma tumors. It has been established from historical records that there was no treatment procedure for the disease. However, a surgical intervention was performed to clear away the surface tumors as is done in the present day (Sudhakar, 2009).

The process of gaining the ability to grow abnormally is termed ‘neoplasia’. There are two courses of neoplasms, either they are benign, meaning they reside in one place, or they become malignant which refers to invasion of the surrounding tissue and travel to distant organs (W.L., Kemp).

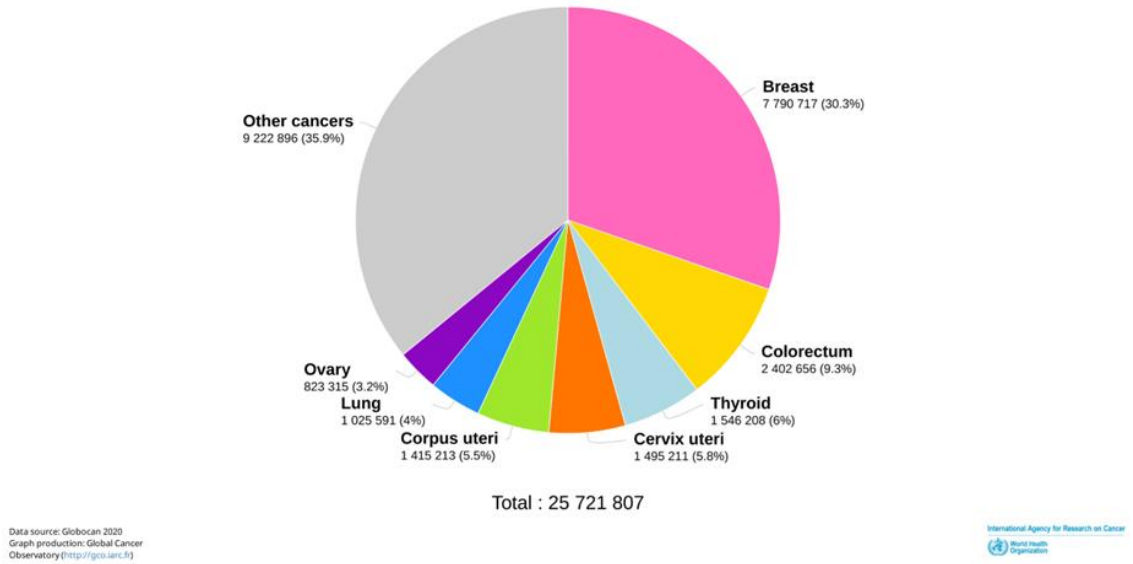
#### 2.1.1 Types of Cancer

There are various types of cancer, based on the organ where it originated. The disease can be classified into five broad classes. These include **carcinoma** (the cancer of body tissue), **sarcoma** (the cancer of connective tissues), **lymphoma** (the cancer of nodes or glands of the lymphatic system), **leukemia** (the cancer of bone marrow), and **myeloma** (the cancer of plasma cells of bone marrow) (Stanford Medicine).

### 2.2 Breast Cancer

According to the World Health Organization, breast cancer is the most prevalent type of carcinoma in females of all ages followed by colorectal cancer and other carcinomas (WHO).

Estimated number of prevalent cases (5-year) in 2020, World, females, all ages



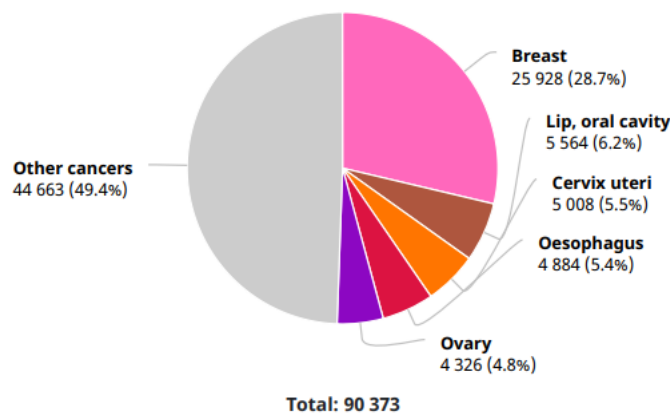
**Figure 2.1** Global Breast Cancer Prevalence (WHO)

Breast cancer is also has a high prevalence in Pakistan to the extent that 1 in every 9 females suffer from the disease at any age in their lives. In Asia, Pakistan holds the first place in breast cancer prevalence where there might be a lot of reasons including poor hygienic conditions and lack of awareness (Menhas, 2020.)

The International Agency of Research on Cancer reports that in the year 2018, the disease had been diagnosed in 34,066 females. Due to lack of proper healthcare facilities and issues related to delayed diagnosis has resulted in a high mortality rate (Khan et al., 2021).

According to the World Health Organization Globocan 2020 data, breast cancer in females remains on top of all new cancer incidences (WHO Globocan 2020).

## Number of new cases in 2020, females, all ages



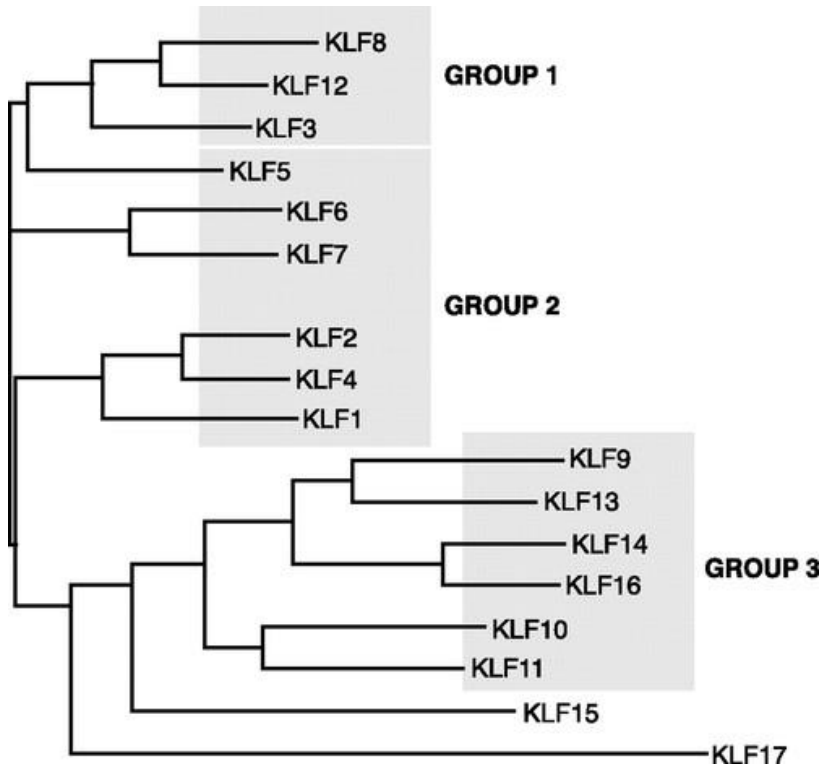
**Figure 2.2** Cancer Incidences in Pakistan (WHO Globocan 2020)

### 2.3 The KLF Gene Family

Gene expression is an important process through which the functioning of a gene is determined. The process primarily involves producing a product from a gene. The product is normally a protein. Gene expression involves the processes of transcription of DNA and translation of mRNA to produce proteins. The process of transcription happens within a nucleus of a cell in which an enzyme uses a coded section of a gene to produce mRNA which is further used by translation process outside of nucleus. During the transcription process the RNA polymerase with the help of proteins called transcription factors binds to a specific section within a gene to produce mRNA. The understanding of gene expression allows us to determine the function of a gene and its purpose through which we can develop a better understanding of the biological processes (Nature Education). Transcription factors that are used to generate mRNA are very important as they act as regulators or controlling factors in the gene expression process. Transcription factors also act as initiators in the gene expression process. The regulation or control of the gene expression process is an important as well as a complex part that ensures the creation of correct proteins (Wakim et al., 2012).

There are different types of transcriptional factors. Krüppel-like factors (KLFs) are transcription factors that bind to DNA and modulate cellular mechanisms and metabolism.

These are zinc finger protein structure-like transcriptional factors that act as regulators in the gene expression process. KLF can enhance as well as repress gene expression. There are 18 types of KLFs which are categorized into 3 groups. This classification is based on their role in regulating the process of transcription and their structural characteristics. Group 1 comprises KLF3, KLF8, and KLF12, Group 2 comprise KLF1, KLF2, KLF4, KLF5, KLF6, and KLF7, and Group 3 comprises KLF9, KLF10, KLF11, KLF13, KLF14, and KLF16. KLF15 and KLF17 have not been included in these groups due to poor knowledge about their association with other factors. Each KLF group has some functional characteristics which are categorized into specific groups. KLFs groups 1 and 3 primarily acts as transcriptional repressors whereas KLFs group 2 act as transcriptional activators (Pollak et al., 2018).



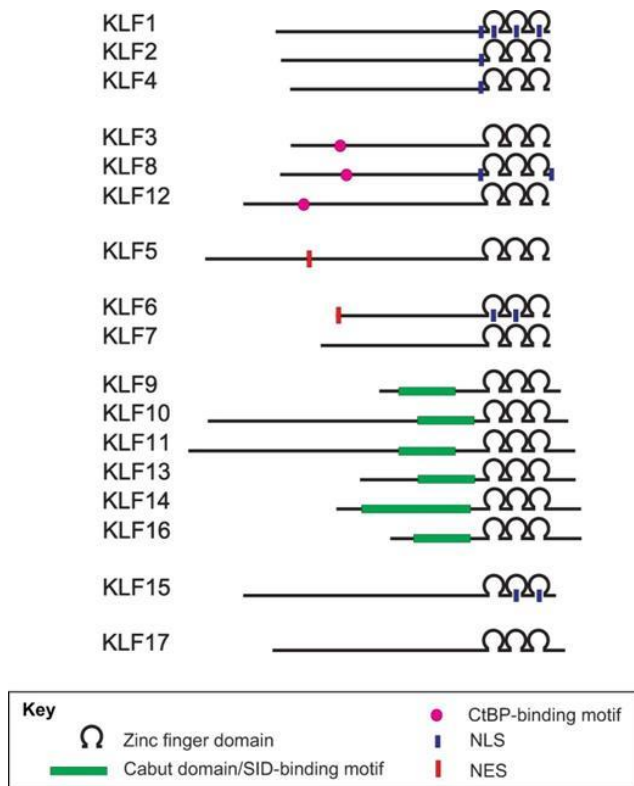
**Figure 2.1:** Phylogenetic Tree of KLF Family (McConnell et al, 2010)

The KLF family members work with other transcriptional regulators and play a vast role in metabolism in approximately every organ of the human body. They work to exert

physiologic effects where required. These are involved in processes such as intake of nutrients, catabolism of amino acids, lipid partitioning, glucose content fluctuation, and energy production through oxidation (Hsieh et al., 2019). These are also involved in important cellular functions including cell proliferation, differentiation, and apoptosis (Bialkowska et al., 2017).

The Kruppel-like factors carry three highly conserved C2H2 zinc finger domains at carboxy-terminals. These facilitate the activation or halting of transcription by interacting with GC-rich consensus sequences including 5'-CACCC-3' sequences of the DNA (Lomberk & Urrutia, 2005). On the other hand, the amino terminus of these factors controls the specific protein-protein and protein-DNA interactions, contrary to the carboxy terminus (Suzuki et al., 2005)

KLFs also share homology with another transcription factor called SP1. This was one of the initially identified and characterized transcription factor which binds to the GC-rich regions of the DNA through three C2H2-type zinc fingers. Due to the structural similarity of KLFs with SP1, these KLFs are categorized among the SP1/KLF family (McConnell et al., 2010).



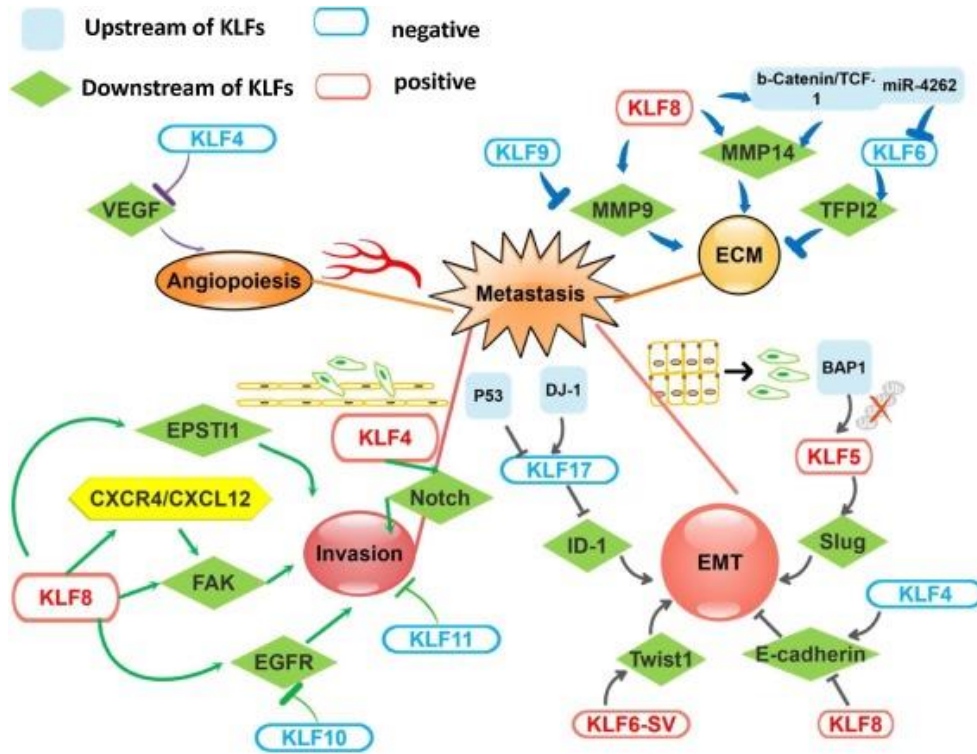
**Figure 2.2:** Domain structure of KLFs in humans (Bialkowska et al., 2017)

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**Figure 2.3:** The Role of KLFs in breast cancer metastasis (Zhang et al., 2020)

## 2.4 KLF8

### 2.4.1 Structure

KLF8 belongs to the family of KLF (Krüppel-like Factors) which act as transcription factors in the regulation of gene expression process. The main purpose of KLF is to regulate or control gene expression by binding to DNA sequences by interacting with promoter regions of genes. KLF8 belongs to group 1 of the KLF family which contains KLF3 and

KLF12 factors too (Satadru et al., 2012.)). In humans, the KLF gene family member, the KLF8 is located on the X-chromosome. It comprises of 359 amino acid residues and was isolated from K562 leukemia cell line for the first time. A transcription factor contains three main domains, a DNA binding domain, which in the case of KLF8 is highly conserved, a transcription regulatory domain, and a nuclear localization signal (Vliet, Turner, & Crossley, 2000).

