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# Relationships between acute generalized eruptive pustulosis, age, sex, primary disease, and prescription medications: Further analysis of the Japanese Adverse Drug Event Report database

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# ABSTRACT

Background: Acute generalized eruptive pustulosis (AGEP) is a rare, but serious, drug-related adverse event. This study aimed to determine the relationships between AGEP and age, sex, primary disease, and prescription medications using a database of adverse drug events, namely, the Japanese Adverse Drug Event Report (JADER) database.

*Methods*: In this retrospective observational study, we analyzed AGEP reports extracted from the JADER database based on the preferred term for AGEP (code 10048799). We evaluated the effects of causative drugs, underlying diseases, age, and sex. The association between AGEP and prescription drugs was analyzed using the reporting odds ratio and adjusted for covariates using multiple logistic regression. Association rule mining was performed to evaluate the correlation between each combination of factors and AGEP.

*Results*: Between April 2004 and March 2023, 823,662 reports, including 869 reports on AGEP, were entered in the JADER database. The highest percentage of reports in each age group was in males aged <10 years, and clarithromycin was the most used drug in males aged <10 years. Nasopharyngitis was the most common reason for use, and Kawasaki disease was reported as a reason for use among males under 10 years of age, but not among females.

Conclusions: In boys aged <10 years, attention should be paid to the occurrence of AGEP when prescribing clarithromycin and treating Kawasaki disease.

# 1. Introduction

Acute generalized eruptive pustulosis (AGEP) is a severe cutaneous adverse event (AE) classified as a T cell-associated neutrophilic inflammatory reaction whose etiology has not been fully established [1]. The initial phase involves the stimulation and migration of drug-specific T cells into the skin. These T cells, together with natural killer T cells and natural killer cells, are activated in the skin and induce keratinocyte apoptosis through interactions between cytotoxic proteins and Fas/Fas ligands, forming subclassified vesicles. Various cytokines and chemokines are released from innate immune cells, acquired immune cells, and resident cells (keratinocytes, dendritic cells, mast cells, and neutrophils) that primarily cause neutrophilic inflammation and pustule formation [1]. AGEP is associated with drug intake in >90% of the cases, with the most common drugs being antibiotics, primarily beta-lactams and macrolides. Other causative agents include aminoglycosides, sulphonamides quinolones, antifungals, antimalarials, acetaminophen,

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https://doi.org/10.1016/j.heliyon.2024.e27800

Received 7 November 2023; Received in revised form 8 February 2024; Accepted 6 March 2024

Available online 15 March 2024

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nonsteroidal anti-inflammatory drugs (NSAIDs), calcium channel blockers, and antiepileptic drugs [2]. AGEP can occur within hours or days of ingesting the causative drug if the patient is already sensitized to it or 1–2 weeks after taking the drug if the patient had not taken the drug before [3].

AGEP is a rare AE with an estimated incidence of 1–5 cases per million patients per year and a mortality rate of <5% [1,4]. In general, AGEP spontaneously resolves within 2 weeks after discontinuation of the causative agent and has a favorable prognosis. However, organ involvement, mucosal involvement, secondary infection, delayed diagnosis, and comorbidities can lead to poor prognosis [5]. Therefore, healthcare professionals should be aware of the risk of developing AGEP. Therefore, early monitoring and appropriate intervention by healthcare providers are important. However, few studies have examined the relationships between AGEP and age, sex, primary disease, and prescription drugs. Therefore, analysis of the AGEP incidence profile using various factors as axes of analysis is justified.

Large-scale spontaneous reporting systems (SRSs) for AEs are important sources of information for post-marketing drug safety surveillance. Pharmaceutical regulatory authorities in various countries have established systems to record AEs voluntarily reported by drug manufacturers and distributors, healthcare professionals, and general users and to utilize them for safety monitoring. The Japanese Adverse Drug Event Report (JADER) is a database of drug reactions suitable for investigating rare AEs.

