

# Translation of Indian Sign Language to Text-A Comprehensive Review

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**Abstract:** Deaf and mute persons across the world uses gestures, non-manual features to interact with fellow persons. This way of communication is called Gesture language or Sign language. Gesture languages are local in nature because of their dependency on geographical area, syntax, pragmatics, and other attributes. The focus of this paper is to present a comprehensive review of conventional as well as contemporary Indian sign language translation system. The process of literature review has been carried out in accordance with Preferred Reporting Items for Systematic reviews and Meta Analysis (PRISMA) guidelines by searching in Scopus, google scholar, Science direct and Lensorg databases. Different articles were included between the years 2010 to 2023 for the purpose of literature review. The study was based on four themes-dataset, technique, result and previous literature reviews. This is the first detailed review conducted in the field of Indian sign language translation system which solely analyses literature related to ISL as per author's knowledge. The findings of this research article may contribute to gain insights and form a blueprint for future areas in the arena of Indian Sign Language translation/recognition system.

**Keywords:** Comprehensive review, ISL Translation, ISL Recognition, Indian Sign Language, PRISMA

## 1. Introduction

Nonverbal communication encompassing body language, gestures, facial emotions is a vital aspect of human interaction. However, people with special needs are solely reliant on this form of communication. Deaf and dumb persons across the world uses gestures, non-manual features to interact with fellow persons. This way of communication is called Gesture language or Sign language[1]. These languages have evolved over the years because of their natural existence. Due to lack of resources for these especially abled persons, they used sign language to communicate with their families. With technological advances, to aid these persons, schools, medical facilities came into existence. Gesture languages are local in nature because of their dependency on geographical area, syntax, pragmatics, and other attributes[2]. Indian Sign Language (ISL) came into existence in 2018, after a long battle by deaf and dumb community[3]. There are many popular sign languages used in India apart from ISL such as Bangla Sign Language (BSL), Tamil Sign Language (TSL), Panjabi Sign Language (PSL) and Malayalam Sign Language (MSL) etc. Currently, Sign language translation is most popular domain with the potential to provide automatic and effective communication tool for hearing disabled persons[4]. It translates the given input sign language gesture into corresponding text. Promptness and exactness are the

important parameters for the decision makers to determine the efficacy of the proposed system[5].

Machine learning is an offshoot of artificial intelligence which aims to simulate human intelligence in machines using various algorithms. However, recently advanced form of machine learning is deep learning which relies on artificial neural network to emulate human neurons for image processing tasks. Machine Learning (ML) and Deep learning (DL) paradigms can process huge amount of data in a reasonable time limit and build an efficient translation system. Consequently, ML and DL practices are getting immensely popular in the discipline of sign language processing [6].

Although a lot of literature reviews has already been conducted in the field of sign language but dearth of an exhaustive literature review in the field of ISL was one of the major motivation factors for this research article using standard Preferred Reporting Items for Systematic reviews and Meta Analysis (PRISMA) guidelines [7]. The contributions of this research paper are as follows.

- In this paper, we have studied the work done in the field of Indian Sign Language Translation for the vicennial period along with their shortcomings.
- Various datasets available in the domain of Indian Sign Language Translation System (ISLTS) has been explored with the focus on available open access dataset.
- Different levels of translation such as Alphabet level, word level and sentence level has been discussed.
- This paper examines current trends in ISLTS and

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provide suggestions to researchers for future works.

The rest of the research paper is organized as follows- Section II describes methodology followed by the authors to conduct this literature review. Section III describes the results followed by conclusion in future scope in Section IV.

## 2. Materials and Methods

We carried out a PRISMA comprehensive review to analyze technological advancements in the field of translation of Indian sign language into text. The research questions were formulated in the first step to initiate the process of conducting literature review.

- RQ1- What is the focus of study in previous literature review in the field of ISLTS?
- RQ2-What are various types of datasets available for researchers in the field of ISLTS?
- RQ3- What are the number of research paper published per year on ISLT/RS?
- RQ4- What are existing techniques for translating ISL gestures and their performance?