#### The DNA-binding Domain of KLF8

This factor contains a DNA binding part which allows it to detect and bind to promoter region of a gene. Its structure allows it to bind to CACCC, a specific pattern in promoter region of a gene (Funnell et al., 2012).

#### The Transcription Regulatory Domain

Transcription activation part allows KLF8 to regulate or control the transcription of genes. These parts allow transcriptive factors to interact with other factors or proteins to enhance the process of gene transcription. It has been observed that KLF8 acts as a transcriptive activator as well as repressor in certain scenarios (Aiqin Sun & Jie Hao, 2019.).

#### The Nuclear Localization Signal

Nuclear Localization Signal (NLS) is the process of transporting certain proteins to the nucleus of a cell. This process is very important as regulation of proteins in and out of the nucleus is crucial for gene expression (Cokol, Nair, & Rost, 2000). Since KLF8 is a transcriptive factor and it needs to be transported to the nucleus of a cell in order to regulate gene expression, it contains a nuclear localization signal that helps it to enter into the nucleus. nuclear localization signals of proteins like KLF8 are detected by importins (transport proteins) when formed in the cytoplasm and transported to the nucleus through nucleus pores (Mehta et al., 2009).

#### *2.4.2 Functions of KLF8*

KLF8 plays various roles based on its interaction with different genes. KLF8 plays a critical role in different cell processes for example cell proliferation, cell differentiation, and cell

transformation. Cell proliferation is the process by which cells are divided and their numbers are increased. The cell proliferation process plays an important role in tissue development and maintenance in the human body. KLF8 plays its role in cell differentiation in which cells are assigned specialized roles or functions and switching from one cell type to another in order to perform various tasks. Many transcription factors play an important role in the cell differentiation process helping a single cell to develop in various tissues and organs. Along with cell proliferation and differentiation, KLF8 also plays its role in the cell transformation process which involves the change in characteristics of a cell (Phenotype) as a result of inserting a new genetic material in a cell (M.-D. Wang et al., 2020).

A dysregulation of KLF8 leads to diseases. For instance, KLF8 has been observed to be involved in Alzheimer's Disease (AD) which is related to the brain affecting brain memory and brain cell structure. KLF8 is expressed in the brain as well and plays its role in neural processes happening in the brain. Several investigations are underway in order to understand the involvement of KLF8 in neural disorders inside the brain (Yi et al., 2014)

#### *2.4.3 Oncogenic Transformation and KLF8*

It is the process by which healthy cells are transformed or undergo changes which lead to abnormal or unregulated cell proliferation which in turn lead to cancer cells development. This process involves oncogenes which when activated amplifies the cell proliferation process. KLF8 plays its role in oncogenic transformation as it is an important transcription factor involving cell proliferation as described earlier. This leads to tumor progression in the human body (Wang et al., 2007). KLF8's role in the EMT process in tumor progression is the main reason for its involvement in different types of human cancers. Studies have shown KLF8 involvement in different types of cancers in the human body such as breast, gastric, liver etc. KLF8 involvement has been observed in growth of human lung cancer as well. In recent experiments, KLF8 overexpression was involved in growth and proliferation of lung cancer cells however on the contrary the knockdown of KLF8 decreased the lung cancer cell growth (Ma et al., 2019).

### **2.5 KLF8 in Breast Cancer**

KLF8 is a transcription factor which is overexpressed in multiple cancer types including the breast cancer. It plays a significant part in epithelial to mesenchymal transition, proliferation, and DNA damage repair (Tianshu Li et al., 2015a). It has also shown to play a role in breast cancer metastasis (Mukherjee et al., 2016). Another role of KLF8 has been identified to regulate other genes and microRNAs in promoting EMT and stem-cell-like traits to promote invasion (Xianhui et al., 2013)

### In EMT

Epithelial to Mesenchymal Transition (EMT) is a process which allows cells to change their characteristics from epithelial (tightly connected cells lying on the surface of organs) to mesenchymal (mobile cells). This process allows the transition of cells from a structured state to a more mobile state. EMT is involved in the beneficial process such as healing wounds, however, it can also play a role in spreading cancer tumor cells on the contrary side (Kalluri et al., 2009).

It has been observed that KLF8 plays its role as an inducer in the EMT process. KLF8 unregulated expression could be a potential reason for spreading cancer cells in the human body. KLF8 facilitates cell migration and invasion which is a core functionality in the EMT process. KLF8 also plays its role in tumor progression which enables cancer cells to acquire mobility in order to spread to different parts to the body. It does so by directly binding and repressing the promoter of E-cadherin independent of E boxes and SNAIL expression. Therefore, more expression of KLF8, lower is the expression of E-cadherin in invasive human breast carcinoma (Xianhui Wang et al., 2009).

### In Metastasis

Metastasis is the mobilization of tumorigenic cells from their primary site to the secondary one by rupturing the extracellular matrix. This rupturing of the extracellular matrix for invasion of surrounding cells is brought about by the matrix metalloproteases (MMPs).

### MMP9

KLF8 regulates MMPs in breast cancer. This was studied by employing two cell lines, one to check the impact of overexpression of KLF8 on MCF-10A and the second, MDA-MB-231 to check the impact of knocking down the KLF8. High expression of KLF8 enhanced MMP9 expression. Opposite to this, KLF8 knockdown led to reduced expression of MMP9. This signified the role of KLF8 and MMP9 signaling in human breast cancer invasion and metastasis (X Wang et al., 2011).

### MMP14

Another role of KLF8 was identified in MMP14 signaling in promotion of invasiveness and metastasis of human breast cancer. At transcriptional level, the KLF8 promotes MMP14 expression and its knockdown showed the opposite effect. This happens through two mechanisms, either by directly affecting the promoter or by indirectly promoting the nuclear translocation of  $\beta$ -catenin, expressing T-cell factor-1 (TCF1) and successively activating the promoter by  $\beta$ -catenin/TCF1 complex. Moreover, by down-regulating focal adhesion kinase (FAK) or its knockdown led to the deduction that the cell surface presentation of active MMP14 is dependent upon FAK expression and activity downstream of KLF8. This signifies the combined role of KLF8 and FAK to regulate the activity of MMP14 for breast cancer metastasis (H Lu et al., 2014).

### CXCR4

Another method adopted by KLF8 to induce metastasis is through induction of CXCR4. A cell line MDA-MB-231 was used to carry out experiments to check this effect. Treatment of cells in this cell line with a variant CXCL12, of CXCR4 ligand showed the formation of filopodium-like protrusions. Overexpression of the factor, KLF8 as well as CXCR4 led to aggressiveness of tumor through metastasis in immunocompromised mice. Overall, this pointed towards recurrence of breast cancer malignancy (Mukherjee et al., 2016).

## EPSTI1

The Epithelial stromal interaction 1 (EPSTI1) is a newly identified stromal fibroblast-induced gene. It has been shown to play a role in invasive breast cancer malignancy by being overexpressed. Two cell lines were used to check the effect of KLF8 expression on EPSTI1. One cell line was used to check the impact of overexpression of KLF8 on EPSTI1 in MCF-10A and the second cell line, MDA-MB-231 was used to check the impact of knocking down the KLF8. High expression of KLF8 enhanced EPSTI1 expression. Opposite to this, KLF8 knockdown led to its reduced expression. KLF8 attached to the promoter region of EPSTI1. EPSTI1 was explored to interact with the valosin containing protein (VCP), degrading the I $\kappa$ B $\alpha$  which ultimately activated of NF- $\kappa$ B in the nucleus. This however led to the conclusion that KLF8 signaling mechanism along with EPSTI1, VCP, and NF- $\kappa$ B is critical in invasion and metastasis of human breast carcinoma (T Li et al., 2014).

## In Drug-resistant Metastasis

KLF8 was observed to be over expressed by treatment of LM2 cells with chemotherapy drugs. This over-induction of KLF8 or CXCR4 is capable of causing increased metastasis to various organs and led to excessive secretion of VEGF, establishment of new vasculature, formation of filopodium-like protrusions, and colonization. Colonization is changing of micro-metastasis to macro-metastasis at distant sites. Such enhancement of KLF8 upon chemotherapy and the association of KLF8/CXCR4 indicated drug-resistant metastasis and can prove to be a new drug target (Hao et al., 2021).

## In DNA Repair

A novel role of KLF8 includes protecting the breast cancer cells' from dying and DNA from being damaged by harmful effects of doxorubicin. In this way, it repairs the DNA of breast cancer cells. As soon as the DNA of these cells gets damaged, KLF8 is phosphorylated by DNA-dependent protein kinase's catalytic subunit. It is then sumoylated

by SUMO E3 ligase protein inhibitors of activated STAT. the process of DNA repair then starts as the KLF8 is recruited to the site of DNA damage (Heng Lu et al., 2012).

### Stem cell-like Activity

Stem cells are the cells that have the capability to self-renew and to differentiate into mature cell types. These mature cells are specialized to perform specific functions. In case of tumors, the sub-population of stem cells are called the cancer stem cells (CSCs). These illustrate the properties of stem cells as well as cancer cells. The breast cancer stem cells (BCSCs), promote breast cancer invasion and metastasis. The tumor microenvironment thus established, plays a critical role in growth signaling and proliferation (He et al., 2021).

In the case of KLF8, it was observed to be overexpressed in triple-negative breast cancer cells, which further enhanced the population of CSCs. On the contrary, downregulating the expression of kLF8 reduced the CSC population. Recently, it has been observed that a nutrient sensor O-GlcNAc transferase (OGT) and O-GlcNAcylation for increased in the population of cancer stem cells. This promoted stemness and tumorigenesis of breast cancer cells, both in-vivo and in-vitro. This increment was interlinked with increased expression of KLF8 suggesting its potential role in rendering CSC traits (Le Minh et al., 2023).

### Invasion through targeting MicroRNAs

MicroRNAs (miRs) are short non-coding regulatory RNAs. They target the 3'UTR (untranslated region) of mRNA. They are tightly regulated in normal cell. However, their role as critical regulators of tumor progression and as cancer stem cells has been explored in the recent past (Mo, 2012).

### miRNA-146a

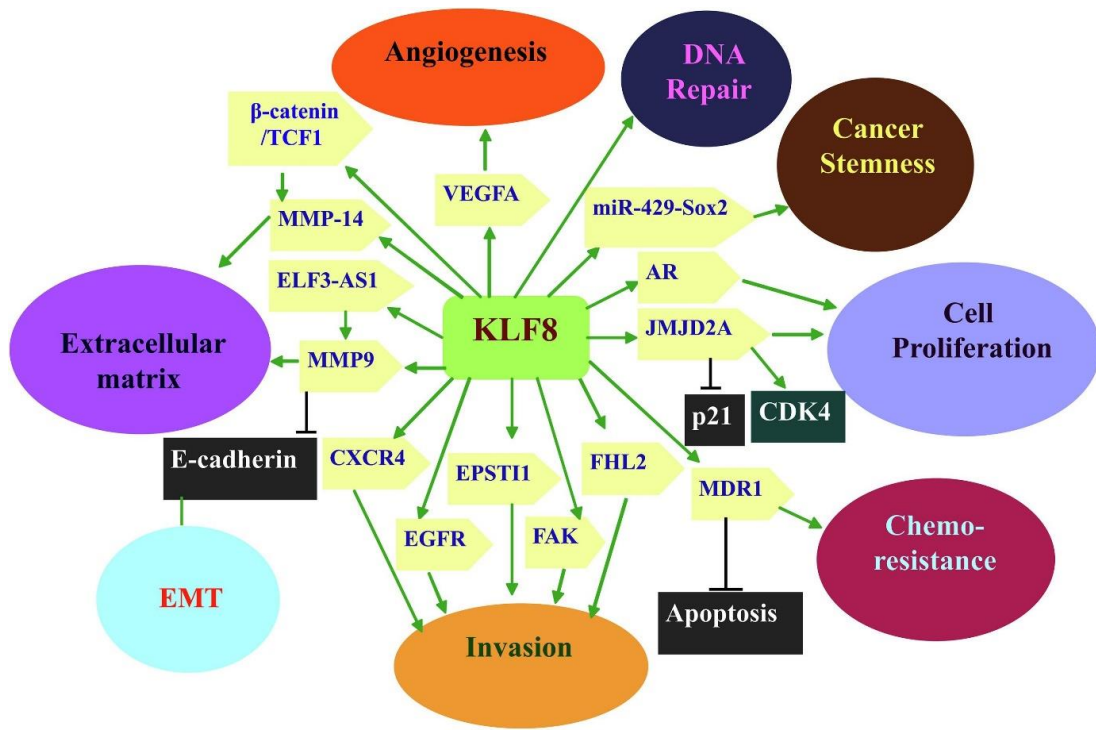
The KLF8 has emerged as a promoter of breast cancer through rendering the cancer stem cell ability to the mammary stem cells. This ability of stem cells can be induced during epithelial to mesenchymal transition (EMT). The miR-146a has been explored

as a target of KLF8 and TGF- $\beta$  which are responsible for inducing EMT. It was observed that the MCF-10A cells acquired the stem cell properties when KLF8 was over-expressed or treated with TGF- $\beta$ . This led to overexpression of miR-146a in invasive breast cancer cell lines. On the other hand, if the cells had been treated with miR-146a inhibitors, the MCF-10A cells were incapable of having the stem cell traits. The reason was identified as miR-146a targeting the 3'UTR of Notch signaling inhibitor NUMB for translational inhibition. Hence, the miR-146a potentially activated Notch signaling by being stimulated by KLF8 (Wang et al., 2013).

#### miRNA-141

The KLF8 is found to target the miR-141. It represses the promoter of miR-141 which targets the 3'-UTR of Epidermal Growth Factor Receptor (EGFR) to stop its translation. EGFR overexpression had been observed in invasive triple-negative breast cancer. Upon upregulating the expression of KLF8 in non-tumorigenic MCF-10A cells, led to overexpression of EGFR whereas knocking it down in another cell line, MDA-MB-231 downregulated it. This shows the impact of KLF8 on expression of miR-141/EGFR signaling pathway in breast cancer malignancy (Tianshu Li et al., 2015).