Several pharmacovigilance indices, such as reporting odds ratio (ROR), have been developed for drug-associated AEs [6]. The ROR is a powerful technique that allows for adjustments through multivariate logistic regression analysis and the partial control of confounding factors [7,8]. Time-to-onset analysis using the Weibull shape parameter (WSP) is useful in detecting AE signals [9].

### Table 1

Number and reporting ratio of reported acute generalized eruptive pustulosis (AGEP<sup>a</sup>) under 10 years of age.

	Drugs	Case reports (%)
Male	Cases in male AGEP <sup>a</sup> patients under 10 years of age	113 (100.0)
	Clarithromycin	19 (16.8)
	Aspirin	13 (11.5)
	Acetaminophen	11 (9.7)
	Famotidine	9 (8.0)
	Cefotaxime sodium	7 (6.2)
	Azithromycin hydrate	6 (5.3)
	Cefcapene pivoxil hydrochloride hydrate	6 (5.3)
	L-Carbocisteine	4 (3.5)
	Flomoxef sodium	4 (3.5)
	Pranlukast hydrate	4 (3.5)
	Combination cold remedy (OTC <sup>b</sup> Drugs)	4 (3.5)
	Pabron	3 (2.7)
	Tipepidine hibenzate	2(1.8)
	Freeze-dried sulfonated human normal immunoglobulin	2 (1.8)
	Amoxicillin hydrate	2 (1.8)
	Cyprohentadine hydrochloride hydrate	2(1.8)
	Chlorpheniramine maleate	2(1.8)
	Cefmetazole sodium	1 (0.9)
	Tulobuterol	1 (0.9)
	Amovicillin hydrate/Potassium clavulanate	1 (0.9)
	Carbinovamine maleate	1 (0.9)
	Achirin and dialuminate	1 (0.9)
	Antipyretic analgesic anti-inflammatory agent (OTC <sup>b</sup> Druge)	1 (0.9)
	Freeze-dried polyethylene glycol treated human normal immunoglobulin	1 (0.9)
	(Dr)	1 (0.9)
	Non purin cold remedy	1 (0.9)
	Others	4 (2.5)
Formala	Cases in female ACED <sup>a</sup> nationts under 10 years of age	F0 (100 0)
remute	Amoviaillia hudrata	50 (100.0) 6 (12.0)
	Coilealaishisa	5 (12.0)
	Saikokeisiiito	5 (10.0)
	Iranexamic acid	5 (10.0)
	Dextrometnorphan hydrobromide hydrate	5 (10.0)
		5 (10.0)
	L-Cardocisteine	5 (10.0)
	Clarithromycin	4 (8.0)
	Acetaminophen	3 (6.0)
	Antibiotics-resistant lactic acid bacteriae	2 (4.0)
	Tosufloxacin tosilate hydrate	2 (4.0)
	Cefcapene pivoxil hydrochloride hydrate	2 (4.0)
	Cetotaxime sodium	2 (4.0)
	Clemastine fumarate	2 (4.0)
	Piperacillin sodium	1 (2.0)
	Ampicillin sodium	1 (2.0)

<sup>a</sup> Acute generalized eruptive pustulosis.

<sup>b</sup> Over the counter.

Association rule mining (ARM) has been proposed as a new analytical approach to identify undetected associations between variables in large databases, such as potential risk factors [10,11]. This algorithm has been used to assess the associations between the AEs in the JADER database [11,12]. Here, we used pharmacovigilance analysis to determine the expression profile of AEGP and influencing factors, such as age, sex, comorbidities, and prescribed drugs.

# 2. Methods

## 2.1. Data source

The JADER database, containing data from April 2004 to March 2023, was obtained from the Pharmaceuticals and Medical Devices Agency website (www.pmda.go.jp). The structure of the JADER database is based on International Safety Reporting Guidelines. The database comprises four tables: 1) case list (demo), 2) drug information (drug), 3) drug reaction information table (reac), and 4) primary disease (hist). As the JADER database does not contain codes for identifying case reports (A1.11), we could not exclude duplicate case reports from the same patient (www.pmda.go.jp/files/000145474.pdf). Age was recorded in a "demo" table, which included patient demographic data. The following age-related items are entered in the "demo" table: <10, 10–19, 20–29, 30–39, 40–49, 50–59, 60–69, 70–79, 80–89, 90–99, or  $\geq$ 100 years. In the drug table, the causality of each drug was assigned a code according to its association with AEs, such as "suspected drug," "concomitant drug," or "interacting drug." In this study, data coded as "suspect drugs" were analyzed. Data from duplicate drugs were excluded from analysis.

