

Computational Investigation of Regulatory Region SNPs of Autophagy Gene *BECN1*

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ABSTRACT

The autophagy process plays a cytoprotective role and ensures the healthy survival of a cell. The role of autophagy has been implicated in various diseases, making it an essential candidate for therapeutic interventions. Beclin 1, a candidate autophagy protein, plays a critical role during autophagy initiation and maturation by interacting with various other autophagy proteins. Beclin1 has been reported to be involved in various human diseases. This study uses a computational approach to study the effect of non-coding region single nucleotide polymorphisms (SNPs) of gene encoding beclin1. RegulomeDB, SNP2TFBS, and PROMO ALLGEN were used to predict the effect of promoter region variants on transcription factor binding sites. SNPs located within 3'UTR were analysed by miRdSNP, PolymiRTS Database 3.0, miRNASNP-V3, MicroSNIPER, and miRmap. Nine promoter region variants that alter the transcription factor binding sites and 4 variants in 3'UTR were identified that either create a new target site for miRNA or disrupt an existing one. The functional analysis of these identified SNPs could be done experimentally to unravel their relation with a particular disease and the genetic predisposition of human subjects for a disease.

Keywords: Hypoxia; BECN1; Single nucleotide polymorphism; Promoter variants; 3'UTR

1. INTRODUCTION

The human body possesses an intricately capable and advanced intracellular system to manage and facilitate responses to acute stressors and dangerous situations. Under certain extreme conditions, specific intracellular pathways activate to promote optimal conditions for sustaining health and ensuring survival. Macroautophagy (hereafter referred to as autophagy) is a conserved intracellular, pro-survival mechanism that is fundamental for ensuring the healthy survival of a cell when exposed to stressful conditions. This intricately orchestrated process holds a central position in removing the elements that have the potential to cause harm to the cells¹. Autophagy encompasses three fundamental stages. Initiation begins with activating a protein complex, which initiates the formation of the phagophore, a precursor structure. This phagophore expands and encapsulates targeted cellular components in the elongation phase, culminating in an autophagosome with a double-membraned structure. Subsequently, during maturation, autophagosomes merge with lysosomes to create autophagolysosomes, where the enclosed cellular contents are subjected to degradation, leading to the recycling of vital biomolecules essential for maintaining cellular health and homeostasis¹. Each step of autophagy is precisely controlled and intricately operated

by the actions of proteins encoded by autophagy-related genes (*ATGs*)^{2,3}. The basal autophagy functions to uphold cellular homeostasis, consequently, any aberration in the process can inflict damage upon the cell³. Autophagy is a well-reported contributing factor in a range of diseases, including neurodegenerative disorders, cancer, hepatic, muscular, infectious, and cardiovascular diseases as well as lung diseases and asthma³⁻⁵.

India is mandated to deploy a substantial military contingent in the western Himalayan region primarily to safeguard national security interests and preserve territorial integrity. Except for a minority representing hill tribes, the majority of these troops hail from lowland regions and have been nurtured in tropical and subtropical environments. Consequently, they confront significant challenges arising from rigorous environmental conditions, prevalent in the western Himalayas. These challenges encompass extremely frigid temperatures, hypobaric hypoxia (characterised by low oxygen pressure at elevated altitudes), pronounced aridity, elevated levels of solar ultraviolet radiation, and formidable winds^{6,7}. To tackle such conditions, both external and physiological factors play a significant role. Autophagy is reported to be induced in alveoli cells during exposure to high-altitude hypoxia. Moreover, autophagy is also a reported contributing factor for high-altitude hypoxia-induced lung injury⁸. The genetic makeup of each individual plays a critical role in their ability to respond to various scenarios, including