### 2.1 Search Query

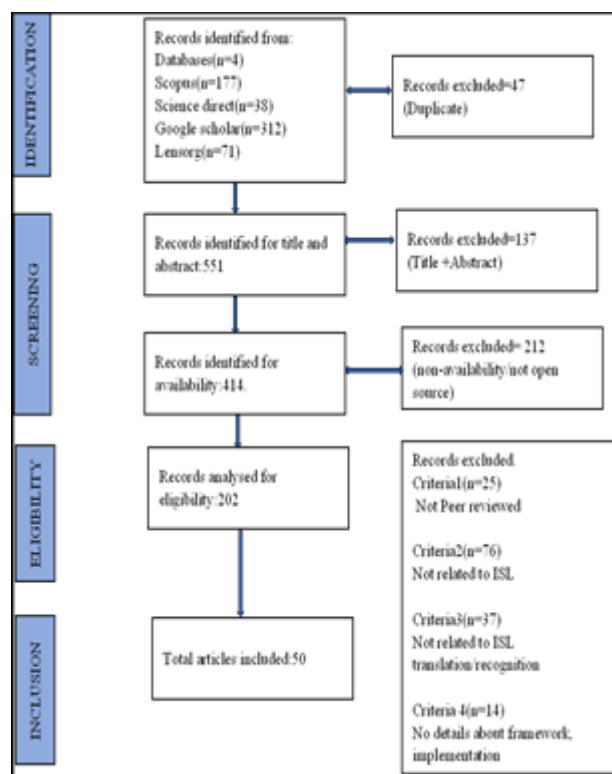
Different research papers related to ISLTS/ Indian Sign Language Recognition System (ISLRS) were searched on four popular research databases such as Scopus, Google Scholar, Science Direct and Lensorg. The principal objective of this literature survey is to examine translation/recognition attempts made in the ISL sector. Open Access articles in English language has been selected for this review process from year 2010 to 2023 based on the search queries mentioned in Table 1.

**Table 1.** Search Query

Name of Dataset	Query
Scopus	TITLE-ABS-KEY ( "INDIAN SIGN LANGUAGE TRANSLATION" OR "ISL TRANSLATION" OR "ISL RECOGNITION" OR "INDIAN SIGN LANGUAGE RECOGNITION" ) AND ( LIMIT-TO ( EXACTKEYWORD , "Indian Sign Languages" ) OR LIMIT-TO ( EXACTKEYWORD , "Indian Sign Language" ) )
Science Direct	"INDIAN SIGN LANGUAGE TRANSLATION" OR "ISL TRANSLATION" OR "ISL RECOGNITION" OR "INDIAN
LensOrg	TRANSLATION" OR "ISL TRANSLATION" OR "ISL RECOGNITION" OR "INDIAN
Google Scholar	RECOGNITION" OR "INDIAN

Name of Dataset	Query
	SIGN LANGUAGE RECOGNITION"

The workflow of the literature review process has been shown using **Error! Reference source not found..** A total of 598 research articles were identified for the purpose of literature review using above mentioned queries from four major research databases. In the next stage, 47 articles were excluded from the study because of redundancy. Upon preliminary literature investigation, every research article’s title and abstract were examined manually and then 414 pertinent papers were selected for further assessment criteria. Subsequently, 202 research articles were selected based on the criteria of availability of research paper or whether it is open access. Four eligibility criteria were adopted for this literature survey in the next phase.



**Fig. 1.** Flowchart of comprehensive review process using PRISMA guidelines

- The article should be peer reviewed.
- The article should be related to Indian Sign Language.
- The theme of the article should be related to recognition/translation of ISL to text.
- The details about framework/implementation should have been mentioned in the research article.

After stringent eligibility criteria more than 50 articles has been selected for this literature review.

### 3. Results and Discussion

In this section, we will try to answer all the research questions on the basis of literature.

#### 3.1 RQ1- What is the focus of study in previous literature review in the field of ISLTS?

To analyze previous literature surveys in this field and their focus areas on which review has been performed a list of review articles has been crafted in the field of ISLRS/ISLTS process. Table 2. shows various research articles published along with their year and focus of review from the period of 2010 to 2023. It has been observed that no literature review article has been published in the field of ISLTS/ISLRS using PRISMA guidelines as per the author's knowledge. It has been observed that most of the literature review articles [8]–[12] included less than 10 research articles for their analysis due to lack of standard research work in the domain of ISL. [13] contemplated 29 research articles related to dynamic recognition and compared different methodologies. However, [9], [13] reviewed various methodologies and [12], [14] discussed various feature extraction techniques in ISLTS/ISLRS. [15] examined different articles related to dataset acquisition techniques and concluded that area of non-manual features is yet to be explored in case of ISL. [16] studied few ISLR articles along with other sign languages to conclude there are limited work in alphanumeric recognition, dynamic sign recognition. It has been analyzed that 90% of review papers in ISL translation/recognition process considered less than 20–25.

