### TOUCHLESS TOUCHSCREEN TECHNOLOGY

#### **Abstract:**

Touchless touchscreen technology has undergone a transformative evolution, revolutionizing our interaction with screens and presenting numerous benefits. Beyond the evident improvement in hygiene, touchless screens provide users with a versatile and natural interaction experience, recognizing a diverse range of gestures for intuitive and dynamic interactions. These screens find application in various sectors, from self-service kiosks to gaming and medical devices. Existing technologies like infrared sensors, cameras, and radar have paved the way for touchless touchscreens, with infrared sensors being cost-effective but having accuracy limitations in low-light conditions. Cameras, offering higher accuracy in different lighting scenarios, may entail a higher installation cost, while radar, known for precision in low-light conditions, adds capability but can be more expensive and demanding on processing power. Future developments may introduce sophisticated features, such as integration with haptic feedback mechanisms to simulate tactile sensations, advancing the user experience. Progress in AI and machine learning could enable touchless screens to comprehend and respond to user emotions, personalizing interactions further. Enhanced security measures, including biometric recognition, can heighten security for sensitive applications. The integration of 3D scanning and augmented reality capabilities may open new avenues for immersive experiences. However, these advancements bring ethical considerations, including data privacy, consent, and potential biases in AI algorithms. Maintaining a balance between convenience and meaningful human interaction is crucial, emphasizing user education on responsible usage and digital well-being. Touchless touchscreen technology represents a multifaceted approach to human-computer interaction, offering benefits such as improved hygiene, durability, versatility, and accessibility. Ethical considerations underscore the importance of responsible design and user education, and as this technology progresses, it holds the potential to redefine our engagement with screens and digital interfaces, shaping a more inclusive, efficient, and immersive future.

**Keywords:** Touchless Touchscreen Technology, Gesture Recognition, Infrared Sensors, Cameras, Radar Technology, Haptic Feedback, AI and Machine Learning Algorithms, Biometric Recognition, 3D Scanning, Digital Well-Being.

### **Introduction:**

Touchless touchscreen technology has emerged as a transformative force, reshaping our interaction with screens and offering a range of benefits, from improved hygiene to versatile and natural user experiences. This introduction will delve into the intricacies of this technology, exploring key components such as gesture recognition, infrared sensors, cameras, radar technology, haptic feedback, and the integration of AI and machine learning algorithms. The discussion will extend to the challenges posed by existing technologies, highlighting their performance metrics and paving the way for an evaluation of their efficacy. Additionally, ethical considerations, data privacy concerns, and the importance of responsible usage will be addressed, emphasizing the need for a balanced approach that prioritizes both convenience and meaningful human interaction.

In the realm of touchless touchscreen technology, the methodology of algorithms plays a pivotal role in ensuring seamless and accurate interactions. Gesture recognition, a fundamental aspect, involves the identification and interpretation of various hand movements and gestures to facilitate user input. Infrared sensors, known for their cost-effectiveness, capture these gestures by detecting changes in infrared light patterns. Cameras offer a higher level of accuracy, capturing detailed visual data to interpret gestures in diverse lighting conditions. Radar technology, on the other hand, employs radio waves to precisely detect and analyze hand movements, particularly excelling in low-light conditions. The integration of AI and machine learning algorithms takes this a step further, enabling touchless screens to not only recognize gestures but also understand and respond to user emotions, thus enhancing the overall user experience.

The performance of touchless touchscreen algorithms is evaluated based on several key metrics. Accuracy, a critical parameter, measures the algorithm's ability to correctly identify and interpret user gestures. Precision and recall assess the algorithm's capacity to minimize false positives and negatives, ensuring reliable interaction. Response time is another crucial metric, gauging the speed at which the system translates gestures into actions, contributing to a seamless user experience. These metrics collectively determine the efficiency and reliability of the touchless technology, influencing its applicability across various domains such as self-service kiosks, gaming, and medical devices. The evaluation of touchless touchscreen technology involves assessing its real-world performance and user satisfaction. User feedback, gathered through surveys and usability studies, provides valuable insights into the practicality and effectiveness of the technology in different applications. Furthermore, service quality is measured by factors such as system uptime, maintenance requirements, and adaptability to user

needs. As touchless technology advances, ongoing evaluation becomes imperative to address emerging challenges, enhance system capabilities, and ensure continued user satisfaction.

