# The Use of C4.5 Algorithm for Classification of New Student Admissions in Vocational High Schools (Case Study: Piramida Rancaekek Vocational High School)

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#### Abstract

The New Student Admission is a mandatory process conducted by all educational institutions to maintain the continuity of student cohorts within the educational framework. At Vocational High School Piramida, The New Student Admission process adheres to a traditional methodology, resulting in subjective and protracted selection procedures. Consequently, the author employs the C4.5 algorithm as a classification technique that produces decision trees from potential student data. The created system is intended to manage The New Student Admission process with greater efficiency and objectivity. This study utilizes descriptive analysis, and the system development adheres to Object-Oriented Software Engineering (OOSE) principles. The system architecture employs the PHP programming language in conjunction with the Laravel framework and a MySQL database. This study demonstrates that utilizing the C4.5 algorithm can enhance the efficiency and effectiveness of the student selection process. The method enhances information accessibility for prospective students, consequently augmenting its attractiveness to candidates.

Keywords: System, Classification, C4.5 Algorithm, Selection

#### INTRODUCTION

The New Student Admission is a procedure for enrolling students in educational institutions, encompassing both formal and non-formal The New Student Admission is a settings. mandatory procedure that every educational institution must follow without exception. The New Student Admission process functions as an educational ecosystem that preserves the continuity of student cohorts, which progressively evolves.

The guidelines for the execution of The New Student Admission are explicitly defined in the Ministry of Education and Culture Regulation No. 51 of 2018. In the subsequent year, these restrictions were amended in the Ministry of Education and Culture Regulation No. 44 of 2019. This refinement underscores the necessity of this activity for all educational institutions, particularly schools.

Vocational High School Piramida Rancaekek is one of the institutions that implement The New Student Admission procedure following the conclusion of annual examinations each year. The New Student Admission procedure at Vocational High School Piramida Rancaekek follows a conventional method, where prospective students, accompanied by their parents, attend the school to finalize their initial registration, submit the requisite supporting documents, and participate in essential assessments. The outcomes of new student admissions must be more objective and efficient to conserve time, as the existing approaches can yield subjective results and protracted processes.

The New Student Admission system at Vocational High School Piramida Rancaekek utilizes a classification method through the application of the C4.5 algorithm. The C4.5 method generates decision trees from categorization samples. This can be summarised as a decision tree methodology that converts extensive datasets into a decision tree that encapsulates rules.

Prior investigations into data mining in PPDB were conducted by Amelia Yusnita, Siti Lailiyah, and Khoiri Saumahudi, who employed the Naive Bayes method. The researchers determined that the decision support system employing this strategy can facilitate the student selection process in an entirely objective, impartial, and transparent manner in decision-A study was undertaken by Linda making. Monizah Fitriani and Andik Setyono, entitled "Application of the C4.5 Algorithm for New Student Admissions at SD Islam Terpadu Permata Bunda Demak." This study employed RapidMiner software as a modeling tool to build rules for classifying incoming student admissions. The application, designed in the Java programming language, effectively assisted the school in identifying new pupils based on their specific requirements .

#### Data Mining

Data mining is the procedure of gathering and analyzing data to derive important insights. This is accomplished by software that uses computations, statistics, or artificial intelligence (AI). Multiple factors are driving the ongoing evolution of data mining, including the rapid proliferation of data sets, the storage of data in warehouses that enable nearly all companies to access high-quality databases, and the competitive pressures within the business sector to enhance market dominance in the global economy.

# C4.5 Algorithm

The C4.5 algorithm is an enhancement of the ID3 algorithm, which is utilized for constructing CART, C4.5, and ID3 are decision trees. algorithms employed in the construction of decision trees. (Chen, 2021; Tempola et al., 2022) In the work by Anwar (2025), titled "Application of K-Means and C4.5 Algorithms for dropout risk prediction in vocational high schools," Yan (2022) elucidate the fundamental methodology for constructing a decision tree utilizing the C4.5 algorithm. This approach initiates by designating an attribute as the root, thereafter forming branches for each value. Cases are subsequently allocated to these branches, and the process is reiterated for each branch until all cases within a branch are classified into the same category.

