

Effects of li doping on superconducting properties of citrate-gel prepared $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ Compound

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Received: 29 May 2022 / Accepted: 20 October 2022 © The Author(s), under exclusive licence to Springer-Verlag GmbH, DE part of Springer Nature 2022

Abstract

The $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ polycrystalline bulk superconductors doped with Li substituting at the Y site at different concentrations (x=0, 0.01, 0.02, 0.1) were prepared using the citrate-gel method to study the effects of doping on the superconducting temperature and critical current density. The question was whether Li addition characterized by a high Debye frequency would have any positive effects on T_c . The optimum citrate-gel and heat treatment conditions were identified as those yielding samples with a maximum grain size on the order of 50 µm (up to the optimum Li doping level, x=0.01). Li substitution at the Y site was verified by structural, electrical, and magnetic measurements of the produced samples, whereas X-ray diffraction (XRD) analysis revealed the formation of a pure phase with no visible impurity phases. Moreover, AC magnetic susceptibility measurements showed no increase in the superconducting transition temperature T_c , consistent with the predicted results obtained by machine learning method, although it was theoretically expected to increase owing to the high Debye frequency of Li. This observation is consistent with magnetic coupling models for pairing mechanism in cuprates. Finally, because of the optimum conditions of the preparation procedure, nearly identical values of the critical current density (J_c) were recorded for samples with different Li doping levels (up to the optimum Li doping level). It was found that improved compound preparation conditions would have a critical and extensive effect on J_c enhancement, with nearly no T_c suppression.

Keywords High- T_c superconductivity \cdot Cuprate \cdot Li doping \cdot Citrate-gel \cdot YBCO \cdot Machine learning

1 Introduction

Superconductivity is among the most important discoveries of the twentieth century [1]. High-temperature superconductors (HTS) have found numerous technological applications such as MRI systems, electromagnets, transformers, and high-speed trains [2]. Many studies have focused on YBa₂Cu₃O_{7- δ} (Y-123) cuprates and their modifications to improve their T_c and J_c [3–5]. The great interest in the study of Y–123 properties stems from the fact that the physical properties of Y–123 systems can be modified over a wide range of doping levels without any significant changes in the lattice structure [6, 7].

The fundamental mechanism of superconductivity of the cuprate family is not yet well-known and numerous theoretical models attempt to explain the phenomenon. The substitution of different elements in the Y-123 structure is generally meant to understand mechanism of cuprates, to improve its superconducting properties as well as its physical properties for various applications. One such property is its high critical current density, which is essential for the industrial applications of high-temperature superconductors (HTS). Since the discovery of Y-based cuprate superconductors, substitution of elements at different atomic sites, especially at the Cu site, has been widely studied to find that elements such as Li, Zn, or Ni substituted at the different cationic sites in the Y-123 structure would act as current pinning centers [8-10]. Furthermore, elements with suitable covalent radii (Co, Li, Ni, and Pd) substituted for Cu have been reported to improve the value for J_c but to reduce that of T_c [11]. The doping suppresses superconductivity locally and can act as pinning

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centers resulting in an increase in critical current density. It has also been reported that flux pinning and J_c enhancement due to Zn and Li doping at the Cu site $(YBa_2(Cu_{1-v}Zn_v)_3O_{6.64})$ with y > 0.5% for Zn and in $YBa_2Cu(_{100\%-x}Li_x)_{3/100\%}O_7$ with x > 0.06 at.% for Li) could be attributed to the formation of local magnetic moments in the proximity of non-magnetic ions such as Li [11–14]. Neutron diffraction experiments have revealed that Cu in CuO₂ planes is replaced with Li [15–18], thereby reducing the number of holes in the copper planes that ultimately leads to T_c degradation [19]. Also, Cu substituted with Li and Ba substituted with K or Na have been reported to lead to declines in T_c with the lowest T_c obtained with Li substitution (in YBa₂Cu_{3-x}Li_xO₇ with x=0, 0.05 and 0.2) [20]. Changes in the distribution of carriers have also been claimed as another cause underlying T_c reduction in Li-Cu materials. In general, Li substitution for Cu in YBCO reportedly reduces T_c , c-axis length, and unit cell volume [15–17, 19–21], indicating that Li actually enters the crystal structure.