The integration of advanced technologies in touchless touchscreens raises ethical considerations that demand careful attention. Data privacy, consent, and potential biases in AI algorithms must be thoroughly addressed to mitigate risks and safeguard user information. Responsible usage, coupled with user education on digital well-being, becomes paramount in striking a balance between convenience and meaningful human interaction. As the technology evolves, it is essential to uphold ethical standards and prioritize the user's right to privacy and security.

touchless touchscreen technology represents a multifaceted approach to human-computer interaction. From the methodology of algorithms to existing performance metrics and evaluations, this technology offers a glimpse into a future where screens and digital interfaces are more inclusive, efficient, and immersive. However, the journey forward must be guided by responsible design, ethical considerations, and ongoing evaluation to ensure a positive and transformative impact on our digital interactions.

### **Literacy survey:**

TITLE	AUTHOR	TECHNOLOGY	EXPLANATION	INFERENCE
Machine Learning for Touchless Touch Screen	Y. Vamsi Krishna, Akshatha K P, Meghana B	Machine Learning (SURF Algorithm, PROSAC Algorithm)	A semi-supervised action recognition system for touchless touchscreens is proposed. The system is trained on a dataset of labeled and unlabeled videos of various actions. Actions are recognized using SURF features and PROSAC comparison. Sensors are not used, reducing cost and complexity.	The development of touchless touchscreen technology raises ethical concerns about data privacy, user consent, and bias in AI algorithms. Both machine learning and rule-based approaches to touchless touchscreen technology present unique ethical challenges. Ethical considerations extend to system security, biometric recognition, and user education. A responsible and ethical approach to touchless touchscreen technology requires a balance between innovation and the protection of user privacy and well-being.
A SURVEY ON MACHINE LEARNING APPROACHE S AND ITS TECHNIQUES	Thomas. Rincy. N Dr. Roopam Gupta	Machine learning, Supervised learning, Unsupervised learning, Semi- supervised learning, Reinforcement Learning.	Machine learning is a field of study that enables computers to learn without being explicitly programmed. Machine learning is a subfield of computer science that is related to pattern recognition and artificial intelligence.	Different machine learning techniques and their approaches were examined. The study's key contributions include classifying machine learning approaches into supervised learning, unsupervised learning, semi-supervised learning, and reinforcement learning, as well as identifying their various

				algorithms.The researchers plan to create a model based on machine learning techniques in the future.
Hand gesture recognition for human computer interaction	Meenakshi Panwar, Pawan Singh Mehra	Machine Learning	The overall algorithm divided into four main steps, which includes segmentation, orientation detection, feature extraction and classification. The proposed algorithm is independent of user characteristics. It does not require any kind of training of sample data. The proposed Implemented algorithm has been tested on 390 images, gives a recognition rate of approximately 92% and average elapsed time of 2.76 sec. It takes a less computation time as compare to other approaches	The Future work will emphasize on removing of these constraints as these constraints limits the user from freely forming the gestures. And also the proposed approach will extend to apply on images including two hands so that most of the key pressing events can be generated by classifying more number of gestures. A

Machine Learning and Computer Vision-Enabled Traffic Sensing Data Analysis and Quality Enhancement	Guohui Zhang , Yinhai Wang	Machine Learning and Computer Vision	Traffic sensor data are essential for informed, scientific decision-making processes in traffic operation, pavement design, and transportation planning. In the current traffic detection infrastructure, inductance loop detectors and surveillance cameras are two commonly deployed sensors	a machine learning approach is developed to establish an artificial neural network (ANN) to better extract classified vehicle volumes from single-loop measurements. In addition, a set of computer vision-based algorithms are developed to extract background image from a video sequence, detect presence of vehicles, identify and remove shadows, and calculate pixel-based vehicle lengths for classification based on widely available surveillance camera signals.
Different paradigm for Touch-Screen technology: A Survey	Prof. Kamalakannan J. Chepuri Saikiran	Machine learning	Touch, the present giant ruling the world. One may think how touch is focused on all aspects. Traditionally, people used to work on keypad type systems by pressing hard buttons, but as the days and time progresses the user needs increasing in a huge demand for so it required of doing work in quick and efficient manner. My point in this paper is to describe that how touch evolved from past to present and even to future where the different types of touch-screen technology available in the market	Although now at this point these technology being facing with continuous feedback i.e., a initial touch point not exactly reflected when it is done multiple points. Researchers now continuing their growth of study at more and more touch technologies to work out on multi-touch support and ability to provide high resolution touch screen.