**Figure 2.1:** Role of KLF8 in Cancer (Kumar et al. 2021)

## CHAPTER 3: MATERIALS AND METHODS

### **In-Silico Analysis**

#### *3.1.1 Data Retrieval*

The SNP data for KLF8 was retrieved from the ENSEMBL database under the transcript ID KLF-204-ENST00000468660.6. The information included variant ID, base and residue alterations, genome coordinates and amino acid coordinates. The genome assembly number was GRCh38.p14 (GCA\_000001405.29). the transcript had 6 exons is annotated with 24 domains and features, is associated with 17623 variant alleles and maps to 668 oligo probes. This gene has 8 splice variants, orthologues and paralogues.

Further data retrieval was done from the Genome Aggregation Database (gnomAD) and, the Catalogue of Somatic Mutations In Cancer (COSMIC) databases. The KLF8 gene variants were selected excluding the in-frame variants. All the variants were then categorized into two categories. These were named coding variants and regulatory variants. The synonymous, non-synonymous, truncated and frameshift variants were included in the coding variants category while splice site and untranslated region variants (UTRs) were included in regulatory variants. Then the selection criteria was narrowed down to missense variants that give altered protein products. These selected variants were analysed further.

#### *3.1.2 Identification of unique missense variants*

From the three databases, ENSEMBL, gnomAD and COSMIC, the number of selected KLF8 variants was summed up. The total number of variants from each database was represented through a graph. These total variants were then compared and the common ones were merged to avoid repetition. All other unique variants were assigned a separate number even if they resided on the same location.

#### *3.1.3 Evaluation of deleterious variants*

The missense variants were then filtered using five software tools. These tools include SIFT, PolyPhen, MetaLR, REVEL and Mutation Assessor for determining pathogenicity. These tools give scores to determine the pathogenicity of the variants.

SIFT score predicts whether an amino acid substitution might cause changes in function of the resulting protein. This score ranges from 0.0 to 1.0 where 0.0 to 0.05 score range is predicted deleterious, 0 being the highest value and 0.05 to 1.0 is predicted to be tolerated, 1 being highly tolerated. PolyPhen also applies a score range from 0.0 to 1.0 where 0 indicates benign and 1 indicates deleterious. MetaLR divides the variants into either a 'damaging' or a 'tolerated' category. The scores range from 0 to 1 where a higher value indicates higher chances of being deleterious. REVEL applies a similar score range from 0 to 1 where scores above 0.5 indicate that the variant is 'likely disease-causing' and scores below 0.5 indicate they are 'likely benign'.

After filtering the variants, these were rendered for further analysis. This led to considering two pathogenic variants of the gene KLF8 that were subjected to wet lab analysis. These variants had rs IDs 'rs1196572444' and 'rs868781835'.

#### *3.1.4 Analysis of stability of filtered variants*

For the analysis of stability of filtered variants, different software tools were used. These tools determined the effect of missense mutation on the resulting protein. These softwares include Mupro, Mutpred2 and DynaMut.

For retrieving data from I-Mutant, protein sequence was fed in FASTA format as well as the amino acids and their coordinates. The results were obtained in the form of DDG values (Kcal/mole). A DDG value less than 0 indicates destabilization in protein whereas greater than 0 indicates increasing protein stability.

MUpro is a software that provides a prediction on changes in protein stability due to a variation in a single amino acid. It gives a change in energy score (DDG) which ranges from -1 to 1, where a score less than 0 indicates decreasing stability while greater than 0 indicates increasing stability due to presence of an amino acid variant.

DynaMut tool is responsible for evaluation of the impact of amino acid variants on stability and molecular dynamics of resulting protein. Also, the difference in vibrational entropy (DDS) between mutant and wild type proteins is predicted through this tool.

Another software tool, InterPro was used to gain information on the domains of KLF8, its active site, substrate binding site and the amino acids found in each region. For this, the data was entered in FASTA format.

### *3.1.5 Structural and Functional Analysis of Pathogenic Variants*

The structural and functional analysis of pathogenic variants was done using software tools including MutPred2, DynaMut and HOPE. These tools helped in estimating the effect of these pathogenic variants on structure and function of resultant proteins.

MutPred2 is a software tool that predicts the pathogenic influence of amino acid variants on resulting proteins and their molecular activities through calculating the likelihood of pathogenicity in the form of a score. This probability scores greater than 0.5 indicates pathogenicity but this has a chance of being incorrect. For this, a set criterion has been devised according to which, a score greater than 0.68 indicates 10% likelihood of false positive results. On the other hand, a score greater than 0.80 indicates 5% chance of false positive results.

The DynaMut software tool determines protein flexibility and rigidity. However, this was used to predict the alteration in protein structure.

The HOPE software describes the changed protein domain and charge difference between the wild-type and variant under consideration. It also determines the affect on molecular interaction with surrounding residues. The information regarding change in protein size due to amino acid differences is also provided by HOPE.

### *3.1.6 In-silico mutagenesis*

In-silico mutagenesis was carried out in PyMol, molecular visualization system. The structure of wild KLF8 was visualized and an amino acid in the wild type was changed to

create the mutated variant. The changed amino acid was given a different color from the original one. This was done by selecting the ‘Wizard’ option > Mutagenesis > Protein and then selecting the residue and converting it to the amino acid as in the variant. The structure was then saved in ‘pdb’ format. Molecular dynamics simulations were run on wild-type and mutant KLF8 structures.

### *3.1.7 Molecular Dynamics (MD) Simulations*

Molecular dynamics simulations were conducted using GROMACS on a super computer. An open source, terminal emulator, PuTTY was used to give commands to the remote system for performing functions such as cleaning the structure of protein, topology, and labeling of coordinates. Different algorithms performed energy minimization and temperature and pressure equilibrium to carry out 20ns MD. An open source file manager, WinSCP was utilized to transfer files between the local and remote server. When the MD steps were completed in PuTTY, the output data analysis was done to calculate RMSD, RMSF, SASA, Radius of gyration and hydrogen bonds number. The results were depicted graphically.

## **3.2 Primer Designing**

Primers were designed to carry out Tetra ARMS PCR. For this, the bioinformatics tool, Primer 1 was used. A total of four, two inner and two outer primers were designed to amplify the gene region containing the variant under study. These primers were then validated through UCSC genome browser in-silico PCR. Here, primer length, temperature, allele difference and SNP position were chosen.

## **3.3 Experimental analysis**

### *3.3.1 Sample Collection*

For blood sample collection, an approval was taken from the Institutional Review Board of Atta ur Rahman School of Applied Biosciences. A total of 100 samples each, breast cancer patients and healthy controls were collected. This was done following the

recommendations of the ethical review board. Breast cancer patients without comorbidities such as diabetes and hypertension and pregnant and lactating women were not included in the study. The volunteering patients' consent was taken. Data regarding their medical history, age, tumor grade, cancer type, history of smoking and alcohol consumption and treatment status was also collected. For blood collection, a tourniquet was tied 3cm above the visible vein in the arm or hand and the area was sanitized. Then, using a sterile syringe, 3-5ml of blood was drawn and added to a 5ml EDTA tube.

### 3.3.2 DNA extraction

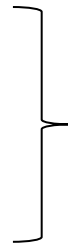
The organic method was used for DNA extraction from blood samples of experimental and control groups. This method employs phenol, chloroform and isoamyl alcohol. For this, four buffers were prepared, named solution A, B, C and D.

#### **Solution A:**

0.32M sucrose (54.72g),

10mM Tris-base pH=7.5 (0.602g),

5mM magnesium chloride (0.238g)



dissolved in autoclaved distilled water to make 500ml of solution

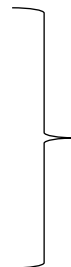
Following this dissolution and autoclaving, 5 mL of 1% V/V Triton X-100 was added and stored at 40° Celsius. This solution breaks cellular membranes to expose the DNA.

#### **Solution B:**

10mM Tris-base pH=7.5 (0.6057g),

400mM sodium chloride (11.685g),

2mM ethylene diamine tetra acetic acid (0.58g)



dissolved in distilled autoclaved water with volume raised to 500ml and pH adjusted to 7

**Solution C** is composed of phenol kept in a dark, properly covered reagent container in aluminum foil, in a cold environment

**Solution D** is composed of 24 ml of chloroform and 1 ml of isoamyl alcohol combined to make 50 mL of solution, diluted to 500mL distilled water.

**Procedure:** The process of extracting the DNA takes two days. On the first day, 750µl blood was taken in an Eppendorf tube with an equal volume of Solution A and incubated at room temperature for 10 minutes. It was then centrifuged for 10 minutes at 13000 rpm. Then, the supernatant was discarded. The pellet was resuspended in 500µl Solution A, centrifuged, and the supernatant was discarded. To the supernatant, 400µl of Solution B was added and centrifuged at 13000 rpm for 10 minutes. This caused the proteins to separate and remove from the pellet while precipitating the DNA in the pellet. Then, the pellet was dissolved in 400µl Solution B, 12µl SDS and 5µl proteinase K. The tube was then kept for incubation at 37 °C for a night.

Next day, the incubated tube was taken, 250 µl of Solution C and the same quantity of Solution D were added and centrifuged at 13000 rpm for 10 minutes. This step was done to create two layers. Solution C separated the aqueous and organic phases in the solution whereas solution D stabilized protein coagulation and aided DNA purification by reducing foam formation. Of these two layers, the upper layer contained DNA and was carefully picked and transferred to a new Eppendorf tube. To this tube, 500µl of isopropanol and 55µl sodium acetate was added and centrifuged at 13000 rpm for 10 minutes. The pellet was suspended in 200µl centrifuged at 13000rpm for 8 minutes. The supernatant was discarded and the tube was inverted for drying. When, dried, 25-30 µl PCR water was added to store the extracted DNA.

### *3.3.3 Tetra ARMS PCR*

This procedure was performed to determine the genotype of the samples obtained. This was done using two types of primers, the inner and the outer primers designed according to the variants created by in-silico mutagenesis. A reaction mixture in the quantity of 20µl was prepared. This mixture contained 1 µl of each primer i.e., a total of 4 µl, 2 µl of the

extracted DNA, 2  $\mu$ l nuclease-free water and 12  $\mu$ l of SolisFAST master mix. The master mix contained PCR buffer comprising of magnesium chloride, dNTPs and additives that maximize PCR efficiency and DNA polymerase.

The tubes comprising the reaction mixture were centrifuged for a few seconds and placed in the thermal cycler. Primer optimization was carried out at various chemical concentrations and melting temperatures in multiple PCRs. For determining optimum temperature ranges, the gradient PCR technique was applied. The PCR stretched over 30 cycles and three key phases, denaturation, annealing, and extension.

In the first phase, the template DNA was denatured from a double-stranded structure to a single-stranded structure when rendered to 95°C for 5 minutes. In the next phase, the daughter DNA strands were denatured. This phase stretched to 30 cycles, 30 seconds each at 94°C. The next phase was primer annealing at 58.5°C for 30 seconds. Following this phase, was the extension phase carried out at 72°C for 30 seconds. The final extension period was 7 minutes conducted at 72 °C. After this, the PCR cycle was paused at 4 °C to cool the PCR product and temporary storage.

#### *3.3.4 Agarose Gel Electrophoresis*

Agarose gel electrophoresis was performed to visualize the DNA extracted from the blood samples collected and the PCR products. Agarose gel was made in the concentration of 2% W/V. for this, 2gm of agarose powder was mixed with 1X TAE buffer and the volume was raised to 100ml and heated to ensure complete dissolution. Then, this was slightly cooled and 5-7  $\mu$ l of ethidium bromide was added for ease of visualization under UV light. The prepared gel was then carefully poured in gel casting tray avoiding bubble formation. A comb was applied to create wells and then the gel was kept to set completely. An electrophoresis tank was filled with 1X TAE buffer. The set gel was taken, comb was removed, and the gel was submerged in the buffer-filled tank. In one of the wells, ladder DNA was injected as a reference in the quantity of 5 $\mu$ l, while in other wells, 8-10  $\mu$ l of the PCR product was injected. The electrophoresis machine was set at 80V and 500A and the gel was run for 40 minutes. Then the gel was visualized under UV transilluminator.



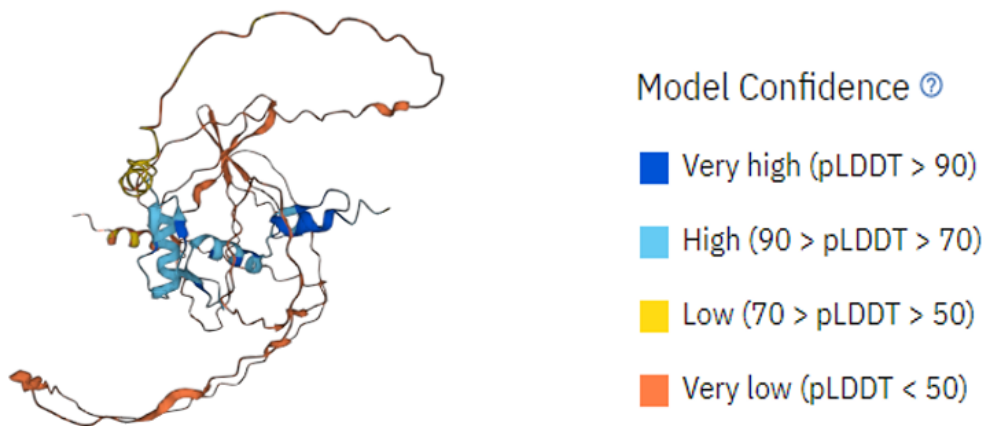
For visualizing the extracted DNA, the concentration of gel was 1.5% W/V, 8-10  $\mu$ l DNA sample was combined with 1  $\mu$ l loading dye and then injected in the wells and visualized under UV light in the similar manner as done for PCR products.

## CHAPTER 4: RESULTS

### 4.1 In-silico Analysis

#### 4.1.1 Structure Prediction of KLF8 Protein

KLF8 protein's structure was predicted by AlphaFold, which used per residue model confidence score (pDDLT) ranging from 0 to 100. Based on this model, a 3D structure was obtained. The interpro analysis provided information on domains of KLF8 protein. There is one N domain from amino acid number 48-216 while three zinc-finger C2H2 domains from 274-303, 304-333 and 334-359. According to AlphaFold, the N domain showed a low confidence score while the other three C2H2 domains showed high and very high confidence score.



**Figure 4.1:** KLF8 protein's 3D predicted structure, C-terminal region domains showed high confidence score (70-90) of predicted structure.