The creation of a decision tree using the C4.5 method Suherman et al., (2023) involves multiple stages:

- Compiling training data. Training data often originates from previous data that has been categorized into specific classifications.
- Determining the root of the tree: The supplementary value of each characteristic is computed, with the highest supplementary value designated as the primary root. The supplementary entropy value is subsequently computed using this formula to determine the entropy value:

$$Entropy(S) = \sum_{i=1}^{n} -p_i * log_2(p_i)$$

Description:

- S = Set of cases
- n = Number of partitions S
- pi = Proportion of Si to S
- 3. Determine the gain value with the subsequent equation:

$$Gain(S, A) = Entropy(S) - \sum_{i=1}^{n} \frac{|S_i|}{|S|} * Entropy(S_i)$$

Description:

S = Case set A = Attribute

n = Number of attribute partitions A

|Si| = Proportion of Si to S

|S| = Number of cases in S

- 4. Reiterate steps 2 and 3 until all records are partitioned.
- 5. The decision tree splitting operation will cease when:
  - a. All records in node N belong to the same class.
  - b. The partitioned records no longer include any properties.
  - c. The vacant branches contain no records.

# Classification

The term "classification" derives from the word "classis," which means "to group similar objects" and "to separate dissimilar objects." The classification process is based on four components (Sari & Safii, 2021): Class: This is the dependent variable that is categorical and indicates the "label" of the object. Predictor: This is the independent variable represented by the characteristics (attributes) of the data. Training Dataset: This is a set of data containing values from the two components mentioned above, used to determine the appropriate class based on the predictor. Testing Dataset: This contains new data that the developed model will classify, and the accuracy of the classification is evaluated. Based on the explanations provided, classification is an activity performed to group or select items that share similarities in order to form a common group.

# Framework

According to Yudhanto & Prasetyo, a framework is a collection of libraries or functions that developers may utilize to make it easier for them to code. (Lintang et al., 2022)

#### Laravel

Laravel is an open-source PHP framework distributed under the MIT license and constructed on the Model-View-Controller (MVC) paradigm. (Abrori et al., 2022) It is a PHP-based web development framework that utilizes the MVC architecture, designed to enhance software quality and minimize maintenance costs. Taylor Otwell created Laravel, recognized as one of the premier PHP frameworks. It is an open-source web development framework characterized by its attractive syntax and expressive code. Laravel is engineered to expedite and streamline web development. (Daulay et al., 2022)

#### UML

The Unified Modelling Language (UML) is a dependable modeling tool utilized in the creation of object-oriented systems. (Sulistiani & Aldino, 2020) UML is an amalgamation of modeling languages created by Booch, Object Modelling Technique (OMT), and Object-Oriented Software Engineering (OOSE). This approach renders the analysis and design process iterative, encompassing the identification of classes and objects, the elucidation of the semantics of the relationships among these entities, and the specification of interfaces and implementations (Aji & Sunyoto, 2020; Muhammad & Rahardja, 2021).

The author aims to investigate the implementation of the C4.5 algorithm for classifying new student admissions at Piramida Rancaekek Vocational School.

#### METHOD

## **Research methods**

The author utilized two primary approaches for data collecting in this study: observation and interviews. Observation involves collecting data through scrutiny of subjects in their natural habitat, allowing the researcher to document genuine behaviors and interactions. Conversely, interviews occur when the researcher gathers data by posing questions to various individuals, directing inquiries toward a set of participants to elicit their responses. Collectively, these methodologies offer an exhaustive framework for comprehending the research subject.

# System Development Methods

Object-Oriented Software Engineering (OOSE) is an object modeling methodology established by Ivar Jacobson in 1992. It is the inaugural object-oriented design technique to employ use cases to enhance software design. A use case delineates a scenario illustrating user interactions with the product or system under development. This approach emphasizes use cases. OOSE has three fundamental stages: the development of requirement and analysis models, design and implementation, and model testing. The value of this method lies in its ease of learning, as it encompasses all elements of software engineering and utilizes straightforward notation.