# 2.2. Definition of AEs

AEs in the JADER database were defined as codes according to the terminology used in the Medical Dictionary for Regulatory Activities/Japanese version 23.1 (www.pmrj.jp/jmo/php/indexj.php). We used the preferred terms for AGEP (code 10048799).

## 2.3. Reporting percentage

To calculate the reporting rate, the number of reports in each age group (<10, 10-19, 20-29, 30-39, 40-49, 50-59, 60-69, 70-79, 80-89, and >90 years) listed in the JADER database and the number of reports recorded as AGEP in each age group were used.

## 2.4. Disproportionality assessment via multiple logistic regression analysis

The authorized pharmacovigilance index RORs [6] was calculated using a two-by-two contingency table for the presence/absence of specific drugs and AEs in case reports. The ROR is the ratio of the odds of reporting adverse events versus all other events associated

#### Table 2

Reasons for drug use in acute generalized eruptive pustulosis (AGEP<sup>a</sup>) patients under 10 years of age.

	Reasons for use	Case reports (%)
Male	Cases in male AGEP <sup>a</sup> patients under 10 years of age	83 (100.0)
	Nasopharyngitis	26 (31.3)
	Kawasaki disease	21 (25.3)
	Fever	10 (12.0)
	Bacterial infection	8 (9.6)
	Sore throat	4 (4.8)
	Upper airway inflammation	2 (2.4)
	Lymphadenitis	2 (2.4)
	Nasal inflammation	1 (1.2)
	Wound	1 (1.2)
	Influenza	1 (1.2)
	Allergic rhinitis	1 (1.2)
	Drug use for unknown indication	4 (4.8)
	Heat treatment	2 (2.4)
Female	Cases in female AGEP <sup>a</sup> patients under 10 years of age	46 (100.0)
	Sore throat	25 (54.3)
	Adenovirus test positive	6 (13.0)
	Fever	3 (6.5)
	Pneumonia	3 (6.5)
	Bacterial infection	3 (6.5)
	Rhinosinusitis	1 (2.2)
	Non-tuberculous mycobacteria	1 (2.2)
	Contact dermatitis	1 (2.2)
	Fasciitis	1 (2.2)
	Prevention of infection	1 (2.2)
	Adenovirus infection	1 (2.2)

<sup>a</sup> Acute generalized eruptive pustulosis.

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with the drug of interest compared with the reporting odds for all other drugs present in the database. Disproportionality analysis was performed by calculating the ROR with corresponding 95% confidence intervals (95% CIs). A signal was considered when the lower limit of the 95% CI exceeded 1. Two or more cases were necessary to positively identify such signals.

Patient demographics and descriptive analyses of the suspected drugs suggested that Kawasaki disease, clarithromycin use, aspirin use, and famotidine use were prevalent factors in boys aged <10 years (Tables 1 and 2). Therefore, we refined the ROR signal using dedicated corrections to detect possible confounders in the database [13,14]. The ROR was adjusted using a multivariate logistic regression model. To calculate the adjusted ROR, only reports with complete information regarding the reporting year, sex, age, Kawasaki disease, aspirin use, and clarithromycin use were extracted from the JADER database. The following formula was used:

# $Log \ (odds) = \beta_0 + \beta_1 Y + \beta_2 S + \beta_3 A + \beta_4 K + \beta_5 Asp + \beta_6 Cla + \beta_7 Fa$

The reporting year (Y), sex (S), stratified ages (<10, 10–19, 20–29, 30–39, 40–49, 50–59, 60–69, 70–79, 80–89, and >90 years) (A), Kawasaki disease (K), aspirin use (Asp), clarithromycin use (Cla), and famotidine use (Fa) were defined as independent variables. The dependent variable was a binary response to the absence or presence of AGEP in each report. To comparatively evaluate the effect of factors, we selected explanatory variables using a stepwise method [15,16] and a significance level of 0.25 (forward and backward). The contributions of the selected variables to the final model were evaluated. A likelihood ratio test was used to assess the influence of explanatory variables. Results were considered statistically significant at p < 0.05.