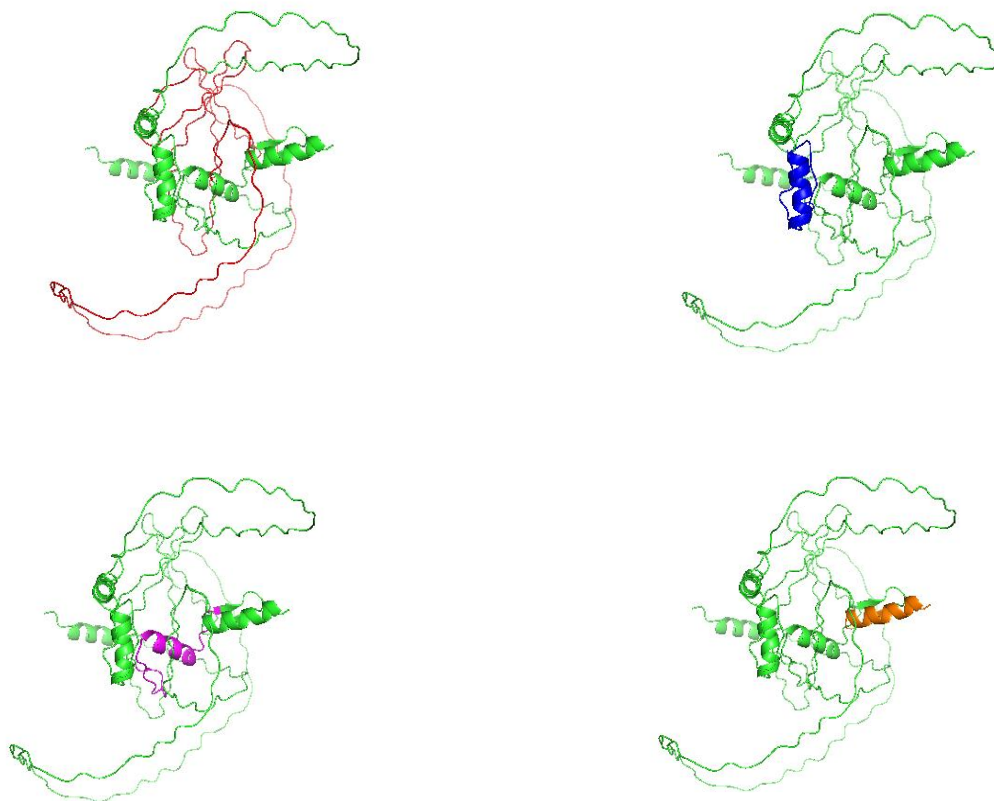
▼ Representative Domains

KLF8\_N

ZINC\_FIN...

ZINC\_FIN...

ZINC\_F...



**Figure 4.2:** Representative domains shown in Interpro

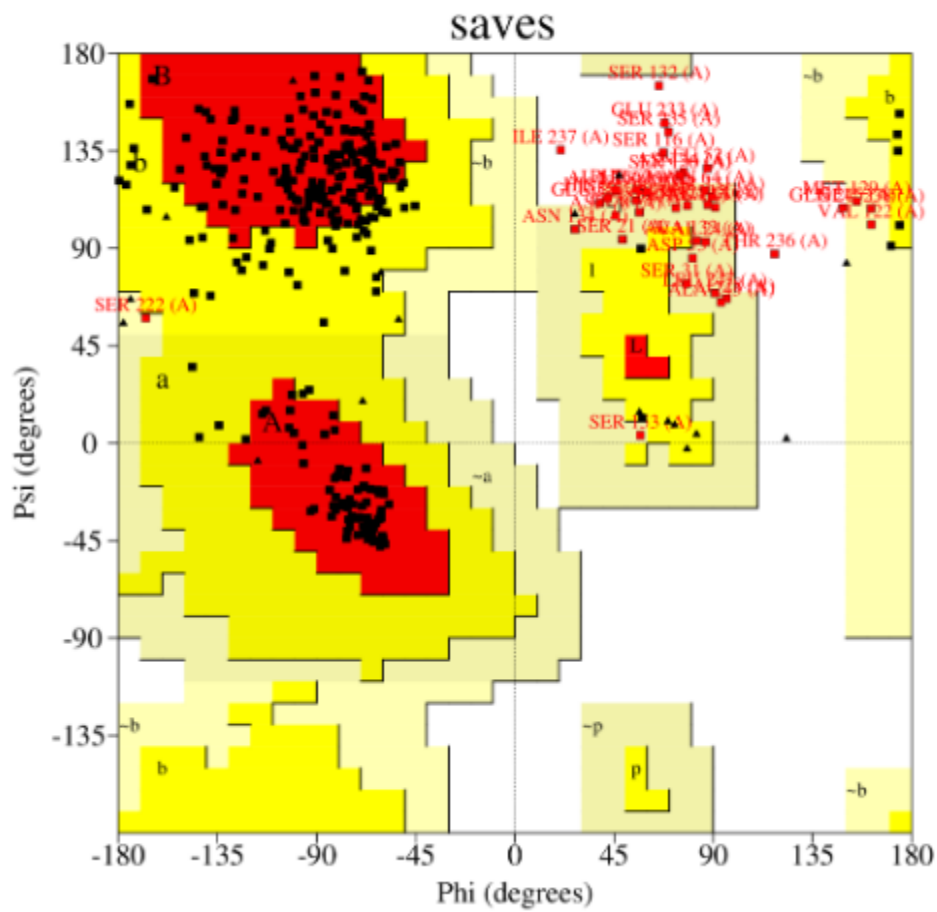
The structure was validated in SAVES and PROCHECK analysis revealed that 70.3 percent

of amino acid residues had phi-psi angles in the most preferred regions, 17.3 percent in

additionally allowed regions, 8.8 percent in generously allowed regions, and 3.6 percent

residue had phi-psi angles in the disallowed regions of the Ramachandran plot

# Ramachandran Plot



**Figure 4.3** Ramachandran Plot showing the measurements of angles in KLF8

## 4.1.2 Subcellular Localization and Phylogenetic Tree

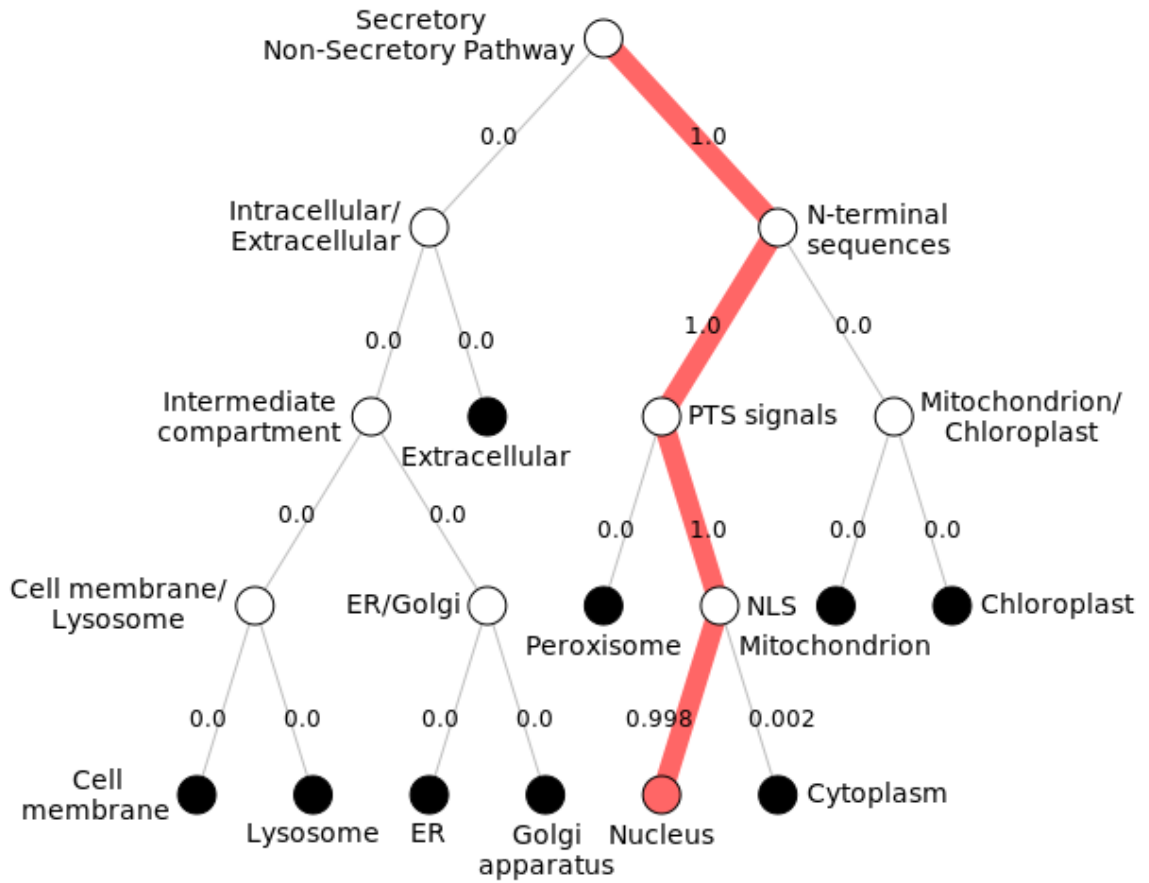
Deeploc 1.0 predicted the subcellular localization of the KLF8. Figure 4.3 and Table 4.1

below depict the path of KLF8 localization. The path taken by the protein to localize to its

compartment within the cell is represented by the red line. The score represents the

probability/likelihood of the event. As a result of the score, it's assumed that KLF8 is

found in the nucleus.



**Figure 4.4:** KLF8 protein's localization route and the likelihood score. The path of localization is shown in red

**Table 4.1** KLF8 Localization and its Likelihood

<b>Localization</b>	<b>Likelihood</b>
Cytoplasm	0.0041
Nucleus	0.9958
Cell membrane	0
Lysosome/Vacuole	0
Golgi Apparatus	0
Peroxisome	0
Extracellular space	0
Endoplasmic Reticulum	0
Mitochondrion	0

Clustal Omega software was used to find the evolutionary relationship between KLF family members. Figure 4.4 depicts the evolutionary profile of KLFs, which reveals that all members of KLF8's descended from a single ancestor protein. The score measures the substitution per site, which indicates how much a specific protein has evolved through time in comparison to its family members. The conserved regions of KLF8 were visualized through ConSurf as shown in the figure below.

```

1      11      21      31      41
MVDM4DK1LI3NN4  LEVQ2LN2SE6GG2  SM5QV4FK3QV4TA3  SV6RNRD3PP2EI2  EY3RS3NMT2S3PT2
e e e e e e b b e e e  b e b e b e e e e e e  b b e b b e e b e e e  e b e b e e e e e e e  e e e e e e e e e e e
  f f                    f                                 f

51     61     71     81     91
L2LDANP2ME1NP2  AL2FN2DI2KI2EP2  PE2EEL2AS2DF2S2  LP2QV2EP2VD2LS2  FHK2PKA1PL1Q1Q1P1
b e e e e e e e e e e  e b b e b e e e e e e  e e e e e e e e e e e  e e e e e e e e e e e  e e e e e e e e e e e
                                      f                   f f f f f

101    111    121    131    141
AS2ML1QAP2IR1P1  PK1PQ2SS2P1Q1TL1  VV2ST2ST2S2D2MS2  TSAN2IPT1VL1T1  PGS2VLT2SS2QS2
e e e e e e e e e e e  e e e e e e e e e e e  e b e b e e e e e e e  e e e e e e e b b b b  e b e b b e e e e e e

151    161    171    181    191
TGS2QQ1IL1H1VI1  HT2I1PS2VS2L1PN1  KMG2GL2KT2IP1V1  VV2QSL1PM1V1YT1  TLPAD2GG2PAA2
e e e e e b b b b b b  b b b e b e e e e e e  e e e e b e e b e e e  e b e e b e b b b b e  b b e e e e e e e e e
                                      f

201    211    221    231    241
ITV1PL1IG2GD2G2  KNAG2SV2KV2DP2  TSM2SP2LE2IP2S2  DSEEST2IES2G2  SSALQ2SLQ2GL2
b e b b b b e e e e e  e e e e e b e e e e e  e e e e e e e e e e e  e e e e e e e e e e e  e e e b e e b e e e e

251    261    271    281    291
QE2EPA2AMA2QM2  QG2ES2SL2DL2KR2  RRI2H2QC2DF2AG2  CS2KV2Y2TK2SS2H2  LKA2H2RR2I2HT2G2
e e e e e e e e e e e  e e e e e e e e e e e  e e b e e b e e e e e  b e e e e e e e e e e  b e e e e e e e e e e
                                      f                   f f f s s f f         s f f f f f f f f f f s f f f f f f f f

301    311    321    331    341
EK2PK2K2CT2WD2G2  CS2WK2F2ARS2DE2  LTR2HF2RK2HT2G2  IK2PF2R2CT2DC2N2  RS2FS2RS2D2HL2S2
e e e e e b e e e e e  b e e e b b e e e e e  b e e e b e e e e e e  e e e b e e e e e e e  e e b e e e e e b e e
f f f f s f f s f s f f f f  s f f f f f f f f f f  s f f f f f f f f f f  f f f s s f         f f f s f f f f f s f

351
LH2RR2RR2HD2TM2
e b e e e e e e e e e
f s f f f f

```

The conservation scale:

Variable     Average     Conserved

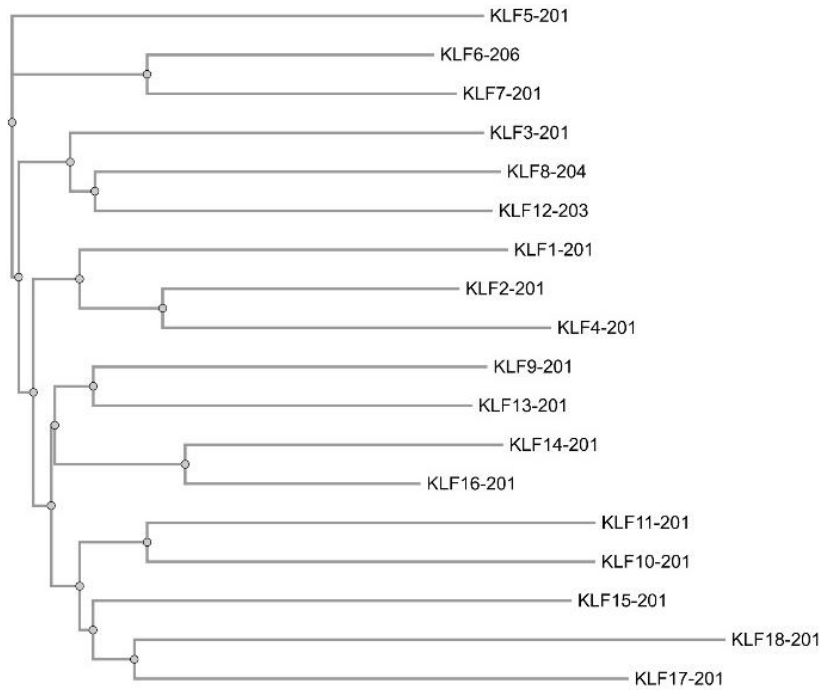
```

 1 2 3 4 5 6 7 8 9

```

- e** - An exposed residue according to the neural network algorithm.
- b** - A buried residue according to the neural network algorithm.
- f** - A predicted functional residue (highly conserved and exposed).
- s** - A predicted structural residue (highly conserved and buried).
- x** - Insufficient data - the calculation for this site was performed on less than 10% of the sequences.