**Table 2.** Prior literature reviews in ISL

Ref	Type	Year	Criteria for Review
[8]	J	2015	Gesture set and technique
[9]	J	2023	Methodology
[10]	J	2013	Input, segmentation, Feature vector, classification, recognition rate, platform
[11]	C	2015	Challenges
[12]	C	2021	Feature extraction
[13]	C	2021	Dynamic ISLRS with focus on methodology
[14]	C	2022	Feature Extraction
[15]	C	2022	Dataset acquisition techniques
[16]	C	2019	Input, dataset, segmentation, method, number of gestures, output, limitation, recognition percentage

#### 3.2 RQ2- What are various types of datasets available for researchers in the field of ISLTS?

A standard well annotated dataset is very important for any sign language processing system. In case of ISLTS, lack of standard, open access datasets are few of the major challenges in translation/recognition process. Table 3. describes list of open access datasets available in the domain of ISLTS along with their characteristics. We have also included some dataset with limited access.

**Table 3.** List of open access datasets available in ISL

Ref	Year	Dataset	Type		
			Alphanumeric	Word	Sentence
[17]	2010	ISL		22	
[18]	2020	Include	-	15W	
[19]	2021	ISL-CSLTR	-	1036	100
[20]	2021	ISLAN	24 A	-	-
[21]	2021	Emergency	-	8W	-
[22]	2021	INSIGNVID	-	55W	15S
[23]	2022	IISL2020	-	11W	-

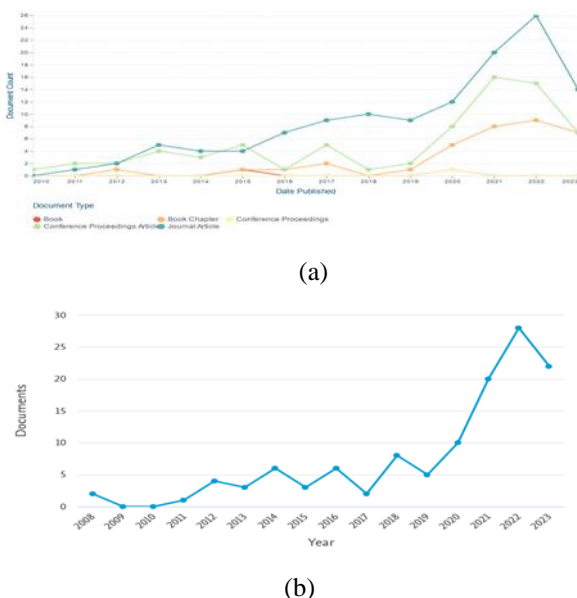
In 2010, [17], created RGB video dataset of 23 different ISL word gestures at 30 frames per second (fps) under various background and lightning conditions. The access of this dataset has been restricted and only given to educational institutions based on agreement through their website by the creators. In 2020, [18] presented Indian lexicon sign language dataset (INCLUDE) with the help of 7 experienced signers. This word level dataset consists of 263 classes of 15 categories, 4287 videos with 1920x1080 resolution and 25 fps. A subset of the above dataset having 50 signs across 15 categories was also proposed with same specifications called INCLUDE 50. Both the datasets include 15 words in total. In 2021, several other researchers came up with their own ISL datasets available freely like [19] Elakkiya et al developed first sentence level Indian sign language dataset for continuous Sign language translation and recognition i.e., ISL-CSLTR. The dataset contained 700 videos of 100 sentences made up with the help of 7 signers. Secondly, another ISL dataset for Alphanumeric (ISLAN) signs was developed by [20] comprising of 350 unique sign images and 12 unique videos compassing 24 alphabets of English language (except J, Z) and numbers totaling to 700 images and 24 videos by 6 signers. Another sign language dataset for emergency domain has been developed by [21]. It included 824 videos of 8 words by 12 males and 14 females. Indian Sign Language Video (INSIGNVID) dataset was developed by [22] for efficient recognition of 55 words of ISL. The dataset was created by 4 right-handed

persons and consists of videos with 30fps, 1920\*1088 resolution and common background conditions. In 2022, Kothadiya et al [23] proposed a permission based Isolated ISL dataset (IISL2020) made up of 11 words from 16 persons and 1100 videos and average 28fps.