#### 2.5. Time-to-onset analysis

The time-to-onset duration from the JADER database was calculated from the time of the patient's first prescription until the occurrence of AE. Median duration, quartiles, and WSP were used to evaluate the time-to-onset data. The scale parameter  $\alpha$  of the Weibull distribution determines the scale of the distribution function. A larger-scale value ( $\alpha$ ) stretches the distribution, whereas a smaller scale value shrinks the data distribution. The shape parameter  $\beta$  of the Weibull distribution determines the shape of the distribution function. A larger shape value is indicated by a left-skewed curve, whereas a smaller shape value is indicated by a right-skewed curve. In the SRS analysis, the shape parameter  $\beta$  of the Weibull distribution was used to show the hazards without reference population as follows. When  $\beta$  was equal to 1, the hazard was estimated to be constant over time. If  $\beta$  was >1 and the 95% CI of  $\beta$  excluded the value 1, the hazard was considered to increase over time. Finally, if  $\beta$  was <1 and the 95% CI of  $\beta$  excluded the value 1, the hazard was considered to decrease over time [9].

Data analyses were performed using JMP Pro, version 16 (SAS Institute Inc., Cary, NC, USA).

## 2.6. ARM

The ARM is a useful technique for inferring the relationship between drugs and their possible risk factors [10,11]. Given a set of transactions T (each transaction is a set of items), an association rule can be expressed as X [lhs: left-hand-side, the antecedent of the rule]  $\rightarrow$  Y [rhs: right-hand-side, the consequent of the rule], where X and Y are mutually exclusive sets of items [10,11]. *Support* determines how often a rule is observed in a database; in this case, the rule is a combination of X and Y. *Support* was calculated using the following formula:

# $Support = P(X \cap Y) = \{X \cap Y\} / \{D\}$

where D is the total number of transactions in the database.

The confidence of the association rule demonstrated its strengths and was calculated using the following equation:

*Confidence* =  $P(X \cap Y) / P(X)$ 

Lift represents probability ratio. Given a rule that X and Y occur together with multiples of their two individual probabilities:

$$Lift = P(X \cap Y) / P(X) P(Y)$$

*Lift* evaluates the independence of X and Y, with higher *lift* values indicating stronger relationships. If X and Y are independent, the *lift* is 1. If X and Y are positively or negatively correlated, the *lift* is > 1 or <1, respectively. ARM was performed using the arules package in R software (version 4.1.2). The parameter *maxlen* (i.e., the maximum length of itemset/rule, a parameter in the arules package) is the maximum size of the mined frequent itemsets. To efficiently extract association rules, thresholds for optimized *support*, *confidence*, and *maxlen* were defined depending on factors such as the size of the data, number of items, and research purpose. In this study, we defined the minimum *support* and *confidence* thresholds as 0.00001 and 0.001, respectively, and the *maxlen* threshold was restricted to 5. ARM was performed using the apriori function of the arules package in R version 4.1.2 software [17].

## 3. Results

#### 3.1. Patient demographic

The JADER database contains 823,662 reports published between April 2004 and March 2023. Of these, 869 were AGEP reports. The number of AGEP reports for males and females were 403 (46.4%) and 465 (53.6%), respectively (Fig. 1).

The percentage of AGEP reports in males and females is shown in Fig. 2. According to the 403 reports, the number of AGEP cases reported in males aged <10, 10-19, 20-29, 30-39, 40-49, 50-59, 60-69, 70-79, 80-89, and >90 years were 65 (0.4%), 28 (0.3%), 14 (0.1%), 18 (0.1%), 40 (0.2%), 53 (0.1%), 62 (0.07%), 91 (0.09%), 27 (0.06%), and 3 (0.06%), respectively. Among females, according to the 465 reports, the number of cases reported in the same age groups was 22 (0.2%), 13 (0.1%), 33 (0.2%), 46 (0.2%), 70 (0.2%), 81 (0.2%), 62 (0.09%), 72 (0.09%), 44 (0.09%), and 13 (0.1%), respectively. The number of reported cases was higher in the group of males aged <10 years.