	Famid AI Farid	Machine learning	Researchers have	The recognition of hand
A Structured	Famio Al Fano	Machine learning	recently focused their	gestures is anticipated to
and			attention on vision-	play an important part in
Methodological			based hand gesture	our daily lives. In the
Review on			recognition. However,	modern world, most of
Vision-Based			due to several	the technologies around
Hand Gesture			constraints, achieving	us are mostly controlled
Recognition			an effective vision-	by hand gestures. In the
System			driven hand gesture	future, we want to
			recognition system in	improve our analytical
			real time has remained	approach to learn more
			a challenge. This paper	about gesture
			aims to uncover the	recognition techniques.
			limitations faced in	
			image acquisition	
			through the use of	
			cameras, image	
			segmentation and	
			tracking, feature	
			extraction, and gesture	
			classification stages of	
			vision-driven hand	
			gesture recognition in	
			various camera	
	Oinhana Va *	Machina laomina	orientations.	In costume based HMI a
C	Qiuhong Ke *,	Machine learning	Human–machine	In gesture-based HMI, a sensor such as
Computer	Jun Liu †,		interaction (HMI) refers to the	Microsoft Kinect is used
Vision for Human–			communication and	to capture the human
Machine			interaction between a	postures and motions,
Interaction			human and a machine	which are processed to
interaction			via a user interface.	control a machine. The
			Nowadays, natural user	key task of gesture-
			interfaces such as	based HMI is to
			gestures have gained	recognize the
			increasing attention as	-
			they allow humans to	meaningful
			control machines	expressions of human
			through natural and	motions using the
			intuitive behaviors	data provided by
				Kinect, including RGB
				(red, green, blue),
				depth, and skeleton
				•
	4 15 11	36.11.1	TT 4	information.
	Ayush Purohit	Machine learning	Human Action	Some of the limitations
A Survey on	*, Shardul		Recognition is one of	which are common
Human Action	Singh		the important research	among various
Recognition	Chauhan*		area in computer vision	approaches includes
			and image processing	camera calibrations,
			field. A machine can be	natural gestures, moving
			interacted and	human segmentation
			controlled by gestures	and action vocabulary.
			and facial expression	These problems mainly arise due to insufficient
			using the visual modality. Human	realistic datasets and
			action recognition can	restricted environments
			be seen as a bridge for	while developing the
<u> </u>	l	l	be seen as a bridge for	winie developing the

			machines to understand human body language, an intelligent interacting system between machines and humans which limits majority of input devices such as keyboard and mouse.	system.
study of touch less touch screen technology	Nilofar E. Chanda	Machine learning	Touch screen displays found everywhere in the world. The touch screen display provides a greater flexibility to user but after some years touch screen display becomes less sensitive which causes failure of touch on touch screen display	With the use of this technology the user gets flexibility to use the system appropriately. The proposed touch less touch screen technology is suitable for android mobile phone, laptop, ATM machine etc. By using this technology maintenance work of touch screen display can reduce.
Gesture-Based Touchless Operations: Leveraging MediaPipe and OpenCV	Aryan Gupta,Naman Chawla, Rachna Jain, Narina Thakur, Ajantha Devi	Machine learning	The recognition of human movements is enabled through the implementation of a convolutional neural network (CNN). Within this study, we develop a simple hand tracking method for controlling a surveillance car operating on the Robot Operating System (ROS) by utilizing socket programming. Our	Humans have only recently begun using hand gestures to interact with computers. The integration of the real and digital worlds is the aim of gesture recognition. It is considerably simpler to convey our intents and ideas to the computer via hand gestures