From a different perspective, alkali metals with smaller ionic radii, such as Li, have been observed to cause more superconductivity suppression than the ones with larger ionic radii (i.e., Na and K). From a theoretical perspective, substitution of K gives rise to changes in the electronic band structure and density of states (DOS) that lead to declining T_c [19]. It has been established that substitution of K⁺ ions at the Ba²⁺ site leads to the hole doping effect that leads to a slight increase in T_c [22]. In a computational study, the only study in the literature about Li addition for Y site of YBCO compound, Li substitution was observed to lead to slight increases in T_c at low doping levels due to the strengthened two-dimensional character of the bonds in CuO₂ planes giving rise [19].

Although the present work had been conceived and initiated prior to the authors' learning of these earlier findings, it had been motivated by a study of Li substituted at *the Y* site in $YBa_2Cu_3O_{7-\delta}$ cuprate at low doping levels. (In the literature, all Li doping studies at the Cu site were done by more than 5% amount of Li addition.) Despite the extensive reports on the superconducting properties of YBCO doped with different elements, no systematic study had been reported on Li doping at the Y site, to the best of the authors' knowledge, in the literature, especially regarding the highquality products prepared through the citrate-gel method in this study, compared to the standard solid-state method. The research question raised was whether partial substitution of Li characterized by a high Debye frequency would have any positive effects on T_c and J_c or the electrical and magnetic properties of Y-123. In the BCS theoretical view, it is expected that the superconducting T_c increases owing to high Debye frequency of the atoms in the compound. To answer this question, fully oxygenated Y_{1-x}Li_xBa₂Cu₃O_{7-δ} polycrystalline samples were synthesized with different levels of Li dopant using an optimum citrate-gel and heat treatment process. A renewed investigation was also carried out on a non-magnetic Li-doped system that had shown enhancements in T_c and J_c values in a previous study of magnetic element doping [23, 24]. We show although the Li dopant was non-magnetic and was capable of bearing a high phonon frequency, no remarkable increases were observed in T_c , thereby refuting the hypothesis that T_c increases with phonon frequency. The results were used to shed light on the mechanism of high- T_c superconductivity in cuprates. Moreover, the optimum conditions of the preparation of samples were identified so that we obtained samples with a grain size on the order of 50 µm. This finding of the optimum conditions of the preparation gives nearly identical values of the critical current density (J_c) for samples with different Li doping levels (up to the optimum Li doping level x = 0.01). In a word, apart from the chemical doping effect, improved preparation conditions would have an extensive effect on J_c enhancement. The present study discloses a contribution to the issue of technological implications of chemically doped citrate-gel prepared YBCO compound. Finally, the citrate-gel synthesis method was exploited in this study due to the demand for homogeneous powders and granule size control required for electrical and magnetic applications of ceramics.

2 Materials and methods

2.1 Materials

The reagents of LiNO₃ (99.99%), Y(NO₃)₃.6H₂O (99.99%), Ba(NO₃)₂ (99.99%), and Cu(NO₃)₂.3H₂O (99.99%) powders were purchased from Merck. Deionized water, citric acid (C₆H₈O₇), and ethylenediamine (C₂H₄(NH₂)₂) were used in the synthesis process.

2.2 Citrate-gel synthesis method

Fully oxygenated polycrystalline samples of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ (x=0, 0.01, 0.02, and 0.1) were synthesized using the citrate–gel method. In citrate-gel method, it could be obtained more homogeneous, single-phased samples and granule size control. For a typical sample, stoichiometric proportions of metal nitrates (0.5 M aqueous solutions) were mixed to obtain a light blue solution to which citric acid was added dropwise at a rate of one drop per 5 s. Next, the pH was raised to 6.8 by adding ethylenediamine dropwise. The final gel was obtained by the constant stirring of the solution at 85 °C for 6 h.