Fig. 1. Flow chart of data analysis.

The mean patient age was 52.1 years. Moreover, the mean ages of male and female patients were 49.6 and 54.3 years, respectively.

# 3.2. Descriptive analysis of suspect drugs

In males who developed AGEP, the top three drugs used were acetaminophen (46/668 cases, 6.9%), amoxicillin hydrate (38/668 cases, 5.7%), and clarithromycin (29/668 cases, 4.3%). Among females, acetaminophen was used in 57/776 cases (7.3%), amoxicillin hydrate in 44/776 (5.7%), terbinafine hydrochloride in 20/776 (2.6%), and loxoprofen sodium hydrate in 20/776 (2.6%).

Table 1 shows the percentage of drugs used in males and females under 10 years of age; clarithromycin was the most used drug in males under 10 years of age in 19/113 cases (16.8%). Of the 19 reports involving clarithromycin, the reason for its use was stated in 13 cases, 12 of which involved nasopharyngitis.

The percentages of the reasons for use in males and females under 10 years of age are shown in Table 2. Nasopharyngitis was reported most frequently in males under 10 years of age, accounting for 26/83 cases (31.3%), whereas Kawasaki disease was reported in 21/83 males (25.3%) under 10 years of age; however, no females reported Kawasaki disease as a reason for use. The most frequently used drugs for treating Kawasaki disease are aspirin and famotidine.

#### 3.3. Disproportionality assessment via multiple logistic regression analysis

The RORs of the 20 most frequently used drugs are shown in Table 3. Signals were detected for all 20 drugs. For aspirin, the signal was detected only in males. For diclofenac sodium, tosufloxacin tosilate hydrate and cefazolin sodium, the signal was detected in females.

Using a stepwise logistic regression model, we examined and selected significant AGEP-related variables among the demographic factors (sex, stratified age, and drug use) and reporting year (Table 4). Famotidine was not incorporated into the model equation because it was not a statistically significant term according to results of the stepwise method. The results in the final model indicated significant contributions to AGEP in the reporting year, stratified age (except for ages 80–89 years), Kawasaki disease and clarithromycin. Adjusted RORs for the analyzed groups were as follows: 1.1 for the female group (95% CI: 0.99–1.3); 3.5 for the <10 years group (2.7–4.5); 7.7 for Kawasaki disease (4.1–14.4); and 6.6 for clarithromycin (5.2–8.5).

## 3.4. Time-to-onset analysis

For the time-to-onset analysis, we extracted combinations for which complete information regarding the dates of treatment initiation and AGEP onset were available. The 20 most frequently used drugs were evaluated (Table 5). The median duration (interquartile range) of AGEP onset was 2.0 (1.0–7.0). The median (inter quartile range) for the onset day of AGEP after treatment with clarithromycin in males were 1.5 (0.3–2.0). The lower limit of the 95% CI of the WSP  $\beta$  value for clarithromycin in males was >1.

## 3.5. ARM

We evaluated the association between AGEP and demographic data. The mining algorithm identified 100 rules for AGEP (Supplementary Table S1). {<10 years, Kawasaki disease, male}  $\rightarrow$  {AGEP} and {Kawasaki disease, male}  $\rightarrow$  {AGEP} demonstrated high *lift* value (Supplementary Table S1 [48, 49]).

# 4. Discussion

## 4.1. Association of AGEP with age and sex

According to the results of the multiple logistic regression analysis, it has been suggested that females reported more AGEP cases



Fig. 2. Reporting ratio of acute generalized eruptive pustulosis (AGEP).

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#### Table 3

Number of reports and ROR<sup>a</sup> of acute generalized eruptive pustulosis (AGEP).