### 3.3 RQ3- What are the number of research paper published per year on ISLT/RS?

**Error! Reference source not found..** elucidates an overview of ISLTS studies that are published annually from Lensorg and Scopus source in (a) and (b) parts. It has been observed from the figure that highest articles are published in the year 2023 till date i.e., maximum of 26 journal articles has already been published in the year 2023. The topic of ISLTS/ISLRS has garnered a lot of research attention in the last few years however, the work done in recent two years outshines the previous works quantitatively and qualitatively.

Three different categories of articles i.e., book chapter, conference articles and journal articles are contemplated for this research article as represented with figure 3(a) and 3(b). It has been observed that majority of articles published in the field of ISL recognition/translation domain are from journals i.e., 124 out of total 239 are research articles published in journals followed by numbers of articles in conference. This data has been taken from Lensorg website [24] with the constraint to include only selected articles related to the domain of ISLTS/ISLRS from the period 2010 to 2023.



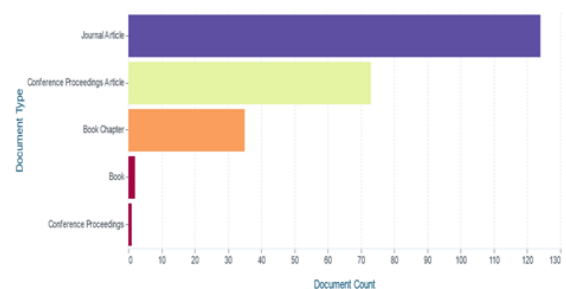
**Fig. 2.** Example of a Publications per year (a)LensOrg (b) Scopus

RQ4- What are existing techniques for translating ISL gestures and their performance?

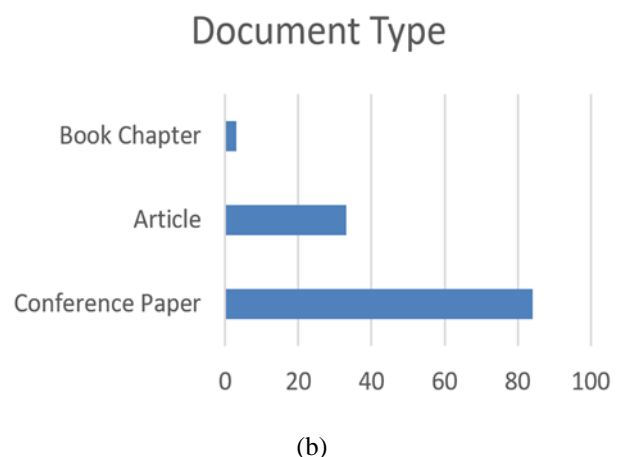
Machine learning has garnered a lot of attention in the field of sign language processing over the last few years.

Majumdar et al [25] in 2011 proposed Indian sign language recognition system with YCbCr segmentation, wavelet packet decomposition, principal curvature-based region as feature extraction, dynamic time warping (DTW) to classify alphabets with an accuracy of 91.3%. [26] concluded that Multi SVM classifier can classify static ISL gestures with recognition rate of 92.6 on a self-made dataset.[27]

proposed a ISL recognition system to classify 24 alphabet level gestures with 97 recognition accuracy using novel Eigen value weighted Euclidean distance. [28] proposed a framework for recognition of two-handed gestures of ISL by employing HOG feature extraction method and four popular pretrained models ALEXNET, VGG-16, VGG-19 and GoogleNet. The model attained highest accuracy of 99.11% with ALEXNET and VGG-19 pretrained transfer learning models to classify alphabets of ISL. [29] proposed model for recognizing alphabets of ISL using extreme learning with an average accuracy of 80.76% on self-made dataset. [30] developed a ISLRS for alphabets using CNN with



(a)



(b)