# **RESULTS AND DISCUSSION**

#### C4.5 Algorithm Analysis

It serves for categorization or categorization and possesses predictive characteristics. The C4.5 algorithm is a data mining technique that classifies data, designed to uncover significant patterns within large datasets. The execution of the C4.5 algorithm for ascertaining new student admittance adheres to the following procedures:

1. Literature Study

At this juncture, the concepts requisite for the research will be elucidated, including an exposition of the C4.5 algorithm and its implementation methodology.

2. Data Sample Collection

At this juncture, the procedure for gathering sample data will involve soliciting information regarding potential new students for the 2024/2025 academic year from Vocational High School Piramida Rancaekek. This prospective student dataset encompasses diverse information, including names, originating schools, parental names, selected majors, and more attributes to be analyzed using the C4.5 algorithm. The overall dataset comprises 133 records of prospective pupils. Table 1 below displays several sample data entries that the author will present in this publication.

Prospective Students	Interest-Aptitude Test	Psychometric Test	Numerical Test	Major Interests	Status		
Student 1	Competent	Quite Competent	Quite Competent	TPK	Passed		
Student 2	Competent	Competent	Quite Competent	TPK	Passed		
Student 3	Competent	Competent	Competent	TPK	Passed		
Student 4	Student 4 Competent No Com		Not Yet Competent	ТРК	Did Not Pass		
Student 5	Quite Competent	Quite Competent	Quite Competent	TPK	Passed		
Student 6	Not Yet Competent	Competent	Not Yet Competent	TPK	Did Not Pass		
Student 7	Quite Competent	Competent	Competent	TEI	Passed		
Student 8			i competent competent		Competent	TEI	Passed
		Competent	Competent	TEI	Passed		
Student 10	Competent	Competent	Competent	TEI	Passed		
Student 11	Quite Competent	Not Yet Competent	Not Yet Competent	TEI	Did Not Pass		
Student 12	Not Yet Competent	Quite Competent	Not Yet Competent	TEI	Did Not Pass		
Student 13	Competent	Competent	Quite Competent	LPS	Passed		
Student 14	Competent	Competent	Competent	LPS	Passed		
Student 15	Quite Competent	Quite Competent	Quite Competent	LPS	Passed		
Student 16	Quite Competent	Competent	Quite Competent	LPS	Passed		
Student 17	Competent	Not Yet Competent Not Yet	Not Yet Competent	LPS	Did Not Pass		
Student 18	Student 1CompetentStudent 5Quite CompetentStudent 6Not Yet CompetentStudent 7Quite CompetentStudent 7CompetentStudent 8CompetentStudent 9CompetentStudent 10CompetentStudent 11Quite CompetentStudent 12Not Yet CompetentStudent 13CompetentStudent 14CompetentStudent 15Quite CompetentStudent 16Quite CompetentStudent 17Competent		Not Yet Competent	LPS	Did Not Pass		

 Table 1. Data on Prospective Students for Vocational High School Piramida 2024/2025

# **Data Sample Analysis**

Upon acquiring the prospective student data, the qualities are delineated and computed according to the predetermined parameters to ascertain the entropy and gain values. The property with the maximum gain value is chosen as the root. The subsequent table presents the computations of entropy and gain values for each variable, derived from the data in Table 1, utilizing the C4.5 algorithm for the prospective students of Vocational High School Piramida for the 2024/2025 academic year:

		Number of Cases (S)	Passed (S1)	Failed (S2)	Entropy	Gain
Total		133	80	53	0,970064152	
Aptitude Test						0,516054161
	competent	53	44	9	0,657272978	
	quite competent	51	36	15	0,873981048	

Table 2. Entropy and Gain

		Number of Cases (S)	Passed (S1)	Failed (S2)	Entropy	Gain
	not yet competent	29	0	29	0	
Psychometric Test						0,426514729
	competent	51	36	15	0,873981048	
	quite competent	67	44	23	0,927926262	
	not yet competent	15	0	15	0	
Numerical Test						0,54157362
	competent	36	27	9	0,811278124	
	quite competent	64	53	11	0,661976064	
	not yet competent	33	0	33	0	
Major Interests						0,390806677
	TPK	49	32	17	0,931304369	
	TEI	45	25	20	0,99107606	
	LPS	39	23	16	0,976634911	

The Total Entropy column row in the table above is calculated using the following entropy formula equation:

Entropy (S) = 
$$\sum_{i=1}^{n} -p_i * \log_2(p_i)$$
  
Entropy (Total) =  $\left(-\frac{80}{133} * \log_2\left(\frac{80}{133}\right)\right) + \left(-\frac{53}{133} * \log_2\left(\frac{53}{133}\right)\right)$ 

# Entropy(Total) = 0,970064152

Meanwhile, the Gain value in the Interest and Talent Test row is calculated using the following gain formula equation:

$$\begin{aligned} Gain\left(S,A\right) &= Entropy\left(S\right) - \sum_{i=1}^{n} \frac{|S_i|}{|S|} * Entropy\left(S_i\right) \\ &= Entropy\left(Total\right) - \sum_{i=1}^{n} \frac{|tes_i|}{|Total|} * Entropy\left(tes_initat_bakat\right) \end{aligned}$$

$$\begin{array}{l} Gain \; (Total, tes\_minat\_bakat) \\ = 0,970064152 - \left(\frac{53}{133}*0,657272978\right) + \left(\frac{51}{133}*0,873981048\right) + \left(\frac{29}{133}*0\right) \\ \end{array}$$

Gain (Total,tes\_minat\_bakat) = 0,516054161 The table above indicates that the numeric

test variable will serve as the initial node in the decision group due to its superior gain value. Subsequently, the second node is chosen, and this process continues until an identical calculation is employed to derive the tree selection. The resultant decision tree is as follows:

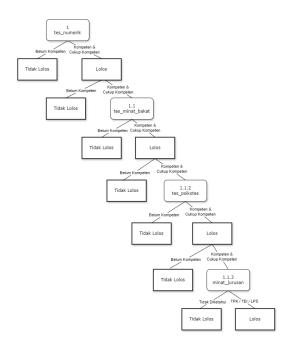


Figure 1. The New Student Admission

# **Classification Decision Tree**

The decision tree results indicate that the quantitative test variable is the most significant factor affecting the graduation of prospective students. This variable serves as the root node of the decision. Consequently, if prospective PPDB applicants achieve elevated numeric exam scores, they will qualify. The scores from the Psychological Test, Aptitude Test, and Major Interest Test are not taken into account.

# Implementation of Website Display

The implementation, utilizing the PHP programming language and the Laravel

framework, has yielded a user-friendly the New Student Admission web page for enrolling pupils aspiring to attend Vocational High School Piramida Rancaekek. This system has been developed to facilitate classification functionalities utilizing the C4.5 Algorithm for the New Student Admission committee and personnel. This facilitates a more equitable approach to entering prospective student data and generating reports.

PPDB SMK Piramida				Home	Login
		Daftar Akun 11 data Anda untuk membuat akun			
	Nama				
	Email				
	Password				
	Confirm Password				
		Deftar			
	Suda	h memiliki akun? Masuk.	J		
,					
	Copyrig	ght 2004   201100597@9tmik-mi.as.id			

Figure 2. Account Registration View



Figure 3. Landing Page View

Login PPDB Online Massik dergan dats yang Anda manukkan saat registraal	
A stringstream	
· · · · · · · · · · · · · · · · · · ·	
Roop no logged in Below newsfilis door! Datus Datas	
Login	

Figure 4. Account Login View



Figure 5. Applicant Dashboard View

	Daftar PPDB		
Selamat Datang di Aplikasi	PPDB SMK Piramida, Silahkan isi formulir dibawah untuk melaku pendaftaran?	kan	
L IDENTITAS PESERTA DIDIK			
NIK			
NISN Nama Lengkap	1		
Jean Kelamin	P+		
Tempet Labir	T		
Tanggol Labir Agama	: dd/mm/syry D		
1.000			
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Email Numor Telepon	1 ()=Ext()=(0)=		
II. SEKOLAH ASAL			
Nama Sekulah / Madrasah			
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Tahun Lulus	4 0000		
Alamat Selected			
Alamat Sekolah	S		
III. IDENTITAS ORANG TUA / Y			
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Nama Lengkap	1		
Pekerjaan			
Alamat			
Bert	N		
Nama Lengkep			
Pekerjasz			
Alamat			
IV. DATA PENDUKUNG	N		
Ninst Jurean	:/Teksh Penhuatas Kain 🔷		
Feto Sizva - Background Merak	Choose File No file shown		
Kirin Reset Formula			