Figure 4.5: Conservation Analysis by ConSurf

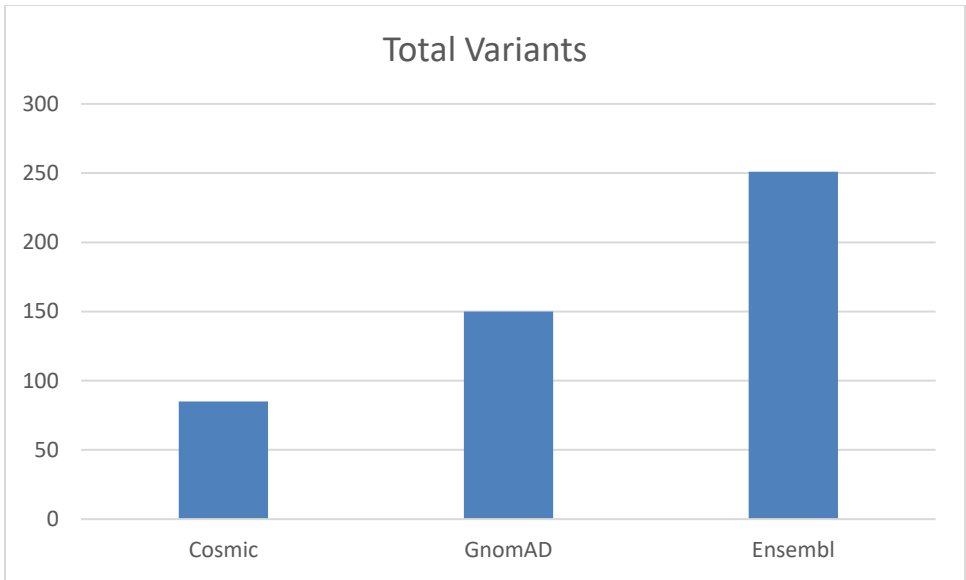


**Figure 4.6** Phylogenetic Tree of the KLF proteins. All the KLFs originated from a common root and then evolve into different classes

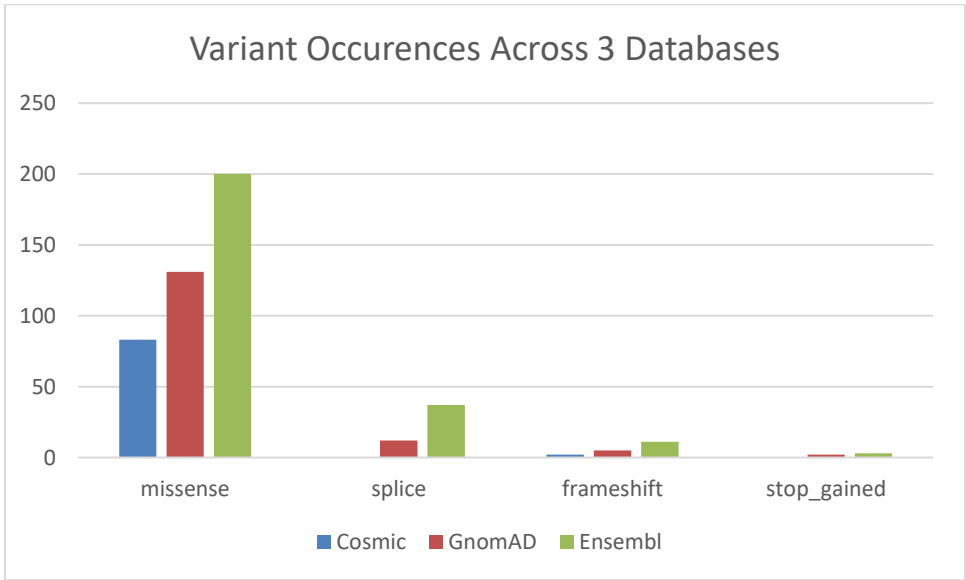
#### 4.1.3 KLF8 Variants Identification

A total of 486 variants of KLF8 gene, including missense, nonsense, spliced and frameshift variants, were retrieved from multiple databases including Ensembl, COSMIC, and genomAD, containing 251, 85 and 150 variants respectively. Although, Ensembl database included 200 missense, 11 frameshift, 37 spliced variants, 3 nonsense variants, while variants from COSMIC database include 83 missense, 2 frameshift, 0 nonsense and splice variants and genomAD had 131 missense, 5 frameshift, 12 spliced variants and 2 nonsense. Out of 486 variants, 200 were unique and 286 variants were redundant. Unique 200 variants include 159 missense, 16 frameshift, 7 nonsense and 18 spliced variants.

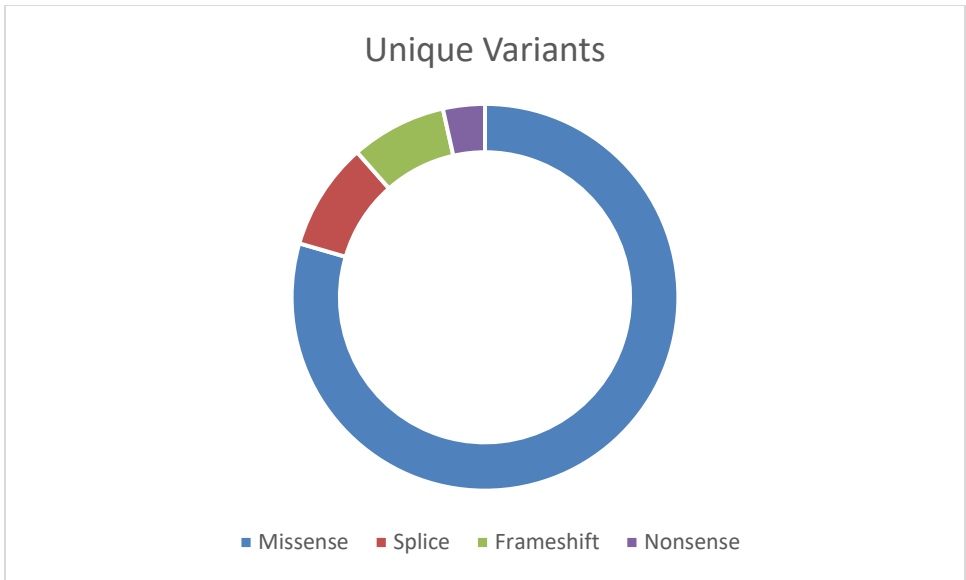




**Figure 4.7** Total variants acquired from all three databases (Cosmic, GnomAD, Ensembl)

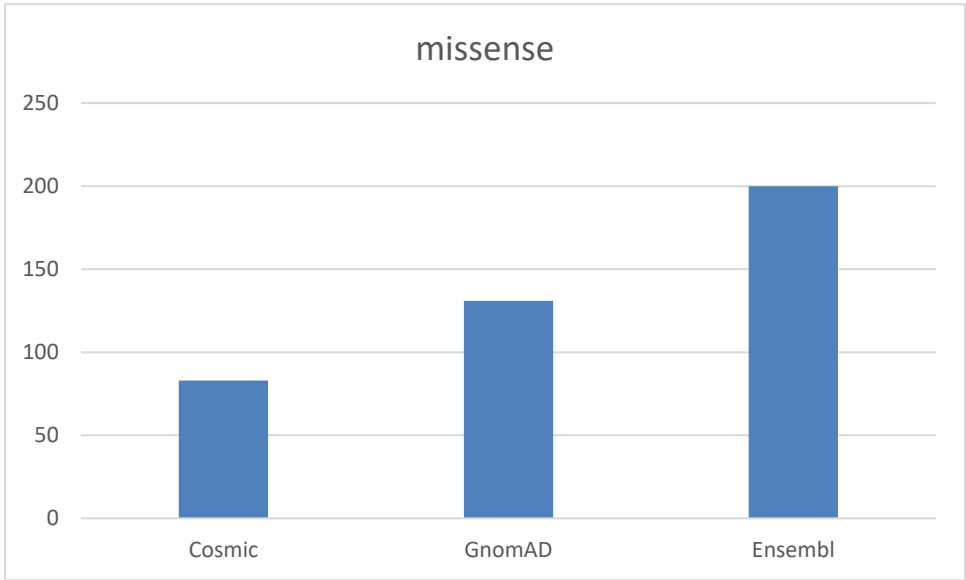


**Figure 4.8** Total Variant Occurrences across all three Databases



**Figure 4.9** Filtered unique variants from all three databases

Missense variants were filtered from these three databases shown in the figure below.



**Figure 4.10** Missense variants filtered from three databases

#### 4.1.4 KLF8 Missense SNPs Filtration

The missense data was further filtered to retrieve the SNPs for further investigation. This filtration was based on their pathogenicity percentages. SIFT, PolyPhen, REVEL, Mutation Assessor, and MetaLR tools were used to calculate pathogenicity scores as stated in the methodology. Following table shows the pathogenicity scoring and percentages.

**Table 4.2** Pathogenicity table of missense variants

Serial no.	Variant ID	AA coord	sift_class	polyphen_class	revel_class	meta_lr_class	mutation_assessor_class	Pathogenicity %
1	rs911255592	3	0	0	-	-	-	0%
2	rs776924073	4	1	0	0	0	0.5	30%
3	rs750116979	5	1	0	0	0	0	20%
4	rs1457982343	7	1	0	0	0	0	20%
5	rs759821417	9	0	0	0	0	0	0%
6	rs1602425063	15	0	1	0	0	0.5	30%
7	rs765607510	17	1	0	0	0	0	20%
8	rs763410667	18	1	0	0	0	0.5	30%
9	rs1409861399	19	1	0	0	0	0	20%
10	rs1329785837	19	1	0	0	0	0	20%
11	rs764459709	20	1	0	0	0	0	20%

12	rs752124107	24	0	0	0	0	0	0%
13	rs1229194237	26	1	0	0	0	0.5	30%
14	rs1438658965	26	1	0	0	0	0.5	30%
15	rs1438658965	26	1	0	0	0	0.5	30%
16	rs1398077363	28	1	1	0	0	0.5	50%
17	rs1040382871	29	0	1	0	0	0.5	30%
18	rs1340917869	29	1	1	0	0	0.5	50%
19	rs368900678	33	1	0	0	0	0.5	30%
20	rs144407506	33	0	0	0	0	0.5	10%
21	rs144407506	33	0	0	0	0	0.5	10%
22	rs774648329	34	1	0	0	0	0.5	30%
23	rs762366439	37	0	0	0	0	0.5	10%
24	rs762366439	37	0	0	0	0	0.5	10%
25	rs768058968	38	0	0	0	0	0.5	10%
26	rs759073766	40	0	0	0	0	0.5	10%
27	rs1417633235	42	0	0	0	0	0.5	10%
28	rs368933949	43	1	1	0	0	0.5	50%
29	rs758061219	48	1	1	0	0	0.5	50%
30	rs777618241	49	0	0	0	0	0	0%
31	rs751362149	50	0	0	0	0	0.5	10%

32	rs757188676	51	0	0	0	0	0.5	10%
33	rs757188676	51	0	0	0	0	0.5	10%
34	rs1253393491	52	1	1	0	0	0.5	50%
35	rs899019890	53	1	1	0	0	0.5	50%
36	rs745893681	53	0	0	0	0	0	0%
37	rs745893681	53	1	0	0	0	0.5	30%
38	rs770010841	53	0	0	0	0	0	0%
39	rs780360757	56	1	0	0	0	0.5	30%
40	rs1342034737	57	1	1	0	0	0.5	50%
41	rs1245069875	58	1	0	0	0	0.5	30%
42	rs139081150	61	0	0	0	0	0	0%
43	rs1050502377	62	1	1	0	0	0.5	50%
44	rs768821062	62	1	1	0	0	0.5	50%
45	rs774753989	69	1	1	0	0	0.5	50%
46	rs887584045	69	1	1	0	0	0.5	50%
47	rs1306007343	70	1	1	0	1	0.5	70%
48	rs1200383021	70	1	1	0	1	0.5	70%
49	rs868781835	71	1	1	1	1	0.5	90%
50	rs143969730	75	1	1	0	0	0.5	50%
51	rs772525064	80	0	0	0	0	0.5	10%

52	rs1379874289	81	0	1	0	0	0	20%
53	rs1160583250	84	1	1	0	0	0.5	50%
54	rs1361224783	87	1	1	0	0	0.5	50%
55	rs1361224783	87	1	1	0	0	0.5	50%
56	rs1404365594	89	1	1	0	0	0.5	50%
57	rs773726642	94	1	1	0	0	0.5	50%
58	rs758988041	95	1	1	0	0	0.5	50%
59	rs1451874481	98	0	1	0	0	0	20%
60	rs752213619	100	0	1	0	0	0.5	30%
61	rs752213619	100	0	1	0	0	0.5	30%
62	rs1330392987	102	0	1	0	0	0.5	30%
63	rs1047950271	104	0	1	0	0	0.5	30%
64	rs1427759514	106	0	0	0	0	0.5	10%
65	rs146429909	108	0	0	0	0	0.5	10%
66	rs139841730	109	0	0	0	0	0	0%
67	rs139841730	109	0	0	0	0	0	0%
68	rs1024570270	109	0	0	0	0	0.5	10%
69	rs753632321	110	0	0	0	0	0.5	10%
70	rs753632321	110	0	0	0	0	0	0%
71	rs756190986	111	1	1	0	0	0.5	50%

72	rs756190986	111	1	1	0	0	0.5	50%
73	rs756190986	111	1	0	0	0	0.5	30%
74	rs1239437887	114	0	0	0	0	0.5	10%
75	rs780171768	115	0	0	0	0	0	0%
76	rs749373892	115	0	0	0	0	0.5	10%
77	rs1383363632	117	0	0	0	0	0.5	10%
78	rs374892798	118	0	1	0	0	0.5	30%
79	rs1180735183	119	0	0	0	0	0.5	10%
80	rs779031240	120	0	0	0	0	0	0%
81	rs748491106	120	0	0	0	0	0.5	10%
82	rs1344440945	122	0	0	0	0	0.5	10%
83	rs143280924	124	0	1	0	0	0.5	30%
84	rs771483872	126	0	1	0	0	0.5	30%
85	rs1397842630	129	0	0	0	0	0.5	10%
86	rs775036329	129	0	0	0	0	0	0%
87	rs775036329	129	0	0	0	0	0	0%
88	rs1221527601	134	0	0	0	0	0	0%
89	rs1221527601	134	0	0	0	0	0.5	10%
90	rs763754140	135	1	0	0	0	0.5	30%
91	rs763754140	135	1	0	0	0	0.5	30%