The citrate–gel method involves the formation of a stable complex of cation nitrates in a solvent of water and citric acid. Citric acid takes part in the reaction with a valence of +1 to generate cit-H₂ ions, whereas it forms a weak bond with Ba²⁺. Ethylenediamine was, therefore, used as a complexing agent to form ethylenediamine nitrate that prevents Ba²⁺ precipitation. The final gel was heated at 520 °C to yield a dark brown powder via pyrolysis. The resulting powder was ground and heated twice at 880 °C with intermediate grinding. Eventually, the final powder was thoroughly ground again and pressed into rectangular pellets $(1 \times 3 \times 10)$ mm³) under a 40 bar load before being subjected to the final heating process at 900 °C for 24 h in a pure oxygen flow in order to obtain fully oxygenated samples [25] that were slowly cooled to room temperature. All the samples were prepared concurrently under the same conditions to ensure

2.3 Characterization

The samples were characterized with respect to their crystal structure, phase purity, and lattice parameters via XRD investigations by $Cu-K_{\alpha}$ radiation using a Philips X'pert MPD system operated at a 2θ range of 10° – 90° . The transport J_c data were obtained using a four-probe technique at 77 K. A Philips XL 30 SEM was used to prepare SEM images of the samples. AC magnetic susceptibility measurements were carried out using a Lakeshore Model 7000 susceptometer in a magnetic field of 0.8 A.m⁻¹ at a frequency of 333 Hz over a temperature range of 77-100 K. Resistivity was measured using the standard DC four-probe technique. The XRD patterns were analyzed in both X'Pert Highscore 3.0.5 and FullProf Suite version 7.30. The SEM images were studied using Digimizer v. 5.3.5. Also, OriginPro 2019b 9.6.5.169 and VESTA v. 3.4.8 (2019) software tools were used for generating graphic illustrations.

meaningful and consistent comparisons. The obtained sam-

ples show the longest grain size to date; see Fig. 4.

3 Results and discussion

3.1 Structural analysis by XRD

Figure 1a shows a typical layered perovskite-like the structure of YBCO compound with two CuO₂ planes separated by a layer of Y atoms bound by the Ba–O and Cu–O layers. Clearly, the O(2) and O(3) oxygen atoms are strongly coupled with Cu(2) in the CuO_2 planes. The XRD patterns were analyzed using the Rietveld refinement method (Fig. 1b). The normalized XRD patterns for pure and Li-doped samples with x = 0.01 and 0.1 are shown in Fig. 1c.

The analyses revealed that the samples were singlephased with an orthorhombic crystal structure of identical perovskite types with *Pmmm* symmetry (JCPDS 85–1877 reference pattern). The lattice parameters for the pure sample were essentially consistent with those reported in the literature. The X-ray analysis showed identical average oxygen



Fig.1 a Shematic of $YBa_2Cu_3O_{7-\delta}$ crystal structure, b Rietveld refinement plot of powder XRD pattern for Y_{0.99}Li_{0.01}Ba₂Cu₃O_{7-δ} sample, and **c** normalized XRD patterns of YBa₂Cu₃O₇₋₈, $Y_{0.99}Li_{0.01}Ba_2Cu_3O_{7-\delta}$ and $Y_{0.9}Li_{0.1}Ba_2Cu_3O_{7-\delta}$ samples

contents of the samples $(7-\delta)$ with a constant value of 6.97 ± 0.02 independent of the substitution degrees [26, 27].

A careful examination in Fig. 1c and comparison of the XRD patterns of both the pure and doped samples reveal strong peaks at (002), (003), (005), (006), and (007), especially for the doped sample. The increasing relative peak intensity in the Li-doped sample indicates grains with a c-axis preferred orientation [28].

Clearly, the ionic radii of Li⁺ at four-, six-, and eightfold coordinated sites are 0.59, 0.76, and 0.92 Å, respectively, while those of Y^{3+} at six- and eightfold coordinated sites are 0.9 and 1.019 Å, respectively. This is while the ionic radii of Cu²⁺ at four- and sixfold coordinated sites are 0.57 and 0.73 Å, respectively, but 0.54 Å for that of Cu⁺ in the sixfold coordinated sites [29]. Because of the similarity of the ionic radii of Y³⁺ and Li⁺ at eightfold coordinated sites, at which Li⁺ is by 9.7% smaller than Y³⁺, and given the fact that Li⁺ is by 3.5% (4.1%) larger than Cu²⁺ at a 4(6)-fold coordinated site, Li^+ ion is seen to move into the Y^{3+} site at low doping levels (as in the present study) [19]. This substitution not only leads to a distortion of CuO₂ planes but might also strengthen the two-dimensional character of the charge carriers in these planes [19]. According to Table 1, the CuO₂ planes are more buckled by 1% Li⁺ doping, while the Cu(2)-O(2)-Cu(2) or Cu(2)-O(3)-Cu(2) angles in the CuO₂ planes have decreased. This led to an increase in T_c , which might be related to orthorhombicity that is known to increase with increasing distance between the (013) and the (103) peaks (see the inset of Fig. 1c). Moreover, because of the Li⁺ and Cu²⁺ similarity (i.e., 0.73 Å vs 0.76 Å in an octahedral coordination and 0.57 Å vs 0.59 Å in a tetrahedral coordination), Li^+ is reportedly able to enter the copper site in the CuO₂ planes [21] (at higher doping levels), reduce hole concentration, and cause a T_c suppression as well. In the present study, both these effects had a synergic impact as a result of adding 1% Li that led to a slight reduction in T_c or to maintain a nearly constant T_c value (see Fig. 2). (It may be noted that in the electrical resistivity measurements, T_c experienced a slight increase by 1% Li doping before it started to decrease at higher doping levels; Fig. 3.) It may be claimed that Li moves mainly to the Y site, which would increase T_c due to the strengthened two-dimensional character of the carriers