both resilience and vulnerability. The genetic diversity among individuals contributes to both advantageous and disadvantageous traits. Certain genetic variations can confer advantages in specific scenarios. To date, more than 40 *ATGs* have been identified, demonstrating a high degree of conservation across eukaryotes. Within this array of autophagy-related genes, *BECNI* emerged as a prominent candidate due to its pivotal role in orchestrating the autophagy process. *BECNI* is situated on the long arm of chromosome 17 at locus 17q21.31⁹. It encodes a 450-amino acid protein known as Beclin 1, characterised by three well-conserved domains: the Bcl-2-homology-3 (BH3) domain spanning amino acids 105 to 130, the coiled-coil domain (CCD) from amino acids 175 to 264, and the evolutionarily conserved domain (ECD) spanning amino acids 248 to 450⁹. Beclin 1 is a candidate autophagy protein that functions as a scaffold for the multiprotein assemblage during the autophagy process¹⁰. Beclin 1 interactome with multiple proteins modulates the autophagy initiation and maturation stages and thus Beclin 1 acts as central and crucial for the process. High levels of beclin 1 have been correlated with enhanced autophagy and beclin 1-mediated modulation of autophagy has been reported to be involved in high-altitude hypoxia^{8,10,11}.

Single nucleotide polymorphism (SNP) is variation at a single position in DNA sequence in more than 1 % population. SNPs affect the structure and function of genes depending upon the region in which these are present. Single nucleotide variations have been reported to play pivotal roles in various diseases. Gene regulatory region variants are important to study as these can directly influence gene expression. Thus, SNPs present in the

non-coding regions such as promoters and 3'untranslated region (UTR) are important to study. These regulatory region variants can affect the expression of a gene in various ways such as alteration of binding sites for transcription factors and altered microRNA (miRNA) binding sites¹²⁻¹⁴.

The promoter polymorphisms of the beclin 1 gene were reported to enhance *BECNI* expression and were related to the early onset of Machado-Joseph's disease¹⁵. Despite its significant importance, the Beclin 1 gene has not been fully elucidated to check the impact of SNPs present within it. Given the pivotal role of Beclin 1 in autophagy and its associations with various diseases, it becomes imperative to investigate the effects of SNPs situated within the regulatory domains of its gene. In this study, we employ a computational approach to elucidate the effects of SNPs located within the *BECNI* promoter and 3'UTR.

2. METHODOLOGY

The study is covered under three major sections: (1) Detection of promoter and 3'UTR SNPs, (2) Analysis of promoter region SNPs to check their effect on transcription factor binding sites, and (3) Analysis of 3'UTR SNPs to check their effect on alteration of miRNA target binding site.

2.1 Data Mining

The information about non-coding region variants of human *BECNI* was obtained from NCBI (<http://ncbi.nlm.nih.gov/>), dbSNP (<https://www.ncbi.nlm.nih.gov/snp/>) and ENSEMBL (<https://asia.ensembl.org/>) database. The flow chart is depicted in Fig. 1.