92	rs757211205	138	1	1	0	0	0.5	50%
93	rs36102105	139	1	1	0	0	0.5	50%
94	rs1314515814	140	1	0	0	0	0.5	30%
95	rs913427803	141	0	0	0	0	0	0%
96	rs767419429	142	1	0	0	0	0.5	30%
97	rs750305226	143	1	1	0	0	0.5	50%
98	rs756101201	145	1	1	0	0	0.5	50%
99	rs200907884	148	1	1	0	0	0.5	50%
100	rs755032067	150	1	0	0	0	0.5	30%
101	rs1433771646	150	1	0	0	0	0.5	30%
102	rs779134931	156	0	0	0	0	0	0%
103	rs1169792053	161	1	1	0	0	0.5	50%
104	rs369282415	162	1	1	0	0	0.5	50%
105	rs778094890	165	1	1	0	0	0.5	50%
106	rs747415809	166	1	0	0	0	0.5	30%
107	rs1209125016	166	1	1	0	0	0.5	50%
108	rs1569186150	167	0	0	0	0	0.5	10%
109	rs1569186150	167	1	0	0	0	0.5	30%
110	rs771429775	168	1	1	0	0	0.5	50%
111	rs777027560	169	0	1	0	0	0.5	30%



112	rs1054205139	169	1	1	0	0	0	40%
113	rs748774811	174	0	0	0	0	0	0%
114	rs1220892845	180	0	0	0	0	0	0%
115	rs1274128070	182	1	0	0	0	0.5	30%
116	rs1321414525	182	1	0	0	0	0.5	30%
117	rs761546600	187	0	1	0	0	0.5	30%
118	rs140327143	188	0	0	0	0	0	0%
119	rs1260887085	189	1	1	0	0	0.5	50%
120	rs1050451843	190	1	0	0	0	0	20%
121	rs1194175582	193	1	1	0	0	0.5	50%
122	rs760574954	195	1	1	0	0	0.5	50%
123	rs766287376	195	1	1	0	0	0.5	50%
124	rs753727815	197	0	0	0	0	0	0%
125	rs753727815	197	0	0	0	0	0	0%
126	rs1569186275	198	0	1	0	0	0	20%
127	rs755018799	199	1	1	0	0	0.5	50%
128	rs1381116096	200	1	1	0	0	0.5	50%
129	rs1418504822	201	1	1	0	0	0.5	50%
130	rs1298150637	205	1	1	0	0	0	40%
131	rs1359243793	206	1	1	0	0	0	40%

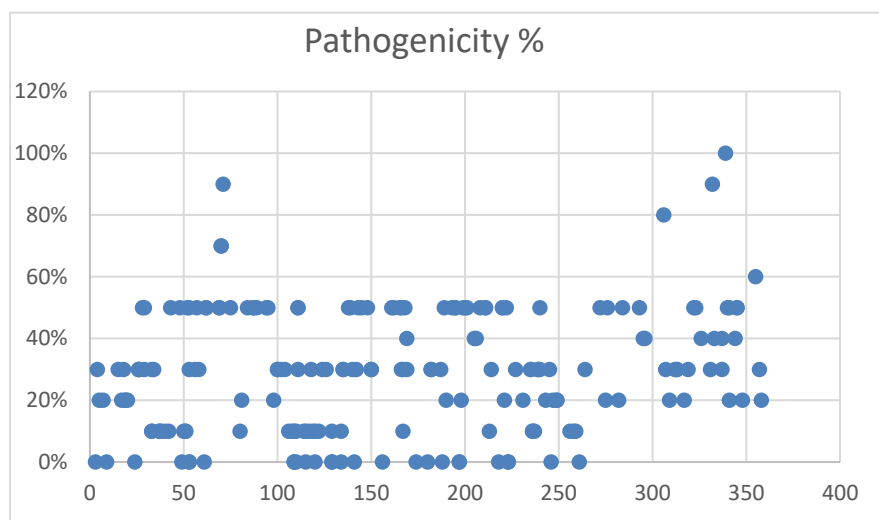
132	rs1402386561	208	1	1	0	0	0.5	50%
133	rs1484964264	211	1	1	0	0	0.5	50%
134	rs1484964264	211	1	1	0	0	0.5	50%
135	rs376737146	213	0	0	0	0	0.5	10%
136	rs754199549	214	0	1	0	0	0.5	30%
137	rs1018147187	218	0	0	0	0	0	0%
138	rs1274299131	220	1	1	0	0	0.5	50%
139	rs1274299131	220	1	1	0	0	0.5	50%
140	rs1220870159	221	1	0	0	0	0	20%
141	rs765267281	222	1	1	0	0	0.5	50%
142	rs771485287	223	0	0	0	0	0	0%
143	rs200006019	223	0	0	0	0	0	0%
144	rs1324948642	227	1	0	0	0	0.5	30%
145	rs370185272	231	0	1	0	0	0	20%
146	rs145056308	235	1	0	0	0	0.5	30%
147	rs1157044023	235	1	0	0	0	0.5	30%
148	rs757569499	236	0	0	0	0	0.5	10%
149	rs1456723587	236	0	0	0	0	0.5	10%
150	rs781401053	237	0	0	0	0	0.5	10%
151	rs763246473	239	1	0	0	0	0.5	30%

152	rs766866162	239	1	0	0	0	0.5	30%
153	rs780734959	240	0	1	0	0	0.5	30%
154	rs747609626	240	1	1	0	0	0.5	50%
155	rs1160586311	243	1	0	0	0	0	20%
156	rs995444916	245	0	1	0	0	0.5	30%
157	rs1402780164	246	0	0	0	0	0	0%
158	rs142294522	247	0	1	0	0	0	20%
159	rs1477147807	249	0	1	0	0	0	20%
160	rs1477147807	249	0	1	0	0	0	20%
161	rs1433525053	256	0	0	0	0	0.5	10%
162	rs755606346	258	0	0	0	0	0.5	10%
163	rs1057099030	259	0	0	0	0	0.5	10%
164	rs1602457545	261	0	0	0	0	0	0%
165	rs1287721402	264	1	0	0	0	0.5	30%
166	rs756827184	272	1	1	0	0	0.5	50%
167	rs745593957	275	0	1	0	0	0	20%
168	rs1379763969	276	1	1	0	0	0.5	50%
169	rs1290497846	282	1	0	0	0	0	20%
170	rs1385427586	284	1	1	0	0	0.5	50%
171	rs749213915	293	1	1	0	0	0.5	50%

172	rs139041929	295	1	1	0	0	0	40%
173	rs774267049	296	1	1	0	0	0	40%
174	rs760877139	306	1	0	1	1	1	80%
175	rs1391902410	307	1	0	0	0	0.5	30%
176	rs969886508	309	1	0	0	0	0	20%
177	rs766842755	312	1	0	0	0	0.5	30%
178	rs754312008	313	1	0	0	0	0.5	30%
179	rs1331025659	313	1	0	0	0	0.5	30%
180	rs1322306475	317	1	0	0	0	0	20%
181	rs1405382114	319	1	0	0	0	0.5	30%
182	rs755518394	322	1	1	0	0	0.5	50%
183	rs1204570188	323	1	1	0	0	0.5	50%
184	rs1286585777	326	1	1	0	0	0	40%
185	rs1380748194	331	0	1	0	0	0.5	30%
186	rs1456582819	332	1	1	1	1	0.5	90%
187	rs1569194785	333	1	1	0	0	0	40%
188	rs1251320193	337	0	1	0	0	0.5	30%
189	rs756805881	337	1	1	0	0	0	40%
190	rs1196572444	339	1	1	1	1	1	100%
191	rs780645349	340	1	1	0	0	0.5	50%

192	rs1016774145	341	1	1	0	0	0.5	50%
193	rs1470090266	341	1	0	0	0	0	20%
194	rs1169656171	344	1	1	0	0	0	40%
195	rs1397734593	345	1	1	0	0	0.5	50%
196	rs745530968	345	1	1	0	0	0.5	50%
197	rs1298321992	348	1	0	0	0	0	20%
198	rs1555939511	355	1	1	1	0	0	60%
199	rs1281874686	357	1	unknown	0	0	0.5	30%
200	rs201887872	358	1	unknown	0	0	0	20%

After that, the percentage pathogenicity for all variants was determined and plotted as shown in the following graph.



**Figure 4.11** Pathogenicity percentages of the missense variants

Out of 200 missense variants, threshold 85% pathogenicity was adjusted to filter out deleterious variants that could affect resultant protein structure and functions. After

applying threshold pathogenicity filter, 3 pathogenic variants were obtained with pathogenicity 85% or above shown in table 4.2. and two variants (rs868781835, and rs1196572444) at amino acid number 71 and 339 with highest pathogenicity 90% and 100% respectively were shortlisted for further protein analysis.

**Table 4.3** missense variants after threshold pathogenicity sorting

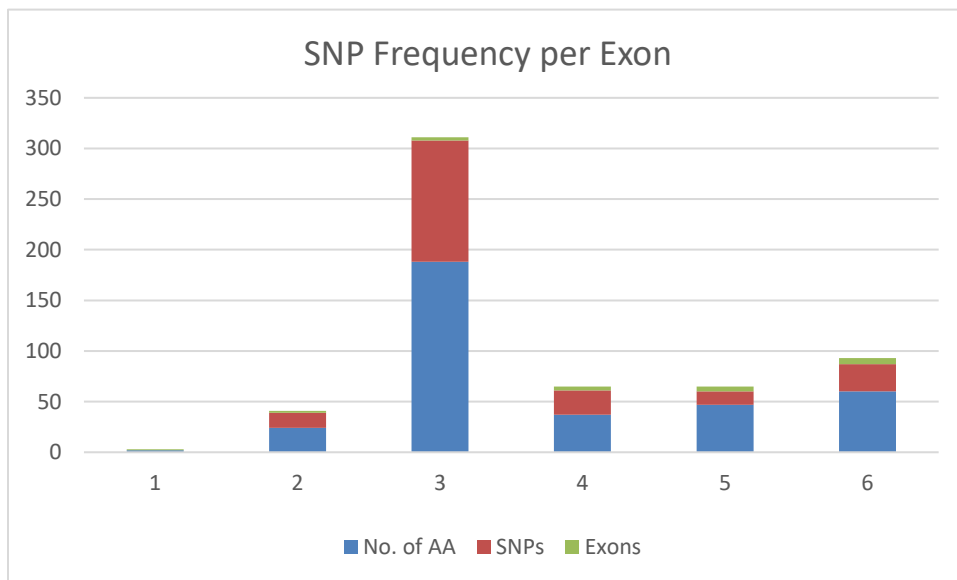
Variant ID	Alleles	AA	AA coord	sift_class	polyphen_class	revel_class	meta_lr_class	mutation_assessor_class
rs868781835	C/T	P/L	71	D	PD	LDC	DA	M
rs1456582819	A/G	K/E	332	D	PD	LDC	DA	M
rs1196572444	T/G	C/G	339	D	PD	LDC	DA	H

(D= Deleterious, PD= Probably damaging, LDC= Likely disease causing, D= damaging, H= High, M= Medium)

**Table 4.4:** Selected Missense Variants according to Pathogenicity

Variant ID	AA coord	Conseq. Type	sift_class	polyphen_class	revel_class	meta_lr_class	mutation_assessor_class	Pathogenicity %
rs868781835	71	missense variant	1	1	1	1	0.5	90%
rs1456582819	332	missense variant	1	1	1	1	0.5	90%
rs1196572444	339	missense variant	1	1	1	1	1	100%

The SNP frequency per Exon was analyzed as shown below.



**Figure 4.12** SNP Frequency per Exon

#### 4.1.5 Protein Stability Analysis

MuPro, DynaMut and Maestroweb were used to do the structural and functional investigation of the SNPs rs1196572444 and rs868781835. MUpro generated a DDG value of -1.7939459 kcal/mol for SNP1, indicating decreasing stability of the protein structure and a value of 0.75721874 kcal/mol, indicating an increasing stability. DynaMut DDG values indicated decreasing stability of both the protein structures associated with the SNPs under study. The DDG value for both the SNPs is 0.383 and 0.856 Kcal/mol, according to MaestroWeb results, indicating an increase in stability.

**Table 4.5:** Protein Stability Analysis

Tools	rs1196572444		rs868781835	
	DDG Value	Consequence	DDG Value	Consequence
<b>MUpro</b>	-1.7939459 kcal/mol	Decreasing Stability	0.75721874 kcal/mol	Increasing Stability
<b>DynaMut</b>	0.056 kcal/mol	Decreasing Stability	-0.292 kcal/mol	Decreasing Stability
<b>MAESTRO WEB</b>	0.383 kcal/mol	Increasing Stability	0.856 kcal/mol	Increasing Stability

#### 4.1.6 Structural and Functional Analysis of Variants

MutPred2, DynaMut, and HOPE were used to check the effect of amino acid variants on the structure and function of total proteins. The SNPs rs1196572444 and rs868781835 had a general probability score of 0.882 and 0.762 according to MutPred2.

According to Mutpred results, inducing the mutation C339G can result in an altered disordered interface and gain of N-linked glycosylation at N340. Also, the induction of mutation P71L can cause loss of intrinsic disorder, loss of B-factor and loss of ubiquitylation at K67.

**Table 4.6:** Mutpred2 Scoring

ID	Substitution	MutPred2 score
<b>KLF8_HUMAN</b>	<b>P71L</b>	<b>0.762</b>
<b>KLF8_HUMAN</b>	<b>C339G</b>	<b>0.882</b>



**Table 4.7** MutPred2 Results

<b>Residual Change</b>	<b>Mechanism</b>	<b>P-value</b>
<b>C339G</b>	Altered Disordered interface	0.02
	Gain of N-linked glycosylation at N340	0.03
<b>P71L</b>	Loss of Intrinsic disorder	9.3e-03
	Loss of B-factor	0.03
	Loss of Ubiquitylation at K67	0.04

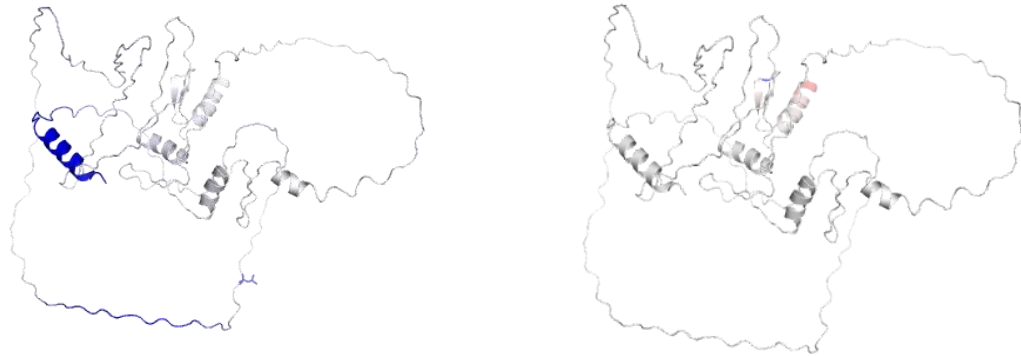
DynaMut2 calculated the vibrational entropy energy between the wild-type and mutant proteins to demonstrate a difference in protein flexibility because of mutation. The results are depicted in the following table.