Touchless Advertising Mobile Application	Dyan Tannoo; Aatish Chiniah; Meekshi Jaunkeepersad; Humaïra Bibi Baichoo	Machine learning	Hand gesture recognition has been a promising area of research in the past decades, especially with breakthroughs in the field of computer vision, but the COVID-19 pandemic has brought much attention to this field. The emphasis on a higher sanitary standard has pushed for more touchless interactions to help mitigate viral contagion.	This technology could prove to be highly beneficial for interactive devices located in public spaces, such as self-serving information kiosks and self-service check-outs. This paper proposes a touchless advertising mobile application as a proof of concept, to display and interact with advertisements, as a proof of concept.
Sustainable society with a touchless solution using UbiMouse under the pandemic of COVID-19	Junichi Takatsu, Ryoji Otsu	Machine Learning	This paper introduces a new artificial intelligence software which is capable of controlling devices using fingers in the air. With Ubimouse, touchpanels, restaurant ordering systems, ATM systems, and etc., which are commonly used by various people in public, can be contact-less devices. These touch-less devices, especially under the harsh conditions of COVID-19, are desired to prevent infections mediated by touch devices.	Thus, in fields using gloves, there is a demand for non-contact device operation. To satisfy these demands, we have developed "UbiMouse". This is an AI software that allows you to operate the device by moving your fingers toward the device. In the AIs in UbiMouse, a convolution model and a regression model are used to identify fingers' features from camera footage and to estimates the position of a detected finger, respectively. We demonstrate an operation of UbiMouse without contact.
Tactile Feedback for Above-Device Gesture Interfaces: Adding Touch to Touchless Interactions	Stephen Brewster, Vuokko Lantz	Machine learning	device gesture interfaces let people interact in the space above mobile devices using hand and finger movements. For example, users could gesture over a mobile phone or wearable without having to use the touchscreen. We look at how above- device interfaces can also give feedback in the space over the device.	Recent haptic and wearable technologies give new ways to provide tactile feedback while gesturing, letting touchless gesture interfaces give touch feedback. In this paper we take a first detailed look at how tactile feedback can be given during above-device interaction. We compare approaches for giving feedback (ultrasound haptics, wearables and direct feedback) and

				also look at feedback design. Our findings show that tactile feedback can enhance above-device gesture interfaces.
TOUCHLESS ELECTRONIC VOTING MACHINE WITH AN AI- FACIAL RECOGNITIO N	Rajendra P Prasad; Sunil K N Kumar; Ravi Gatti; M Pranav; G. Rahul; M. Yatheesh	Machine Learning	This project is used to maintain High level biometric security. The voter details are stored in dataset directory. Before entering the voting process the person should stand in front of the PC, the camera will read the image of the voter. After reading the details PC allows the authorized person to Vote and a signal is sent to the microcontroller and user just need to hover over the voting party to Vote with no need to touch the EVM at all. The application software maintains the person data. In the "smart voting system" once a person casts his vote, the controller informs vote is successfully registered.	Elections are important distinguishing qualities of every democracy that is ruled by the people expressing their choices or their ideas through voting. Voting mechanisms have progressed from simple handwritten ballots to digital voting systems in leaps and bounds. This project aims to build a smart voting system using face recognition technology that allows any voter in INDIA by going to their respective constituency from "ANYWHERE IN INDIA" to the nearest voting booth in the place of stay
PalmSpace: leveraging the palm for touchless interaction on public touch screen devices	Nath, Pinku Deb	Machine Learning	With PalmSpace, UI elements are mapped onto the users' palms and can be accessed by touching various locations directly on the palm. I conducted a series of user studies to evaluate several design options, such as interface layout, item size, preferred item location, and suitable feedback for items.	The touchscreen is the primary solution to interact with public devices such as Automated Teller Machines (ATMs). However, the touch modality raises health concerns since users have to touch the screens, and therefore risk the spread of contagious diseases. I designed PalmSpace, an alternate input technique leveraging users' hand palms to interact with public devices.

# **Machine Learning (ML):**

In the Machine Learning (ML) methodology for touchless touchscreen technology, the system relies on algorithms trained on extensive datasets to recognize and interpret user gestures and emotions. The process begins with the Input Sensing module, where sensors like cameras or infrared arrays capture various user interactions, such as hand movements and facial expressions. This raw data undergoes Data Preprocessing to extract relevant features, eliminate noise, and prepare it for analysis. The AI Processing module, employing ML algorithms, analyzes the preprocessed data, learning patterns and correlations to infer user intent. The system then generates an Output, which could include visual or audio feedback or actions within the software or physical environment. While ML offers personalized interactions, concerns regarding Data Privacy arise due to the vast amounts of user data involved. Additionally, addressing Algorithmic Bias and ensuring Transparency and Explainability become critical considerations for responsible implementation.