Figure 6. PPDB Form Display for Applicants



Figure 7. Registrant Logout View

The Use of C4.5 Algorithm for Classification of New Student Admissions in Vocational High Schools (Case Study: Piramida Rancaekek Vocational High School)



Figure 8. Confirmation of Registrant Logout

Menu Admin		Lagout
Dashboard	Hello Admin	
Homa	Selamat Datang di menu Admin, silahian eksplorasi menu yang tersediar	
Formulir PPDB		
User Management		
Dashboard Admin		
Data User		
Master Data		
Peserta Didik		
Klasifikasi C4.5		

Figure 9. Admin Dashboard View

Menu Admin				Logo
Dashboard	Selama	t Datang di Halaman	Data User	
Home		data akun calon siswa		
Formulir PPD8	Tamboh Data Us			
User Management	Nama User	Email	Rola	
Dashboard Admin	admin	admin@admin.com	admin 👫	Detete
Data User	panitia	paritia@paritia.com	panitia 👘	Detute
Data Oser	indiani	srinehayu200207@gmail.com	user	Detois
Master Data			_	
Peserta Didik				

Figure 10. User Data Display

Menu Admin		Logo
Dashboard	Tambah Data User	
Home	Nama Ular	
Formulir PPDB		
User Management	Enail	
Dashboard Admin	Persword	
Data User		
Master Data	Konfirmusi Password	
Peserta Didă	Drie	
	admin	
Klasifikasi C4.5	Simper Retail	
	Contactil Contact	

Figure 11. Display of the Add User Data Form

bashboard	S	elamat Da	tang g	li halar	man	Pesert	a Didik			
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faster Data	2	3207972060005428	0028355557	Varya Sharei	e.	94 Lugia	En Industri Nu. 349, Padargi 25562, Bartan	Layonan Perbaskan Sumoh	Rearing User	Ditter
Peserta Didk Klasifikasi C4.5	1	3207153754650267	0020229944	Maida Wijayariti	P	949 МД Награди	Di, Bere Kali Burgar No. 346, Tarakan 50548, Sarnat	Tolesk Doktor Inderki	Resonante Western	Diate
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		3208073678005814	0028841154	Armitia Araggumi		SHPN 1 Ratesekok	An Hajasan Nil. 522, Ciutton 76431, Bali	Takok Dakto Fedarat	Ritus Anda Arggan	Deans
		3207820597585239	0025731575	Diani	e .	parega. Harman	k. Bark No. 335. Berginda (19542) Juliar	Trick Deltsi Islashi	Rites Dans Usanah	Deire

Figure 12. Student Menu Display

#### Laporan Data Siswa

No	NIK	NISN	Nama Siswa	Jenis Kelamin	Asal Sekolah	Alamat	Minat Jurusan
1	3201748008512440	0022778124	Puspa Suartini M.Kom.	Р	SMP BPI Rancaekek	Dk. Laswi No. 412, Lubuklinggau 61483, Lampung	Teknik Elektro Industri
2	3207972060805628	0028305557	Vanya Utami	L	SMP Lugina	Jln. Industri No. 348, Padang 25552, Banten	Layanan Perbankan Syariah
3	3207153754610267	0020229144	Maida Wijayanti	Ρ	SMP PGRI Haurpugur	Dk. Baya Kali Bungur No. 346, Tarakan 58748, Sumsel	Teknik Elektro Industri
4	3204776241313353	0025571711	Okto Budiyanto	Р	SMP PGRI Haurpugur	Jr. Wahid No. 925, Bandar Lampung 88583, Kaltara	Layanan Perbankan Syariah
5	3208060085654089	0027695621	Anastasia Puspa Winarsih	L	SMP PGRI Haurpugur	Jln. Barat No. 579, Makassar 68900, Pabar	Layanan Perbankan Syariah
6	3208175114317937	0024877998	Satya Suwarno M.Farm	L	SMP PGRI Haurpugur	Gg. Padang No. 222, Yogyakarta 40701, Sultra	Layanan Perbankan Syariah
7	3208705292584618	0028236164	Putu Hakim	L	SMP Pasundan Rancaekek	Dk. Panjaitan No. 399, Samarinda 89715, Kalbar	Teknik Pembuatan Kain
8	3200857033699318	0029059202	Gilda Oni Laksmiwati S.E.	L	SMP Bakti Ilham	Ki. Astana Anyar No. 553, Gorontalo 66537, DKI	Teknik Elektro Industri