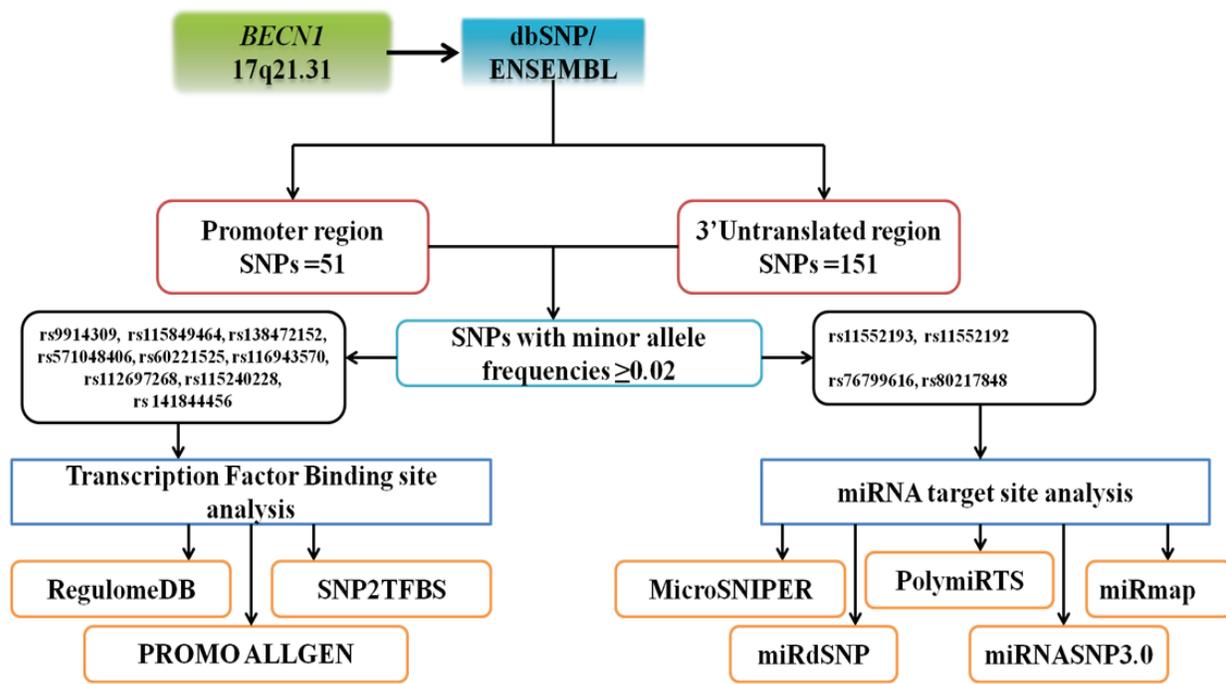


Figure 1. Flowchart depicting selection and analysis of regulatory region SNPs.

2.2 Prediction of Effect of *BECN1* Promoter Region SNPs RegulomeDB

(<https://regulomedb.org/regulome-search/>) database was used for the annotation of SNPs with known and predicted regulatory elements in the non-coding regions of the *beclin 1* gene¹⁶. **SNP2TFBS** (<http://ccg.vital-it.ch/snp2tfbs/>) was used to predict the effect of SNPs on the binding affinity of transcription factors¹⁷. dbSNP IDs of variants were provided as input. **PROMO** (<http://algggen.lsi.upc.es/>) was used to predict the effect of SNPs on transcription factor binding sites¹⁸. A string of nucleotides containing mutant and reference alleles was given as input. The default dissimilarity threshold used (the parameter that controls how similar a sequence must be to the matrix to be reported as hit) was 15 %.

2.3 Analysis of Potential 3'UTR Variants of *Beclin 1*

miRdSNP (<http://mirdsnp.ccr.buffalo.edu>) was used to predict the disease-associated SNPs and the miRNA binding site in the 3'UTR of *BECN1*¹⁹. PolymiRTS Database 3.0 (Polymorphism in microRNAs and their TargetSites database version 3.0) (<http://compbio.uthsc.edu/miRSNP>) was used to predict the variants altering miRNA binding site²⁰. miRNASNP-v3 (<http://bioinfo.life.hust.edu.cn/miRNASNP/>) tool provided a lot of information regarding SNPs in the miRNA target site. It also predicted the effect of SNPs on the binding of miRNAs²¹. rsID of variants or the gene symbol (*BECN1*) was given as an input in these three in silico tools. MicroSNiper (<http://cbdb.nimh.nih.gov/microsniper>) another tool that also predicts the effect of SNPs on miRNA binding sites was used to predict the creation or loss of the miRNA binding site²². The binding efficiency of miRNAs predicted by the above tools was calculated by miRmap (<http://cegg.unige.ch/mirmap>)²³. The name of the gene was provided in input and *beclin 1* targeting miRNAs predicted by the above-mentioned tools were searched one by one. Further, the site lost or gained was checked with the results provided by these servers.

3. RESULTS

3.1 Data Retrieval

SNPs of the non-coding region of *BECN1* were obtained from NCBI, dbSNP, and ENSEMBL. There were 51 upstream gene variants, 2372 intronic region variants, and 151 variants in 3'UTR. Since the minor allele frequencies (MAF) of most of the SNPs were not available, variants with a given MAF (≥ 0.02) and available literature were selected for analysis. In total, nine promoter region SNPs and four 3'UTR SNPs were analysed computationally. We have also performed in silico analysis of intronic region variants and genotyped one of the SNP (data submitted for publication). So, the current study is limited to promoter region variants and 3'UTR variants.