### **Rule-Based Systems (RBS):**

In the Rule-Based Systems (RBS) approach, predefined rules govern user interactions, providing a structured and predictable framework for the touchless touchscreen system. The Input Sensing module captures user interactions through sensors, and Data Preprocessing ensures that the data aligns with the predetermined rules. The AI Processing module interprets user intent based on these rules and generates an Output accordingly. Rule-based systems offer reliability and consistency in interactions, but their rigid structure may limit adaptability and inclusivity. The ethical considerations include ensuring that the predefined rules strike a balance between providing a clear framework and avoiding undue restrictions on user freedom. Continuous refinement of rules based on user feedback contributes to an iterative and adaptive system.

## **General Architecture for a Touchless Touchscreen System:**

### **Input Sensing:**

In the initial stage of the touchless touchscreen system, sensors, including cameras, depth sensors, or infrared arrays, serve as the capture mechanisms to record user interactions. These interactions encompass a diverse range of data types, including hand gestures, facial expressions, and body movements, providing a comprehensive dataset for subsequent analysis.

# **Data Preprocessing:**

Moving to the Data Preprocessing phase, the raw data undergoes several critical steps. Relevant features are extracted through Feature Extraction, ensuring that the information essential for analysis is isolated. Simultaneously, Noise Reduction mechanisms are employed to filter out any extraneous or erroneous data, enhancing the overall accuracy of the subsequent processing stages.

# **AI Processing:**

The heart of the touchless touchscreen system lies in AI Processing, where the processed data takes different paths based on the chosen methodology. In the case of a Machine Learning (ML) approach, the system engages in Algorithmic Training. ML algorithms are trained on the preprocessed data, enabling them to discern patterns and correlations within the dataset. Conversely, in a Rule-Based System (RBS), the predefined rules are applied, allowing the system to interpret user intent and guide responses accordingly.

# **Output Generation:**

The culmination of the process is the Output Generation stage. Here, the system generates responses, offering feedback or initiating actions based on the AI's interpretation of user intent. These outputs manifest as visual feedback, audio cues, or actions within the software or physical environment, contributing to the overall user experience.

## **Data Privacy:**

In terms of ethical considerations, Data Privacy takes precedence. User data collection mandates explicit consent, adhering to stringent privacy regulations. Secure Storage mechanisms are imperative, safeguarding user data from unauthorized access and potential breaches.

## **Algorithmic Bias:**

Addressing Algorithmic Bias is crucial for fairness. Continuous evaluation and refinement of ML algorithms are essential to minimize biases that may lead to unfair outcomes. Ensuring Diverse Representation in training datasets helps prevent the reinforcement of societal biases.

### **Transparency and Explainability:**

To foster trust and understanding, Transparency and Explainability are key. Users should have a clear understanding of how the AI system interprets their interactions and makes decisions. Transparent communication about the system's logic and functioning further enhances user trust.

#### **User Education:**

Educating users about the system's capabilities and limitations is pivotal. Providing insights into System Capabilities and informing users about the Ethical Implications of their interactions encourages responsible and informed usage.

# **Biometric Recognition:**

Incorporating Biometric Recognition for security purposes introduces its own set of ethical considerations. Robust safeguards are essential to protect user identities, and mechanisms must be in place to prevent Unauthorized Access through biometric data.

## **System Security:**

The overarching ethical responsibility extends to System Security. Robust Security Measures should be incorporated to guard against potential misuse or unauthorized access to sensitive user data. Regular Vulnerability Management ensures continuous assessment and enhancement of the system's overall security posture.

# **DISCUSSION:**

The emergence of touchless touchscreen technology signifies a transformative shift in how users engage with digital interfaces. This discussion explores the various facets of touchless touchscreen systems, encompassing their architectural components, ethical considerations, and the broader implications for human-computer interaction. The general architecture outlined encompasses a comprehensive journey of data, commencing with Input Sensing utilizing sensors like cameras, depth sensors, and infrared arrays to capture diverse user gestures and expressions. Data Preprocessing refines the raw data, employing Feature Extraction and Noise Reduction to ensure accuracy. In AI Processing, the system interprets user intent through either Machine Learning (ML) algorithms or predefined rules, leading to Output Generation that ranges from visual feedback to tangible actions.