Figure 13. Display of Student Report Print Results

beard	Se	elamat Datan	g di H	Halaman	Klasifika	asi C4.5		
me	Hala	rrian ini barisi prosas MasiFikusi r	unggunakar	Algoritma C4.5				
muliv PPDB	Da	ataset Klasifikasi						
Management	No	Nama Sirwa	Jenis Kelamin	Nilai Text Minat Bakat	Nilai Test Psikotes	Nilai Test Numerik	Minut Jarasan	Hasil Klasifikasi
shboard Admin	1	Gangsa Budiyanto	L	kompeten	cukup kompeten	cukup kompeten	Layanan Perbankan Syariah	Lolos
ta User	2	Zahra Handayani S.Pd	L	kompeten	Rompeten	kompeten	Teknik Pembuatan Kain	Lolos
	3	Gairran Tamba	L	kompeten	Rompeten	belan kompeten	Teknik Pembuatan Kain	Tidak Lolos
or Data	4	Jatmiko Eja Dabukke S.Farm			cukup kompeten	cukup kompeten	Teknik Pembuatan Kain	Lolos
erta Didik	5	Hari Rahayu	P	kompeten	Nompeten	cukup kompeten	Teknik Pembuatan Kain	Lolos
sifikasi C4.5	6	Kemal Suryona	P	kompeten	cullup kompeten	cukup kompeten	Teknik Elektra Industri	Lolos
	7	Cakrawala Permadi	L		cukup kompeten	belam kompeten	Layanan Perbankan Syariah	Tidak Lolos
	8	Salman Wahyudin	P	cukup kompeten		cukup kompeten	Layanan Perbankan Syariah	Lelos
		Setya Putra	Ł	kompeten	kompeten	kompeten	Layanan Perbankan Syariah	Lalos
	10	Cinthia Aryani	L	cukup kompeten	kompeten	kompeten	Teknik Pembuatan Kain	Lelos
	Nam	ata klasifikasi berhasil disimpan. M Lengkep						
	Pl	ih Nema Sawa						
	Aurie	Relamin						
	NU	Test Minut Bekat						
	Nie	Test Pulcates						
	Nilai	Text Numeric						
	Mina	it Saranam						
		kokan Klasifikasi C4.5						
	Pol	hon Keputusan untuk	Gangsa	Budiyanto				
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	Nita	i Entropy Total : 0,970084152 i Gain Tas Minat Bakat = 0,516 i Gain Tes Psikotes = 0,426514 i Gain Tes Namerik = 0,541573	29					

Figure 14. C4.5 Classification Menu Display



Figure 15. Admin and Officer Logout Display

This is a comparative table assessing the effectiveness of the existing system against the proposed system devised by the author:

Table 3 Comparison	of Effectiveness Levels
--------------------	-------------------------

		Running System		Proposed System	
No	Job description	Y/N	Working Process	Y/N	Working Process
1	Registrationn		60-180 Minute		15 Minute
2	Filling out the Registration Form		30-60 Minute		2-5 Minute
3	The completeness of document		30-60 Minute		2-5 Minute
4	Selection Process		1-2 Sunday		15 Minute

		Running System		Proposed System	
No	Job description	Y/N	Working Process	Y/N	Working Process
5	C4.5 Algorithm Selection Results		-		15 Minute
6	Report		1-4 Sunday		5-15 Minute

#### CONCLUSION

The system has performed satisfactorily, as anticipated. Testing has verified the efficacy of both input and output, particularly in terms of the precision and effectiveness of the C4.5 classification algorithm. Consequently, this system can aid the school in objectively evaluating students as a decision-making support tool in the new student admission process. Furthermore, the report automation feature improves operational efficiency, expediting and streamlining the new student admission process at Vocational High School Piramida.

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