**Fig. 3.** Publication types (a) LensOrg (b) Scopus

### 3.4 RQ4- What are existing techniques for translating ISL gestures and their performance?

Machine learning has garnered a lot of attention in the field of sign language processing over the last few years. Majumdar et al [25] in 2011 proposed Indian sign language recognition system with YCbCr segmentation, wavelet packet decomposition, principal curvature-based region as

feature extraction, dynamic time warping (DTW) to classify alphabets with an accuracy of 91.3%. [26] concluded that Multi SVM classifier can classify static ISL gestures with recognition rate of 92.6 on a self-made dataset.[27] proposed a ISL recognition system to classify 24 alphabet level gestures with 97 recognition accuracy using novel Eigen value weighted Euclidean distance. [28] proposed a framework for recognition of two-handed gestures of ISL by employing HOG feature extraction method and four popular pretrained models ALEXNET, VGG-16, VGG-19 and GoogleNet. The model attained highest accuracy of 99.11% with ALEXNET and VGG-19 pretrained transfer learning models to classify alphabets of ISL. [29] proposed model for recognizing alphabets of ISL using extreme learning with an average accuracy of 80.76% on self-made dataset. [30] developed a ISLRS for alphabets using CNN with diffGrad optimizer and stochastic pooling to achieve validation accuracy of 99.64%. [31] proposed a framework for recognition of alphabets of ISL using correlation coefficient feature extraction and neurofuzzy algorithm as classifier to achieve an average accuracy of 92.3%. [32] proposed transfer learning based recognition of ISL alphabets with an accuracy of 95%. The VGG16 pretrained model consists of 13 convolution layers, average, max pooling, dropout layer for controlling overfitting, Adam optimizer and softmax as classifier layer.

In numeric ISLTS, [33] in 2014, proposed ISL numeric digit (0-9) recognition system on a self-made ISL dataset using KNN classifier and an accuracy of 97.1%. However, [34] proposed Kinect sensor based ISLRS using scale, rotation, and background lightning invariant ORB feature extraction method and KNN machine learning algorithm to classify (0-9) digits of ISL on a self-made dataset with an accuracy of 93.26% outperforming standard feature extraction techniques like SIFT and SURF.

In the domain of alphanumeric level recognition, [35] proposed ISLRS framework using fingertip algorithm and PCA to obtain 94% accuracy. In 2013, [36] used Fourier descriptors, distance transform and artificial neural network with four layers to classify 36 alphanumeric gestures of ISL with an average accuracy of 91.11%. [37] Geetha et al suggested alphanumeric ISL sign recognition system with B-spline approximation and SVM classification algorithm. [38] proposed novel fusion descriptor for classification of ISL numeric signs with Nearest Mean classifier and an accuracy of 100%. The novel fusion descriptor comprises of two contour (Boundary, Fourier descriptor) and one region based (7Hu) descriptors. [39] classified gestures of ISL using SVM machine learning algorithm. [40], [41], [42] used Kinect sensor to classify gestures at alphanumeric and word level along with popular classification algorithms such as PCA, SVM to attain remarkable accuracies. [43] suggested translation of word level ISL gestures by extensive training of humanoid robot

HOAP-2 along with direction histogram feature extraction, Euclidean distance metric has been used to attain an average accuracy of 90%. [23] Kothadiya et al. in 2022 classified 11 words of ISL using sequential combination of LSTM and GRU with accuracy of 97% on their dataset IISL2020. [44] developed a model for classifying 24 dynamic word gestures of ISL using novel dynamic time warping recognition technique along with accuracy of around 90%. 20 different gestures were classified by [45] using 3D CNN and attaining 88% validation accuracy in 100 epochs. The model

comprises of 3 convolution layers, max pooling, dropout and softmax activation function. [46] Subramaniam suggested integrated model of Media pipe with optimized GRU model for recognition of 13 ISL gestures to attain average accuracy of 95%. The proposed system has been compared with RNN, LSTM, standard GRU, BiGRU, and BiLSTM models.

Hybrid ISLT paradigm comprising of combination of word and alphanumeric and sentence level. In this, [22] suggested transfer learning approach using MobileNetV2 to transcribe clips of ISL into English language. The proposed system was analyzed using other pretrained models such as MobileNet, VGG16 and ResNet50 using 25 epochs with 9 trainable layers to attain testing accuracy of 93.89%. Although proposed system achieves better accuracy but time to train the system was comparative high i.e., more than 12 hours.