Touchless touchscreen systems bring forth a myriad of ethical considerations. Data Privacy is a paramount concern, necessitating explicit user consent and robust Secure Storage mechanisms. Algorithmic Bias, especially in ML approaches, demands continuous evaluation and a commitment to Diverse Representation in training datasets. Transparency and Explainability become crucial in fostering user trust, ensuring users comprehend how the system interprets their interactions. User Education assumes a pivotal role, emphasizing the capabilities and limitations of the system and informing users about the Ethical Implications of their interactions. The inclusion of Biometric Recognition introduces complexity, requiring safeguards to protect user identities and prevent Unauthorized Access. System Security encompasses comprehensive measures to thwart potential misuse or breaches, necessitating ongoing Vulnerability Management.

Touchless touchscreen systems hold significant potential for transforming human-computer interaction. ML methodologies offer adaptability and personalization, contributing to a seamless and immersive user experience. The structured and predictable nature of Rule-Based Systems ensures reliability, making touchless technology accessible to a broader user base. The emphasis on transparency and user education fosters a sense of empowerment and trust, shaping a more inclusive and efficient digital future, the discussion on touchless touchscreen systems

highlights their transformative potential and the ethical considerations inherent in their development. This convergence of technology and human-centric design underscores the need for responsible design, ethical usage, and an ongoing dialogue to shape the trajectory of touchless touchscreen technology and its impact on our digital interactions.

#### **Ethical considerations:**

Touchless touchscreen technology has ushered in a myriad of applications across various domains, presenting a range of ethical considerations that are crucial to address. The overall applications and ethical considerations span diverse fields, from healthcare and public spaces to entertainment and security.

#### Healthcare:

In healthcare settings, touchless touchscreens find applications in interactive medical devices and self-service kiosks. The ethical considerations here revolve around patient privacy, data security, and the responsible use of biometric data. Ensuring that sensitive health information is handled with the utmost care and that patients explicitly consent to data collection becomes paramount.

# **Public Spaces and Retail:**

Touchless technology is increasingly prevalent in public spaces, such as airports, malls, and public transportation. The ethical considerations include user consent for data collection, transparent communication about how the technology operates, and measures to prevent unauthorized access or misuse of user information. Striking a balance between convenience and privacy is essential in these public settings.

#### **Education:**

In educational institutions, touchless touchscreens may be utilized in interactive learning environments. Ethical considerations involve safeguarding student data, obtaining parental consent for minors, and ensuring that the technology is used responsibly to enhance the learning experience without compromising individual privacy.

# **Entertainment and Gaming:**

Gaming and entertainment applications of touchless touchscreens introduce ethical considerations related to user engagement and potential addictive behaviors. Designing these applications responsibly includes considering the mental well-being of users, providing transparent information about data usage, and incorporating features that promote healthy usage patterns.

# **Security and Access Control:**

Touchless technology is integral to security systems, including biometric recognition for access control. Ethical considerations encompass user consent, preventing the misuse of biometric data, and implementing robust security measures to protect against unauthorized access. Striking a balance between heightened security and individual privacy is crucial in these applications.

# Workplaces:

In corporate environments, touchless technology may be deployed for access control, time tracking, and collaborative workspaces. Ethical considerations involve ensuring employee consent, transparent communication about data usage, and safeguarding against potential workplace surveillance concerns. Respecting individual privacy rights in the workplace is essential.

# **Overall Ethical Considerations:**

• Informed Consent: Users should be informed about the collection, storage, and use of their data, and explicit consent should be obtained before deploying touchless technology.

- Transparency: Clear communication about how the technology operates, the purpose of data collection, and the implications of user interactions is crucial to building trust.
- Data Security: Robust measures should be in place to secure user data from unauthorized access or breaches, minimizing the risk of data misuse.
- Bias Mitigation: In applications using machine learning, efforts should be made to identify and mitigate biases in algorithms to ensure fair and equitable user experiences.
- User Empowerment: Educating users about the capabilities, limitations, and ethical implications of touchless technology empowers them to make informed decisions about its usage.
- Accessibility: Ensuring that touchless technology is designed to be inclusive and accessible to individuals
  with diverse abilities and backgrounds is an ethical imperative.

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