Table 1 Unit cell parameters, bonding intervals and angles in $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ samples for x=0, 0.01, and 0.1 as obtained by XRD Rietveld refinement. T_{conset} (K) and $T_{c mid}$ (K) are experimental superconducting transition temperature obtained from susceptibility measurements

Li-doping level	0%	1%	10%
Lattice constants (Å)			
а	3.8139	3.8130	3.8183
b	3.8836	3.8783	3.8827
С	11.6643	11.6762	11.6583
Unit cell volume (Å3)			
V	172.7680	172.6698	172.8398
Bond lengths (Å)			
Cu(2)–O(2)	1.9208 (3)	1.9237 (3)	1.9564 (5)
Cu(2)–O(3)	1.9683 (4)	2.0076 (6)	1.9524 (3)
Cu(2)–O(4)	2.1745 (18)	2.1846 (18)	2.3537 (18)
Cu(1)–O(4)	1.9641 (16)	1.9861 (16)	1.7758 (16)
Cu(2)–Cu(2)	3.3830 (19)	3.3295 (19)	3.4244 (19)
Bond angles (deg.)			
Cu(2)-O(2)-Cu(2)	166.75 (12)	165.10 (12)	155.04 (12)
Cu(2)–O(3)–Cu(2)	161.63 (12)	150.66 (12)	168.59 (13)
T_{conset} (K)	92.61	92.01	92.34
T _{cmid} (K)	92	91.5	90.4



Fig. 2 AC magnetic susceptibility vs temperature of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ samples with $x=0.0, 0.01, 0.02, 0.1. T_{conset}$ is defined as the temperature below which the real part of AC susceptibility deviates from its normal-state linear behavior and drops to a diamagnetic superconducting phase

of CuO₂ planes [19], but that it is partially substituted at the Cu²⁺ site, which would drastically decrease T_c ; hence, the overall result is a slight reduction in T_c .

Finally, if the Li⁺ dopant goes into the Cu(1) site of the chains, it would result in a lattice expansion along the *a*-axis, causing a great decrease in T_c [15–17, 19–21]; this was not, however, the case observed in the present study (see Table 1). It should be noted that the addition of impurities to high- T_c cuprates gives rise to variations in T_c mainly due to the related changes in carrier density and oxygen content.



Fig. 3 The electrical resistivity vs temperature of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ samples with x=0.0, 0.01, 0.1

This does not, however, mean that the effects of impurity on pair breaking and scattering could be ignored [30].