[47] classified alphanumeric, word gestures of ISL using Fourier descriptor feature extractor and distance metric to attain overall accuracy of 92.91%. The proposed system [48] recognized 26 alphabets, 10 numbers and 10 distinct phrases on self-made skinpixel segmentation, Moment based feature extraction and SVM algorithm to classify dynamic gestures. The system obtained an accuracy of 97.5% recognition rate in classifying 4 signs (3 alphabet and one word). [49] presents a signer independent communication model for real time using YCbCr segmentation, Zernike moments feature vector and SVM as classifier. [50] Deshpande et al. classified 56 signs real time using CNN into text and audio with 98% accuracy with the constraint of plain background. The proposed model had 5 convolution layers, ReLU activation function, max pooling, dropout layer, single valued stride function and softmax layer to classify different signs. [51] recognized gestures of 7 days of week using Kinect sensor and random forest classifier algorithm to give an accuracy of 74.29% with focus on low cost and maximum efficiency. [52] Nareshkumar attained an accuracy of 98.77% in translating alphanumeric gestures of ISL and American Sign language using novel pretrained model for mobile MobileNetV2 consisting of pointwise convolution layers, separable depthwise convolution, ReLU activation function, swish

activation function, batch normalization, dropout layer and softmax as final classifier layer. [53] developed ISLRS using HSV segmentation, PCA with OH 18 and 36 bins feature extraction mechanism to classify 10 ISL sentences having 2,3 or 4 gestures using six different distance metrics. Euclidean distance topped the performance chart with 93% recognition rate (RR) on 36 bins of orientation histogram. [54] proposed a lightweight framework for translating sentence level ISL gestures into text and audio, LiST, and used pretrained model InceptionV3, two layered LSTM architecture on open access dataset with a translation accuracy of 91.2%. [55] proposed a framework for translation of 10 signs of ISL using HMM and DWT to

achieve lowest accuracy of 80 and highest accuracy of 100%. [56] proposed a Leap motion sensor based ISLTS for 35 words and 942 sentences using four gated cell LSTM with 2 dimensional CNN to attain average accuracy of 89.5% and 72.3% respectively.[57] proposed gesture recognition mechanism for 42 signs of ISL using KNN and SVM machine learning classification techniques. HSV, Otsu thresholding for segmentation and novel MFCC feature extraction method has been used along with wavelet descriptor to translate 42 static and dynamic gestures of ISL. The authors concluded that SVM with MFCC feature extraction mechanism

**Table 4.** Work done in the domain of ISLTS

<i>Ref</i>	<i>Year</i>	<i>Type</i>	<i>Specifications</i>	<i>Features</i>	<i>Results</i>
[25]	2011		26Alphabet	Support Vector Machine (SVM), K-Nearest Neighbour (KNN) and Dynamic Time Warping (DTW)	91.3 accuracy
[26]	2012		26Alphabet	Multi SVM classifier	92.6 Recognition Accuracy
[27]	2013	ALPHABET	Alphabet (24)	Eigen value weighted Euclidean distance	97 Recognition Rate
[28]	2022		Alphabet	Histogram Oriented Gradient (HOG), AlexNet	99.11 Accuracy
[29]	2020		-	Extreme learning	80.76 Accuracy
[30]	2022		26 Alphabet	Convolution Neural Network (CNN) with diffGrad optimiser	99.64
[31]	2019		26 Alphabet	Neurofuzzy algorithm with correlation coefficient feature extractor	92.3
[32]	2022		26 Alphabet	VGG16	95
[33]	2014	NUMBER	10 Numbers	Neural network with KNN	97.1
[34]	2020		10 Numbers	Kinect and Bag of visual words with ORB, KNN	93.26
[35]	2012		36Alphanumeric	Principal Component Analysis (PCA)	94 Recognition Accuracy
[36]	2013	ALPHANUMERIC	36Alphanumeric	4-layer Artificial Neural Network (ANN)	91.11 Recognition Rate
[37]	2012		26Alphabet, 6Numbers	B-spline approximation	-
[38]	2016		36Alphanumeric	Novel fusion descriptor,	99.61
[39]	2022		Alphanumeric	Bag of Visual Words (BOVW), Speeded Up Robust Features (SURF), SVM, CNN	99.64 Accuracy