**Table 4.8:** DynaMut2 Results

<b>Variant</b>	$\Delta\Delta S_{\text{vib}} \text{ ENCoM}$	$\Delta\Delta G$
<b>C339G</b>	-0.070 kcal.mol <sup>-1</sup> .K <sup>-1</sup> (Decrease of molecule flexibility)	0.056 kcal/mol (Destabilizing)
<b>P71L</b>	0.365 kcal.mol <sup>-1</sup> .K <sup>-1</sup> (Increase of molecule flexibility)	-0.292 kcal/mol (Destabilizing)

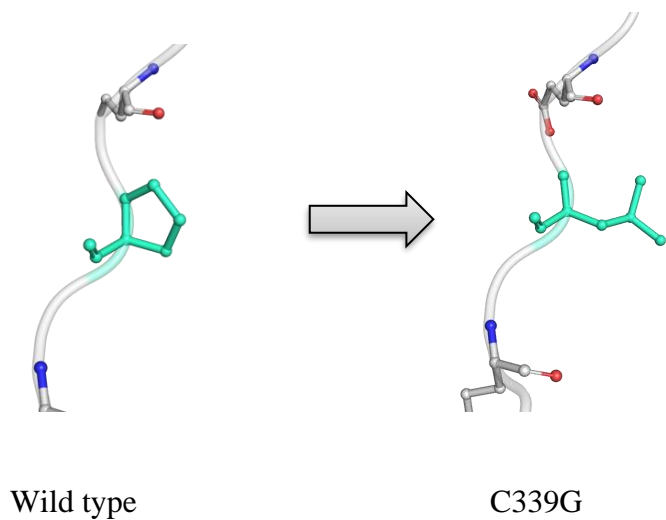
C339G

P71L

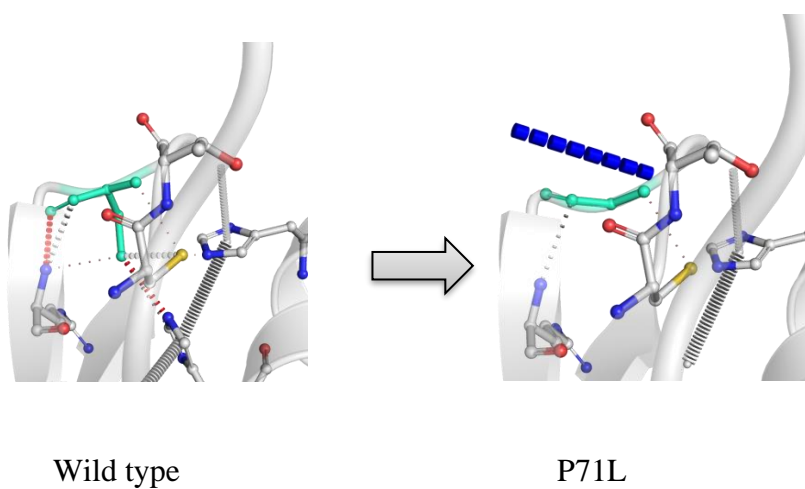


**Figure 4.13:** Variant Structures. Colored Amino acids show change in vibrational entropy due to mutation. **BLUE** represents a rigidification of the structure and **RED** a gain in flexibility

Intra-atomic interactions were also predicted by DynaMut. The results indicate different amino acid interactions when compared to wild type. The wild type contained cysteine which was converted to glycine at 339th position in one of the SNPs. In another selected SNP, proline was converted to leucine at position 71.



**Figure 4.14:** Intra-atomic Interactions of wildtype and C339G variant



**Figure 4.15:** Intra-atomic Interactions of wildtype and P71L variant

In addition to these techniques, the HOPE report shown below in Table revealed that both the variant residue is substantially larger than the wild-type residue, possibly resulting in bumps in the protein structure.

**Table 4.9:** HOPE Report

SNP ID	Variant Amino Acid			Variant Location		Functional	Damage	
	size	charge		hydrophobicity	Domain	Conservation		
rs1196572444	smaller	Wild	Variant	More hydrophobic	Zinc-finger domain	Highly conserved	Effects on DNA binding	damaging
		Neutral	Neutral					
rs868781835	bigger	Neutral	Neutral	-	-	Highly conserved		Probably damaging

The variant and the wild type residue were neutral even after the mutation, C339G and P71L as shown in the figures below.



**Figure 4.16:** Cysteine is shown on the right side which is converted to Glycine located on the left. The size of Cysteine is larger than Glycine as the -SH group is removed.

At this position, mutation converts cysteine to glycine which reduces the rigidity of structure and this disturbs the zinc-finger domain. The resulting residue is less hydrophobic as compared to the wild-type. This will lead to a loss of hydrophobic interactions at the core or surface of the protein.

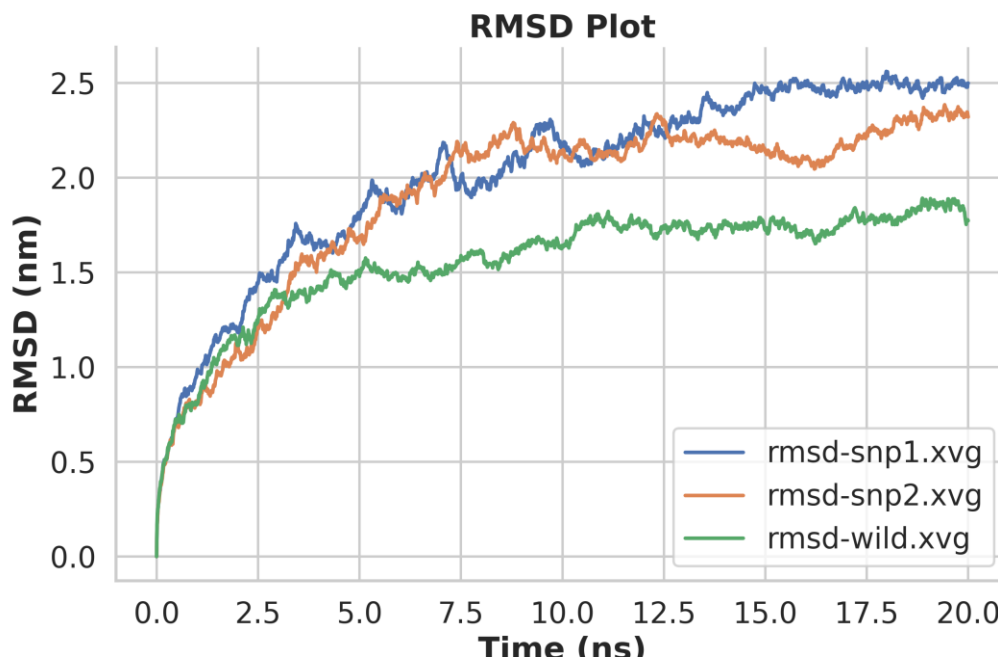


**Figure 4.17** Proline is shown on the right side which mutates to Leucine shown on the left side. The size of Leucine is larger than Proline as the ring structure is opened

This mutation leads to a reduction in rigidity of the protein structure. The mutant residue is larger. This can disturb the structure through the formation of bumps. The HOPE report does not mention the effect of this mutation on hydrophobicity and domains of the protein structure.

#### *4.1.7 Results of Molecular Dynamics (MD) Simulation*

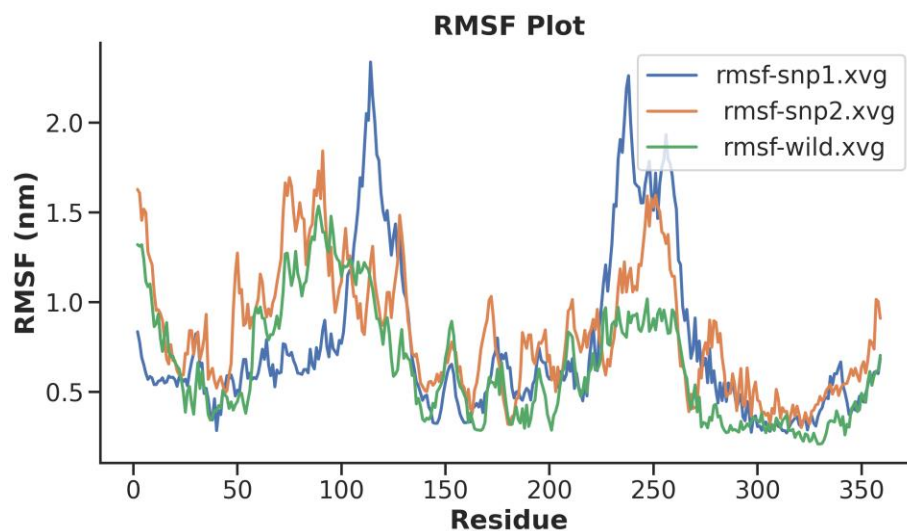
The files containing data on MD simulations were run for the wild-type and variant proteins. To analyze the results, graphs were generated through Gromancer. To examine the differences between wild type and mutant proteins, four characteristics were used: root mean square deviation (RMSD), root mean square fluctuation (RMSF), radius of gyration, number of hydrogen bonds, and solvent accessible surface area (SASA) analysis. The variation of distinct atoms in a protein from its typical position is shown by the root mean square deviation (RMSD). In comparison to wild type, RMSD analysis of the wild and mutant proteins found that the altered protein deviates greatly from its reference position. By comparing the RMSD values, the structural deviations of both wild type and variant proteins over the period of 20ns were determined and plotted in a graph. Both the variant proteins showed deviations when compared to the wildtype. The maximum deviation could be seen around 20 ns starting from around 0.3 ns. This shows that the variants have low stability.



**Figure 4.18:** RMSD plot for wild-type and variant proteins

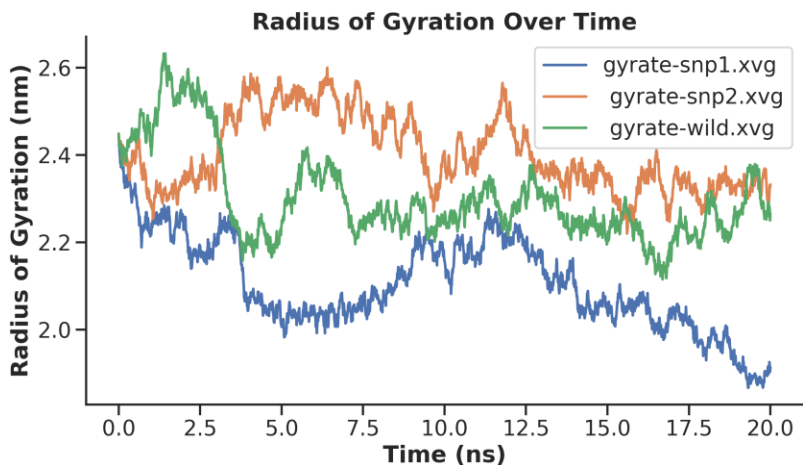
Individual residues' deviation from their mean position is depicted by root mean square fluctuation (RMSF). A higher RMSF number indicates more flexible movements during simulation, whereas a lower RMSF value indicates limited movements in comparison to its typical position. The difference in the fluctuation of wild and mutant protein residues can be seen using RMSF analysis in figure 25. To see the difference in the dynamic behavior of residues caused by the mutations, RMSFs of wild and both mutant KPCG structures were computed. The peaks represent the residues' highest kinetic energies.

The fluctuation in residues of wild and both variant KLF8 structures ranges from 0.1nm to >2nm. The C339G variant shows higher peaks than the wild and other variant, P71L. The region around 100-150 residues and 200-300 show a higher peaks for SNP1 (C339G) and the region 50-100 and 200-250 show highest peaks for the variant P71L. The wild-type shows highest peaks around 50-150 only. The overall graph shows nearly similar fluctuations for SNP2 while SNP1 shows much difference where the fluctuations are decreasing considerably after 300 residues.



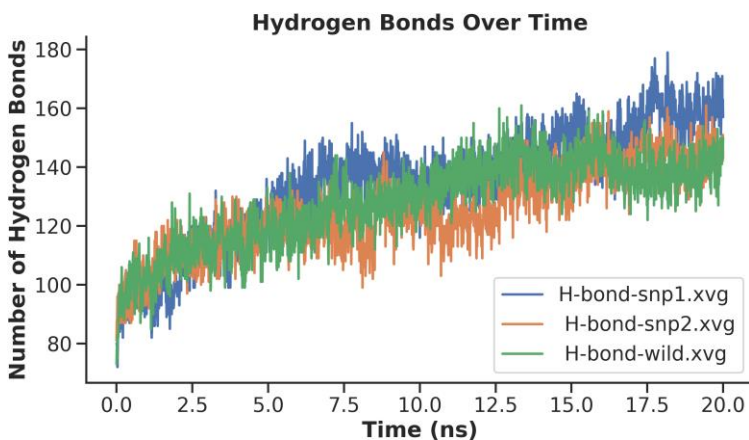
**Figure 4.19:** RMSF plot of wild-type and variant proteins

The radial distance of all the atoms in a protein from their common axis is known as the radius of gyration ( $R_g$ ).  $R_g$  is the radius of a protein structure in a dynamic situation, indicating the protein's compactness and changes in folding over time. The computed radius of gyration for wild and modified proteins is shown in figure 4.20. it can be seen that the  $R_g$  value of wild-type protein increases significantly at 2.5 ns and then fluctuates throughout the course of simulation. However, for SNP1 the values drop in the beginning, then increase and then again decrease around 20ns showing an increase in compactness. For SNP2, the graph fluctuates but the values concluded around 20ns are similar to the wild-type.