Table 1 presents the lattice parameters, bonding intervals, and bond angles for pure YBa₂Cu₃O₇₋₈ and Li-doped samples. Clearly, the Cu(2)-O(4) distance increased with Li doping, as did the lattice constant c. However, Cu(2)–Cu(2)distance (defined as the spacing between CuO₂ planes) has been reported to decrease at low doping levels. According to Anderson's theoretical model [26], T_c is predicted to increase with decreasing separation between the CuO₂ planes. The same trend has also been reported for LaBaCuO and LaSr-CuO, for which a rapid increase has been observed in T_c as a result of either applying an external pressure or substituting a smaller cation for the La site that reduce the distance between CuO₂ planes. This is, however, a controversial point on which there is no universal agreement. Hardy et al. [31], for instance, applied uniaxial pressure along the tetragonal *a*- and *c*-axes for an optimally doped HgBa₂CuO_{4+ δ} (Hg-1201) cuprate. By bringing the CuO₂ planes closer together through the pressure applied along the *c*-axis, T_c was expected to decrease rapidly in accordance with Anderson's theoretical model, while it was observed to increase as the area of the CuO₂ plane declined [26]. Also, Locquet et al. observed an almost twofold increase in T_c in LaSrCuO₄ thin films epitaxially grown on a suitable substrate, which would increase the separation between the CuO_2 planes [32]. In a recent review, Schilling et al. [33] concluded that T_c in optimally doped cuprates could increase in two ways: by increasing the separation between CuO₂ planes or by reducing their plane area. The dependence of T_c on the separation between planes or plane area differs widely from one experimental group to another in Fe-based superconductors that form another class of nearly 2D-layered superconductors in which metallic layers of Fe atoms form a planar square lattice bonded with as atoms above and below them [33]. For instance, by substituting smaller rare-earth cations for La in LaFeAsO_{1-x} F_x (La-1111) and reducing the separation between the conduction FeAs layers, T_c exhibits a great increase [34]. In Y-doped SmFeAsO compounds [35], however, both enhancements and reductions in T_c have been reported despite the reductions shown by the compound in its a and c lattice parameters.

3.2 AC magnetic susceptibility results

An increase in T_c is expected to occur by Li addition due to its high Debye frequency and light atomic weight. As elaborated in the BSC theory, higher Debye frequencies would lead to higher T_c values. The AC magnetic susceptibility results exhibit the T_c values reported in Fig. 2. In this case, T_{conset} is defined as the temperature below which the real part of AC susceptibility deviates from its normal-state linear behavior and drops to a diamagnetic superconducting phase. As illustrated in Fig. 2, the superconducting transition is sharp for pure and Li_{0.01}-doped samples with no secondary steps on the transition curve, showing the samples are homogeneous and single-phased. Moreover, the transition temperatures of the doped samples are only slightly lower than that of the pure sample; for pure and 10% doped samples, $T_{c \text{ mid}}$ is 92 K and 90.4 K, respectively. (see Table 1). In other words, T_c is almost unchanging even for a Li doping of 10%. This is consistent with the magnetic coupling models and observations reported for cuprate superconductors [23, 24], where by magnetic element doping at the Y site (Co, Fe and Ni element doping) T_c increased, while the substitution of two non-magnetic elements, i.e., Ca and Sr, results in the T_c degradation. The increase in T_c through magnetic doping, which was a new observation, confirms the effective role of magnetically mediated pairing in cuprates [24]. However, a remarkable reduction in T_c has been observed for a substitution of Li for Cu that is attributed to the lower hole concentration [29]. To note, comparing with the recent observations in cuprate superconductors [24], where the doped sample with T_c nearly to the pure one is an optimum doped sample, here, we could consider x = 0.01 as an optimum doping level of Li. Therefore, we focused mainly on pure, x = 0.01 and a higher doping level of Li, such as x = 0.1, in most of the analyses.

3.3 Electrical resistivity

Figure 3 exhibits the temperature-dependent resistivity of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ samples with x=0, 0.01, and 0.1 measured in a zero magnetic field. In the normal state down to above T_c , the temperature dependence is linear for all the prepared samples, exhibiting a transition from a metallic behavior to a superconducting one. The linear behavior of $\rho(T)$, a distinctive feature observed at optimal doping levels in high- T_c cuprates, shows that all the samples are fully oxygenated at the optimal carrier concentration [24, 36] such that T_c remains nearly constant when 10% Li is added.

3.4 SEM analysis

Figure 4 presents the SEM micrographs of the samples at resolutions of 10 μ m (Left) and 20 μ m (Right). These polycrystalline samples show the longest grain size to date when compared with the microstructural analyses reported so far on the different bulk preparations of cuprates. These observations bear witness to the optimum preparation conditions employed in the present study. Grains as large as 50 μ m are found in the samples along with closely packed smaller ones (Fig. 4). The large and closely packed grains contribute to the higher J_c values of the samples in comparison with the values reported for similar samples prepared using the solid-state method (Fig. 5). The pure sample exhibits a J_c of



Fig. 4 SEM micrographs (scanning electron microscope) of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ samples with **a**, **b** x=0, **c**, **d** x=0.01 and **e**, **f** x=0.1. The micrographs are shown in 10 µm (Left) and 20 µm (Right) resolutions

around 100 A.cm⁻² [23] for the solid-state sample, whereas a value of 7 times higher was recorded for the citrate–gel samples.