3.2 *Beclin 1* Promoter Region SNPs are Predicted to Alter Transcription Factor Binding Sites

RegulomeDB cross-references variant genomic coordinates with functionally relevant regions from assays like TF

ChIP-seq and DNase-seq and also considers transcription factor footprints and QTL data from the ENCODE database. This comprehensive approach helps assess a variant's functional significance by analysing its proximity to gene-regulating regions, aiding in understanding its impact on gene expression and phenotypic traits. The results obtained were given as ranks and scores. RegulomeDB probability score is determined using a random forest model and ranges from 0 to 1. A score of 1 signifies a higher likelihood of a variant being regulatory, while lower scores indicate decreasing probabilities of being a regulatory variant. The scoring scheme assesses the likelihood of a location or variant being functional based on supporting evidence. More available supporting data results in a higher rank, with 1 indicating a higher likelihood of functionality and 7 indicating a lower score, reflecting decreasing functional potential. Among the studied variants, rs138472152 and rs116943570 had RegulomeDB scores closest to 1 (0.82 and 0.81, respectively), depicting that these were most likely to be regulatory variants, whereas the rest of the variants had scores in the range of 0.6- 0.61 (Table 1). rs9914309 had a high rank and least probability score (comparative to the rest of the variants) of 0.55. SNP2TFBS computed the PWM (position weight matrix) score in both reference (hg19) and alternate human genome assemblies. The alternate genome assembly was generated by incorporating the alternate alleles of common genetic variants from 1000 genome projects. SNP2TFBS also provides high score and low score threshold values. The SNPs that change the PWM score above the threshold were retained.

Variant rs571048406 altered the binding sites of PAX4 and ERG2. However, this SNP had a score above the threshold for ERG2 only and hence it is considered to be retained. rs115849464 altered the binding sites for EGR1 and SP2 and the score was above the threshold in both cases, hence the SNP is considered to be retained. rs60221525 affected the binding sites for transcription factor EGR2, however, there was no score difference between reference and alternate genome. PROMO uses version 8.3 of TRANSFAC to construct specific binding site weight matrices for TFBS prediction. The variants having a low dissimilarity index show a high probability of binding. All of the studied variants were predicted to either create or remove the TFBS. The dissimilarity index indicating a high similarity score was less than 15 % in the case of all variants. The results are summarised in Table 1. Variants rs141844456, rs9914309, and rs112697268 were predicted to alter the maximum number of transcription factor binding sites.

3.3 *BECN1* 3'UTR Variants Alter the Binding Site for MiRNAs

miRdSNP provided the feature to search the position of an SNP with respect to the miRNA binding site in the 3'UTR. The position of conserved miRNAs and their target sites was obtained as an output. The variants rs11552192 (G>A) and rs11552193 (T>A) were shown

Table 1. List of the promoter variants that alter the binding site of transcription factor

S. No.	Variant id	RegulomeDB		SNP2TFBS		PROMO		
		Rank	Score	Match	Score difference	Transcription factor	Loss/ gain of site	Dissimilarity index (%)
1.	rs571048406	2b	0.61	EGR2 PAX4	-88 -45	Egr-3 RXR-alpha	Gain of site	11.37 5.27
2.	rs115849464	2b	0.61	EGR1,SP2	>259 -145	p53 PAX5	Loss of site	6.19 9.55
3.	rs115240228	4	0.60	-	-	E2F	Loss of site	13.89
4.	rs116943570	2b	0.81	-	-	MAZ	Gain of site	14.13
5.	rs138472152	2b	0.82	-	-	Ik-1	Gain of site	11.87
6.	rs141844456	4	0.60	-	-	STAT4	Loss of site	4.4
						cETS-1		0.26
						Elk-1		4.2
						HNF-1C		9.58
						HNF-1B		10.21
cETS-2	5.16							
7.	rs9914309	1f	0.55	-	-	TFII-I	Gain of site	1.82
						AP-2alphaA	Loss of site	3.97
						SRY	Gain of site	13.35
TCF-4E	12.60							
8.	rs60221525	2b	0.61	EGR2	0	LEF-1	Gain of site	9.94
						MAZ		14.13
9.	rs112697268	4	0.60	-	-	PAX5	Loss of sites	1.54
						p53		1.27
						GR-alpha		Gain of sites
AP2-alphaA	2.55							
						T3Rbeta1		4.48