[40]	2020	Alphanumeric	Kinect with SURF, HOG, Local binary pattern	Average Accuracy-71.85
[23]	2022	11Words	Long short-Term Memory (LSTM), Gated	97Accuracy
[41]	2015	37Words	Kinect with SVM	86.16% Validation Accuracy
[42]	2013	10Words	PCA with ALI, Microsoft Kinect (25 key	Best-100A, Average-40, Worst-25
[43]	2010	22Words	Euclidean distance, KNN with 36 bins	Lowest-48.42, Highest-100Accuracy
[44]	2016	24Words	DTW	90Accuracy
[45]	2021	20Words	3D-CNN	88.24Average Accuracy
[46]	2022	13Words	MOPGRU with ELU activation and softsign	99.92Accuracy, 0.21 Loss
[22]	2021	55Word, 15Sentences	MobileNetV2 pretrained model	93 Recognition Rate
[47]	2015	10Signs	Fourier Descriptor with Euclidean distance	Lowest-85, Highest-97Accuracy
[48]	2016	01Word, Alphabet	3SVM, Kinect sensor	97.5 Recognition Accuracy
[49]	2022	26Alphabet, 11Word	Co-articulation, Zernike moment, SVM	Alphabets-91A, W-89A
[50]	2023	36 Alphanumeric, 20Word	Region of Interest, CNN	98Accuracy
[51]	2022	7 Word	Kinect V2 sensor, Random Forest	74.28 Accuracy
[52]	2023	26Alphabet, 3Word	Transfer learning, Modified MobileNet V2	98.77 Accuracy
[53]	2015	10Sentences	PCA with OH	Lowest-85 , Highest -93 Accuracy
[54]	2023	15Sign	Inception v3 CNN with LSTM	95.90 Accuracy
[55]	2015	10Sign	DWT, Hidden Markov Model (HMM)	Lowest-80, Highest-100 Accuracy
[56]	2019	942Sentences, 35Word	CNN with LSTM and Leap motion sensor	Sentences-72.3Avg. Accuracy, Words-89.5 Avg. accuracy
[57]	2017	42Words	KNN, SVM with MFCC feature extraction	97 Accuracy
[58]	2019	80Word, 50Sentences	Fuzzy clustering algorithm	75 Accuracy
[59]	2018	33Alphanumeric, 12 signs	HMM, KNN	99.7 Static Sign

WORD

HYBRID



[60]	2012	80Words Sentences	andPCA feature vector, Fuzzy inference system	96 Accuracy
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classified ISL gestures with better accuracy than KNN. [58] proposed Fuzzy c-means clustering algorithm for classifying 80 words and 50 sentences of ISL with an average accuracy of 75%. [59] proposed ISLTS for 45 sign (alphanumeric and word) using skin color segmentation, Hidden Markov Model, K-nearest neighbor recognition algorithm with an accuracy of 99.7% for static signs and 97.23% for dynamic signs. [60] proposed video gesture recognition of ISL using Gaussian filter, Canny edge detector, Fourier descriptor and Sugeno fuzzy inference system to attain higher accuracy of 100 and lowest accuracy of 60 among total signs. Although there are various researchers who have been working in the domain of Indian sign language translation/recognition to develop optimal framework but there is tradeoff between accuracy and time. The system is affected by so many parameters discussed in next section.

#### 4 Conclusion and Future Scope

In this paper, we have conducted comprehensive literature review on Indian Sign Language Translation/Recognition System using PRISMA guidelines. After rigorous screening, more than 50 papers were selected for this review from four major research databases- Scopus, Google Scholar, Science Direct and LensOrg. There were four main criteria on which this survey was conducted- previous work done, datasets available, number of research articles published per year and summary of important work done. It has been concluded a lot of work has already been done in sign language processing systems, but ISLTS are still lagging in a lot of aspects.

- Lack of well annotated standard open access datasets
- Alphanumeric recognition
- Two-way communication system
- Domain specific translation system
- Lack of quality review papers
- Sensor based devices gives better accuracy but are not comfortable

As there are many challenges but exploitation of new emerging machine learning algorithms is need of the hour in ISLTS as compared to other sign language processing system. We hope that this research paper will help other future researchers in the field of ISL.

#### Author contributions

**Seema Sabharwal:** Conceptualization, Methodology, Software, Field study, Data curation, Writing-Original draft

preparation, Software, Validation, Field study, Visualization **Priti Singla:** Investigation, Writing-Reviewing and Editing.

#### Conflicts of interest

The authors declare no conflicts of interest.

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