3.5 J_c measurements

The V–J characteristic curves for the $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ samples (x = 0, 0.01, 0.1) measured at 77 K are also presented in Fig. 5. The J_c value is determined by extrapolating the linear part of the V–J curve at the intersection with the horizontal current density (J) axis. The maximum J_c values obtained for the x=0 and x=0.01 samples were 680 and 633 A.cm⁻², respectively. The 1% Li doping is the concentration with the least T_c reduction (in the corresponding magnetic susceptibility measurements) and maximum J_c (i.e., optimum concentration); higher concentrations of Li above the optimum value led to a lower pinning strength and reduced J_c (see Fig. 5). By introducing the optimum Li dopant, no remarkable change was detected in the magnitude of J_c , indicating that the sample preparation and heating process was performed under optimum conditions; hence, no considerable differences or effects were observed either in the pinning properties or in J_c values as a result of adding Li. However, some studies have indicated that J_c would exhibit a more than twofold increase in Li-substituted YBCO prepared by the melt-textured method when compared with the plain undoped compound [37]. Adding impurity normally affects the pinning properties in cases where partial



Fig. 5 V–J characteristic curves of $Y_{1-x}Li_xBa_2Cu_3O_{7-6}$ samples (x=0, 0.01 and 0.1) at 77 K. J_c of each sample is where the extrapolation of the linear part of the V–J curve, cut the horizontal axis. By increasing Li dopant, there is no big change in the value of J_c

clustering of the dopant or dislocations appears in the crystal structure.

Despite the remarkable advances made in achieving high critical current densities in cuprates by neutron, electron, or heavy ion irradiation and in spite of the extensive research conducted to determine the effects of chemical substitutions, oxygen deficiency, or structural defects on the pinning mechanism in cuprates, the main problem of increasing flux pinning due to chemical doping in YBCO yet remains to be resolved. It has been found that chemical substitutions into CuO₂ planes and/or partial clustering of the dopant between the grains form effective pinning centers in the YBa₂Cu₃O_{7- δ} compound. The main disadvantage of the conventional preparation method is the need for a large number of pinning centers to increase J_c , which could lead to a drastic T_c suppression. The improved preparation process proposed in the current study, however, led to a profound effect on J_c enhancement, with nearly no T_c suppression. High critical current density is crucial for the technological applications of bulk high-temperature superconductors.

3.6 Prediction: machine learning results

The effect of Li doping on the superconducting temperature of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ samples was studied by machine learning method, too.

Physicists attempt to develop mathematically based models that can predict the future behavior of a system as well as effectively predict the outcomes of experiments that have not yet been measured [38]. Machine learning methods are currently one of the most effective approaches for analyzing data and extracting a suitable model from the dataset for predicting the results of new experiments [39]. In recent years, the availability of huge data and the improvement of machine learning-based algorithms have caused an unprecedented increase in the use of machine learning methods [40–44].

Algorithms and datasets are two essential tools for research in data science. In this work, the CatBoost algorithm was used to predict the transition temperature of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ cuprate compound and to compare it with the presented experimental results. The CatBoost algorithm is a method based on gradient-boosted decision trees (GBDT) machine learning ensemble techniques. GBDT are an effective tool for solving regression and classification problems in big data [45]. Here, utilizing the CatBoost algorithm and tuning its hyperparameters were done for the first time in this field of research.

To predict the transition temperature of superconducting materials, an algorithm must be trained with a dataset. The SuperCon is now the world's largest and most complete superconducting materials database, which is utilized in the field of methods based on artificial intelligence [46]. After the data preprocessing steps, 16,148 superconducting compounds were considered a dataset. 90% of the dataset was utilized for training the algorithm, while 10% of the dataset, i.e., 1614 compound, was reserved for the testing phase.

The ID and characteristics of each superconducting compound are the features that we make for the compounds; therefore, each compound can be known to the algorithm by its features. Here, 132 features were obtained for each compound using the Magpie package, and then, 25 of the most important features were selected utilizing the CatBoost algorithm's capability [47]. Then, considering that the attributes play a crucial role in predicting the transition temperature, 241 other features were generated for each superconducting compound, and among them, 30 of the most important features were selected using the CatBoost algorithm's capability.