but none of the variants was located within the most preferential binding site of highly conserved miRNA that is within the site of miR-30a-5p, miR124-3p-2/506-3p, miR142-5p, and miR-17-5p/93. PolymiRTS Database 3.0 predicted the effect of variants on the miRNA binding site (Table 2). The results were given under two function classes – ‘D’ and ‘C’. Function class ‘D’ refers to the disruption of the miRNA binding site, whereas ‘C’ refers to the creation of a new miRNA site, whereas, the context score predicts the likelihood of the outcome (Table 2). Variant rs11552193 led to the disruption of 2 miRNA binding sites and the creation of 3 new sites. Variant rs11552192 created a new site for miRNA hsa-miR-590-3p, whereas, rs76799616 created a new site for 2 miRNAs and rs80217848 resulted in the loss of site for 2 miRNAs (Table 2). Disruption of the miRNA binding site leads to the modulation of expression whereas the creation of a new site has an inhibitory role in gene expression²⁰. Context score change predicted the

likelihood of the event of creation or disruption of the site. Its more negative value suggests higher chances of site creation or disruption. The miRNASNP-v3 predicted the effect of SNP on the miRNA target binding site as loss or gain of the site (Table 3). Along with that, free energy change values were given in Table 3. rs76799616 only caused the gain of miRNA binding sites whereas the rest three variants resulted in both gain and loss of sites (Table 3). Moreover, most of the sites predicted by miRNASNP-v3 were also predicted by PolymiRTS Database and MicroSNiper. miRmap computed the highest seed match sequence probability for miRNAs hsa-miR-4684-5p and hsa-miR-32-3p with scores of 50 and above. Both of the sites were however predicted to be lost due to variants rs11552192 and rs11552193 (Table 3).

4. DISCUSSION

The investigation of SNPs within non-coding regions carries substantial scientific significance, as it

Table 2. 3'UTR SNPs alter the miRNA binding site predicted by PolymiRTS Database 3.0

S. No.	Variant id	Variation	PolymiRTS Database 3.0			
1.	rs11552193	G>A	miR ID	miR site	Function class	Context ⁺ score change
			hsa-miR-3675-3p	gtgtTAGAGATat	Disruption of site	-0.288
			hsa-miR-520f-5p	gtgTTAGAGAtat	Disruption of site	-0.093
			hsa-miR-302b-5p	gtGTTAAAGAtat	Creation of site	-0.196
			hsa-miR-302c-5p	gTGTTAAAgatat	Creation of site	-0.003
			hsa-miR-302d-5p	gtGTTAAAGAtat	Creation of site	-0.0207
2.	rs11552192	T>A	hsa-miR-590-3p	ctaatTAAAATTt	Creation of site	-0.015
			hsa-miR-3140-5p	aaATTCAGGtat	Creation of site	-0.317
3.	rs76799616	G>A	hsa-miR-4680-3p	aAATTCAGgtaat	Creation of site	-0.195
4.	rs80217848	A>G	hsa-miR-320a-d	catctaAGCTTTA	Disruption of site	-0.035
			hsa-miR-4429	catctaAGCTTTA	Disruption of site	-0.103

Table 3. Effect of 3'UTS SNPs on miRNA binding site predicted by miRNASNP-v3

S. No.	Variant	Variation	Gain/Loss of miRsite	Effect	
				miRNA	dG binding (Kcal/mol)
1.	rs11552193	G>A	Gain	hsa-miR-302b-5p	-15.07
			Gain	hsa-miR-302c-5p	-10.04
			Gain	hsa-miR-302d-5p	-17.88
			Gain	hsa-miR-552-5p	-8.62
			Loss	hsa-miR-3765-3p	-9.15
			Loss	hsa-miR-520f-5p	-10.63
			Loss	hsa-miR-4684-5p	-9.28
2.	rs11552192	T>A	Gain	hsa-miR-590-3p	-3.33
			Loss	hsa-miR-32-3p	-3.36
3.	rs76799616	G>A	Gain	hsa-miR-4680-3p	-7.86
			Gain	hsa-miR-183-3p	-4.67
			Gain	hsa-miR-4427	-8.55
			Gain	hsa-miR-3140-5p	-11.86
			Gain	hsa-miR-4452	-5.13
			Gain	hsa-miR-5187-3p	-6.69
4.	rs80217848	A>G	Gain	hsa-miR-7109-3p	-11.32
			Gain	hsa-miR-3135a	-10.05
			Loss	hsa-miR-320a-d	-8.66
			Loss	hsa-miR-4429	-7.12