**Figure 4.20:** Radius of gyration of wild-type and variant proteins

Between the polar side chains of amino acid residues, hydrogen bonds develop. The amount of hydrogen bonds formed in all KLF8 structures throughout a 20ns simulation period is depicted in figure 4.21. The number of hydrogen bonds in variant P71L can be seen to be decreasing around 7.5 to 15ns when compared to the wild-type. The wild-type KLF8 hydrogen bonds are overall increasing during the 20ns simulation. However, variation is observed for C339G where the hydrogen bonds are even more in number than the wild KLF8.

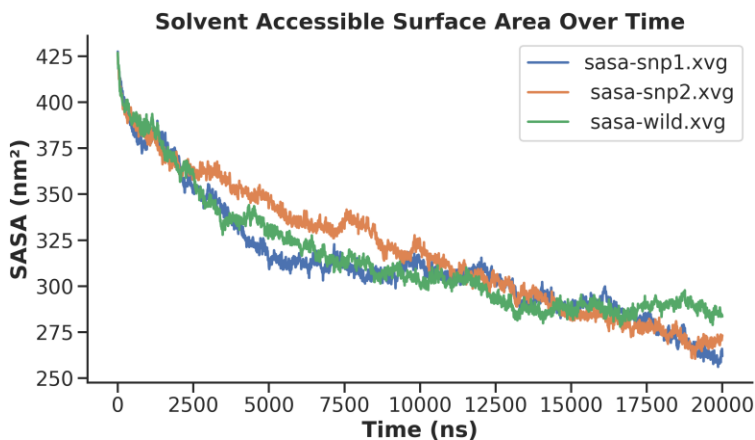


**Figure 4.21:** Difference in number of hydrogen bonds in wild-type and variant proteins

Calculating the solvent accessible surface area (SASA) is a simple technique to address



the surface features of peptides or protein, such as whether the surface is polar or non-polar, or to distinguish between exposed and buried sections or amino acids. the figure 4.22, shows the SASA plot.

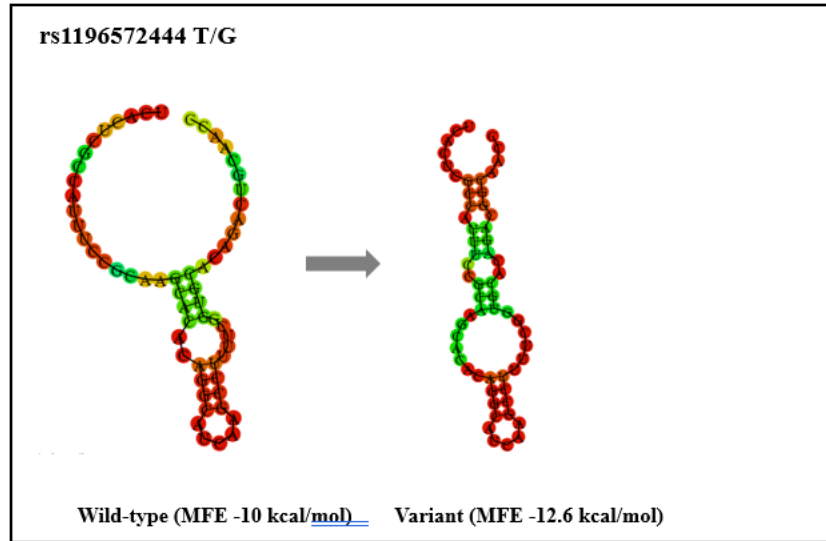


**Figure 4.22:** SASA Analysis of wild-type and variant proteins

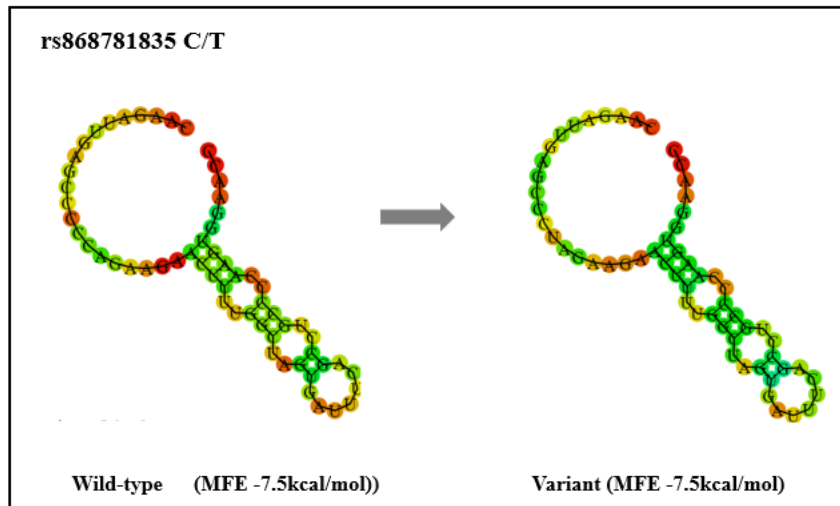
The SASA value of both variant protein structures is overall smaller than that of the wild type protein structure, indicating that variant structures are more stable, particularly the C339G variant. The wild and mutant protein structures SASA range is between 250 and 425 nm<sup>2</sup>. The SASA value begins to decrease for the wild-type as well as the variant proteins throughout the simulation period. Both variations resulted in a decrease in the surface area of protein, indicating a decrease in the solvent accessible surface area, which leads to an increase in protein stability. Figure 4.22 shows the graphical representation of surface area over the stimulation period.

#### 4.1.8 Effects of SNPs on mRNA Secondary Structure

The mRNA secondary structure of KLF8 wild-type and variant was predicted applying in-silico method. For both, the wild-type and variant, minimum free energy was calculated. For the variant rs1196572444 T/G, the minimum free energy (MFE) of RNA secondary structure changed from -10 kcal/mol of wild-type to -12.6 kcal/mol of the variant. As a result, a significant change in structure was also observed. The decrease in minimum free energy signifies an increase in stability of the mRNA structure. In case of the variant rs868781835 C/T, no change in MFE was observed. This showed a value of -7.5 kcal/mol.



**Figure 4.23:** Minimum free energy of wild-type and variant C339G



**Figure 4.24:** Minimum free energy of wild-type and variant P71L

## 4.2 Wet-lab Analysis

### 4.2.1 Genotype Analysis of Breast Cancer Patients and Control Samples

The genotype analysis showed that the genotype TT has an odds ratio and relative risk of 40.16 and 6.143. These values indicate that these homozygous alleles could be related to breast cancer development and association when compared to homozygous GG or heterozygous GT. The homozygous genotype GG shows an odds ratio of 0.1856 and relative risk of 0.3293 which means that this might not be involved in the disease.

**Table 4.10** Genotyping data of C339G (rs1196572444)

Genotype/ Alleles	Cancer %	Control %	Odds Ratio	Relative risk	P-value
<b>GG</b>	3%	14.29%	0.1856	0.3293	<0.005
<b>GT</b>	11%	72.45%	0.047	0.1748	
<b>TT</b>	86%	13.27%	40.16	6.143	
<b>G</b>	8%	23.15%	0.2887	0.4611	<0.005
<b>T</b>	92%	76.85%	3.464	2.169	

### 4.2.2 Association of C339G SNP (rs1196572444) with Metastasis in Breast Cancer

The genotype GG and GT have higher values than the TT genotype. However, the genotype GT has the highest value. Therefore, we can say that the patients possessing this genotype might have a chance of metastasizing.

**Table 4.11:** Statistical Analysis of Genotype relation with metastasis

<b>Genotype</b>	<b>Disease State</b>					
	<b>Metastatic</b>			<b>Non-metastatic</b>		
	<b>P-value</b>	<b>Relative Risk</b>	<b>Odds Ratio</b>	<b>P-value</b>	<b>Relative Risk</b>	<b>Odds Ratio</b>
<b>GG</b>	>0.9999	1.176	1.527	>0.9999	0.7698	0.6548
<b>TT</b>	0.3832	0.7651	0.4821	0.3832	1.587	2.074
<b>GT</b>	0.3428	1.321	2.177	0.3428	0.6068	0.4594

## CHAPTER 5: DISCUSSION

The accumulation of mutations in genes that regulate cell cycle, apoptosis, invasiveness and metastasis leads to cell transformation in the form of malignancy. This happens because such genes act as caretaker genes for maintaining genomic stability (Listgarten et al., 2004). Breast cancer development occurs either due to modifiable or non-modifiable factors. Out of these, a non-modifiable factor is single-nucleotide polymorphism (SNP) (Łukasiewicz et al., 2021b). These SNPs are accountable for increasing cancer susceptibility (Hubner & Houlston, 2017).

The KLF family of proteins is the transcription factor family. These are either transcriptional activators or repressors and are zinc-finger proteins. There are a total of eighteen members of this family. KLF8 is one of them. This protein is involved in cell migration, proliferation, epithelial to mesenchymal transition, DNA repair, as well as resistance to drugs in cancer progression (Yuce & Ozkan, 2024). The aim of this study was to identify the most pathogenic variant of KLF8 and check its impact on overall protein structure and function. It further aimed at checking its link with breast cancer development and prognosis.

To check the localization of this protein, Deeploc 1.0 was utilized. Deeploc 1.0 uses deep learning to predict the subcellular localization of proteins found in eukaryotes. It is able to differentiate among ten different localizations. These include nucleus, cytoplasm, mitochondrion, cell membrane, endoplasmic reticulum, chloroplast, Golgi apparatus, lysosome or vacuole, peroxisome and extracellular. The predicted likelihood score turned out to be 0.998 in nucleus for KLF8. This indicates that this protein is found in the nucleus. The Clustal Omega software was used to find the evolutionary relationship between Kruppel-like Factor family members. KLF8 evolution was analyzed by aligning the sequences of all KPC family proteins to indicate conserved areas among the proteins in this family. The phylogenetic tree showed that all members of this family descended from a common ancestor and have some conserved regions present in all KLF isoforms. The conserved regions were visualized through ConSurf.

We searched the structure of KLF8 in Protein Data Bank (PDB). No structure was found so we used AlphaFold to predict its three-dimensional structure. This software applies scoring system to predict model confidence. While the 3D structure was predicted by AlphaFold, another software, Interpro was used to see the domains in KLF8. One N-terminal domain and three C2H2 domains were identified. according to AlphaFold, the three C2H2 zinc-finger domains showed high confidence score. The structure was validated using SAVES and Procheck analysis showed that majority of the angles lie in the most preferred regions pointing towards structure stability. This was depicted through a Ramachandran plot.

As this research study was aimed at checking the link between KLF8 SNPs and breast cancer, the SNP data was acquired from Ensembl Genome Browser. Then, the missense SNPs were filtered out and considered for further analysis as these have an impact on protein structure and sometimes, on its function as well. Further filtration was conducted to identify highly pathogenic variants. For this, the scores of various softwares were considered. These include SIFT, PolyPhen, Revel, Mutation Assessor, and MetaLR. Pathogenicity percentage of the variants was calculated by self-scoring the variants keeping benign and pathogenic on two extremes as 0 and 1 and a midpoint was set at 0.5. Then, a threshold value was selected as 85%. All these variants were shown in a graph pf pathogenicity percentage. Based on the threshold value, two highly pathogenic variants further analyzed. These had rs IDs, rs868781835(P71L) and rs1196572444(C339G).

Stability analysis of the selected variants was carried out using software tools such as MUpro, DynaMut and MaestroWeb. For the variant rs868781835(P71L), MUpro and MaestroWeb showed increasing stability while DynaMut showed decreasing stability. For the variant rs1196572444(C339G), MUpro and DynaMut showed decreasing stability while MaestroWeb showed increasing stability. The decreasing stability of protein indicates that the SNP might be associated with breast cancer.

For structural and functional analysis of selected SNPs, MutPred2, DynaMut and HOPE were used. These tools predict whether a change in protein structure may affect its function. MutPred2 gives a score for predicting the pathogenicity and has set 0.5 as a threshold for

pathogenicity. On the other hand, the value for our selected variants came out as 0.762 for P71L and 0.882 for C339G which renders them as pathogenic. Moreover for C339G, this showed an altered disordered interface and gain of N-linked glycosylation at N340 and for P71L, it showed loss of intrinsic disorder, loss of B-factor and loss of Ubiquitylation at K67. These mean that there is a decrease in stability of the mutant protein. DynaMut also predicted that the overall protein stability is decreasing due to the mutation. Moreover, the intra-atomic interactions were also altered as compared to the wild-type. Hope analysis revealed that the size of the mutant proteins differs from the wild-type. In case of SNP P71L the size was bigger with probably damaging effect, whereas for C339G, the size was smaller with damaging effect. These results suggest that a change in even a single amino acid can alter the protein structure and it may affect its function, leading to cancer development and disease progression.

The effect of such change on protein structure and function was also evaluated by observing the stability of mRNA secondary structure. In case of P71L, the change does not affect mRNA stability. Contrary to this, in SNP C339G, there is a change in stability. The minimum free energy for this structure changes from -10 kcal/mol to -12.6 kcal/mol. This decrease in MFE indicates an increase in stability of the structure.

As the in-silico analysis predicted that the polymorphism of KLF8 in SNP C339G might have a damaging role, leading to breast cancer development. For practically checking this effect, ARMS-PCR was conducted. Two groups, disease and control were made. Blood samples were collected from 100 subjects each and DNA was extracted. Then, two sets of primers, two inner and two outer were designed through Primer1. The obtained results were analyzed through statistical evaluation. According to these, the genotype GG might have a protective role in breast cancer development whereas the genotype TT might be involved in disease development as the odds ratio and relative risk turned out to be 40.16 and 6.143 for the genotype TT. The statistical analysis suggests that the genotype GG and GT might be involved in metastasis in breast cancer patients.

### *5.1 Conclusion*

Breast cancer is the most prevalent type of cancer in females worldwide and its incidence rate is also higher. There are a lot of ways in which the disease develops. The disease proves to be fatal as it is usually diagnosed at later stages. Early detection can lead to an increased chance of survival. However, biomarkers for detection and therapies might help in this aspect. Single nucleotide polymorphism is an important factor in disease development. In case of KLF8, little is known about its variants and their role in breast cancer.

In our study, the KLF8 SNPs were identified through in-silico analysis. Two of them were selected for structural and functional analysis through computational methods. One of these (rs868781835) turned out to be probably damaging while the other (rs1196572444) turned out to be damaging. This SNP was further chosen for genotype analysis. The results suggested that this was not associated with breast cancer development, but it was involved in disease progression through metastasis. The results might be different for different ethnicities.

This SNP can be used as a prognostic marker and a therapeutic target in breast cancer. The expression profile of this SNP calls is needed for further investigation as it might provide some cutting-edge findings in cancer therapy. Further research is required to validate the acquired results on a large sample size and different ethnicities. The results can also be applied to other cancer types to check the role of this variant of KLF8.



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