In addition, two attributes were added to complete the set of features: the number of elements in each composition and the sum of each composition's subscript.

Depending on the number of various tasks performed on the dataset and choosing the suitable algorithm, the prediction accuracy of developed models based on machine learning differs from one another. In order to compare the accuracy and efficiency of the developed models, there are evaluation criteria, the two most important of which are: R-squared (R^2) and root mean square error (RMSE) [38]:

$$R^{2} = 1 - \frac{\sum_{i=1}^{m} (y_{i} - \hat{y}_{i})^{2}}{\sum_{i=1}^{m} (y_{i} - \overline{y})^{2}} \quad RMSE = \sqrt{\frac{1}{m} \sum_{i=1}^{m} (y_{i} - \hat{y}_{i})^{2}}$$



Fig. 6 Superconducting transition temperature predicted by the model and compared to the true values (for 1614 superconducting compounds)

Table 2 The transition temperatures predicted by the model for $Y_{1-x}Li_xBa_2Cu_3O_7cuprate$

Compound	$T_{\rm c}$ -prediction(K)
YBa ₂ Cu ₃ O ₇	92.46
Y _{0.99} Li _{0.01} Ba ₂ Cu ₃ O ₇	91.25
Y _{0.98} Li _{0.02} Ba ₂ Cu ₃ O ₇	90.11
$Y_{0.9}Li_{0.1}Ba_2Cu3O_7$	83.86

where *m* represents the number of superconducting compounds, and y_i indicates the true transition temperature of the *i*-th sample which exists in the dataset. \hat{y}_i represents the transition temperature of the *i*-th sample that predicted by the model, and \bar{y} is the average value corresponding to the true values of *m* samples. In this study, R² value of 0.949 and RMSE value of 7.59 Kelvin were obtained, which are suitable values comparing with similar studies [47–52], indicating that the model developed in this study is reliable.

Using the model developed in this study, the transition temperature for ten percent of the data set, or 1614 superconducting compounds, is predicted and compared to the true values in Fig. 6. Each point in Fig. 6 represents a superconducting compound's transition temperature.

Finally, the CatBoost algorithm performed the prediction process using 57 features for each compound. The transition temperatures predicted by the model for $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ cuprate compounds are shown in Table 2. It can be seen that T_c is nearly decreasing with Li doping.

More innovations are associated with the model, including:

- using novel features for superconducting materials that were not utilized in prior studies. Some of these features include electron affinity, Pettifor number, magnetic moment, dipole polarizability, thermal conductivity, etc.
- using various of feature selection methods to pick the most important features for model development.
- applying many data preprocessing steps to the dataset.

4 Conclusion

The effects of non-magnetic Li doping at the Y site in $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ (x = 0, 0.01, 0.02, 0.1) on the T_c and J_c values as well as the crystal structure of the compound were investigated. The optimum conditions for the citrategel preparation method were determined. The surface of the samples thus prepared was examined by SEM to find that the grains were closely packed with no porosity, while the size of the largest grain reached 50 µm. Moreover, no variation was observed in J_c values as a result of the optimum Li doping. Rietveld refinement analysis of XRD data showed that the *c* lattice parameter increased with Li doping, while the distance between CuO₂ planes decreased. AC magnetic susceptibility bulk measurements showed that although the Li dopant was non-magnetic and was capable of bearing a high phonon frequency, no remarkable increases were observed in T_c , thereby refuting the hypothesis that T_c increases with phonon frequency. This is consistent with magnetic coupling models and recent observations in cuprate superconductors. The CatBoost algorithm was also used to predict the transition temperature of $Y_{1-x}Li_xBa_2Cu_3O_{7-\delta}$ cuprates. It showed that T_c is nearly decreasing with Li doping consistent with the experimental results. Finally, it was found that improving upon the preparation process would have a profound effect on J_c enhancement (by a factor of 7 compared with the standard solid-state method) with nearly no T_c suppression, which seems promising for further fundamental and applied research. It would be interesting to investigate the effect of lower doping of Li, x < 0.01, on the T_c , J_c and pinning force of $Y_{1,x}Li_x$ -123. The question is worthy to be asked, and further experiments are planned in this regard.

Funding No funding was received for conducting this study.

Declarations

Conflict of interest The authors have no competing interests to declare that are relevant to the content of this article.

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