elucidates essential regulatory elements governing gene expression, exerting influence over a broad spectrum of biological phenomena. In the current study, we have shortlisted and predicted the possible impact of SNPs present in *BECN1* non-coding regions computationally. Analysis of promoter region variant predicted the nine potential variants altering *BECN1* transcription factor binding sites which could eventually lead to altered expression of *beclin 1*. Earlier, Kazachkova et al., (2017) studied *BECN1* promoter region variants rs60221525 and rs116943570 and checked their correlation with Machado Joseph's disease. Apart from their finding of the alteration of TFBS due to these variants, they further checked that the variants affected the expression of *BECN1* which is correlated to early onset of the disease¹⁵. The *BECN1* gene expression is under the regulation of various transcription factors and levels of expression are directly correlated with autophagy initiation³. The effect of SNPs on TFBS predicted by these in silico servers could be further experimentally validated by checking the gene expression. The miRNAs modulate gene expression by binding to 3'UTR^{24,25}. Most of the miRNAs downregulate gene expressions, yet few reported miRNAs can act as positive regulators of gene expression²⁵. The computational examination of the 3'UTR of *Beclin 1* has revealed that the investigated SNPs either introduce or eliminate binding sites for miRNAs. In mammalian genes, the seed sequence is widely recognised as the paramount feature for miRNA-mediated target recognition. Moreover, 3'UTR SNPs were predicted to eliminate the sites for two miRNAs with the highest seed match probability. SNPs that alter miRNA sites have been reported to be prognostic and predictive cancer biomarkers²⁶. So, these SNPs could be studied further for functional validation and could be correlated with a particular trait such as stress response. Expression of *Beclin 1* is directly correlated with induction of autophagy²⁷. Altered autophagy due to altered expression of *beclin 1* contributes as a major factor in the case of various diseases _ENREF_28²⁸. Furthermore, a recent study has reported that autophagy plays a significant role as a mechanism in the development of lung injury induced by high-altitude hypoxia⁸. Additionally, *Beclin 1* expression was seen to be enhanced in the tissues exposed to hypoxia. Thus underlying factors that enhance the expression of candidate genes such as *beclin 1* under various conditions need to be thoroughly investigated. SNPs can confer protection or susceptibility towards physiological traits such as stress responses.

SNPs within the human genome are a rich source of genetic diversity, impacting our understanding of traits, diseases, and personalised medicine. SNPs enhance genetic comprehension and hold the potential for improved healthcare. In silico analysis is vital for initial SNP analysis, facilitating the identification of functionally significant variants, saving time and resources, and offering insights into the genetic foundations of traits and diseases.

This is a preliminary study that predicts regulatory region *beclin 1* SNPs that affect the transcription factor and miRNA binding sites which can be a reason for the altered expression of *beclin 1* and this could be further studied and correlated with the susceptibility of individuals towards environmental stress such as hypoxia. Thus, comprehending an individual's genetic predispositions can provide valuable insights for tailoring personalised healthcare strategies and interventions to optimize well-being while mitigating potential risks. Our study is limited to preliminary computational analysis and lacks experimental validation. However, this study forms a basis for further functional studies such as the study of the effect of only these shortlisted SNPs with predicted impact on *beclin 1* expression. Furthermore, most importantly association studies on these SNPs could be done to check the genetic predisposition of individuals such as troops posted or likely to get posted to high altitudes to ensure their physiologic fitness to face such extreme environmental conditions. In conclusion, this computational approach has provided significant results of analysis of the regulatory region of *BECN1* that could further help find out the underlying mechanism of its malfunction in various diseases.

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