Drugs	Male and Female			Male			Female			
	Total	Case	Crude ROR <sup>a</sup> (95% CI <sup>b</sup> )	Total	Case	Crude ROR <sup>a</sup> (95% CI <sup>b</sup> )	Total	Case	Crude ROR <sup>a</sup> (95% CI <sup>b</sup> )	
Acetaminophen	4552	103	24.7 (20.1-30.4)	2058	46	25.4 (18.6-34.6)	2398	57	22.9 (17.3–30.3)	
Amoxicillin hydrate	2718	83	32.9 (26.1-43.1)	1507	38	28.1 (20.1-39.4)	1566	44	26.4 (19.2-36.1)	
Terbinafine hydrochloride	1553	44	29.0 (21.3–39.5)	742	24	35.1 (23.1-53.3)	761	20	23.3 (14.8-36.7)	
Clarithromycin	3815	44	11.6 (8.5–15.7)	1653	29	18.9 (12.9–27.7)	2086 15 6.2 (3.7–10.3)		6.2 (3.7–10.3)	
Loxoprofen sodium hydrate	8247	39	4.7 (3.4–6.4)	3989	19	4.9 (3.1–7.8)	4217	20	4.1 (2.6-6.4)	
Ampicillin sodium/Sulbactam 1678 35 21.0 (14.9–29.5) sodium		21.0 (14.9–29.5)	952	18	19.9 (12.3–32.0)	700	17	21.4 (13.1–34.9)		
L-Carbocisteine	1788	30	16.7 (11.6–24.1)	784	17	22.8 (14.0-37.3)	993	13	11.3 (6.5–19.7)	
Cefcapene pivoxil hydrochloride		26	13.8 (9.3–20.4)	904	14	16.1 (9.4–27.5)	945	12	10.9 (6.1–19.4)	
Combination cold remedy (OTC <sup>c</sup>	1642	24	14.4 (9.6–21.7)	836	18	22.7 (14.1-36.6)	790	6	6.4 (2.9–14.4)	
Drugs)										
Celecoxib 4364 20 4.4 (2.8–6		4.4 (2.8-6.9)	1493	7	4.7 (2.2–10.0)	2703	13	4.1 (2.4–7.1)		
Cefdinir 761 20 26.1 (16.7-4		26.1 (16.7-41.0)	341	6	18.0 (8.0-40.5)	411	14	30.1 (17.5–51.7)		
Fexofenadine hydrochloride 774 20 25.7 (16.4–40.2)		358	3	8.4 (2.7-26.3)	393	17	38.9 (23.7-63.7)			
Diclofenac sodium 4030 19 4.6 (2.9–7.2)		1841	3	1.6 (0.5–5.0)	2083	16	6.6 (4.0–10.9)			
Diltiazem hydrochloride 606 19 31.3 (19.7–49.7)		31.3 (19.7–49.7)	334	10	31.2 (16.5–59.1)	269	9	29.2 (15.0-57.2)		
Tazobactam sodium/Piperacillin 2369 18 sodium		7.4 (4.6–11.8)	1449	13	9.2 (5.3–16.0)	871	5	4.8 (2.0–11.7)		
Cefditoren pivoxil	Cefditoren pivoxil 999 18 17.7 (11.1–28.4)		435	4	9.3 (3.4–24.9)	548	14	22.4 (13.1-38.3)		
Aspirin	9535	18	1.8 (1.1–2.9)	5633	15	2.7 (1.6-4.5)	3478	3	0.7 (0.2–2.2)	
Tosufloxacin tosilate hydrate	785	17	21.4 (13.1–34.7)	367	0	-	411	17	37.1 (22.6-60.8)	
Cefazolin sodium	1759	17	9.4 (5.8–15.2)	896	1	-	842 16 16.6 (10.0-2		16.6 (10.0-27.5)	
Ceftriaxone sodium hydrate	3358	16	4.6 (2.8–7.6)	1734	6	3.5 (1.5–7.8)	1585	10	5.4 (2.9–10.0)	

<sup>a</sup> Reporting Odds Ratio.

<sup>b</sup> Confidence Interval.

<sup>c</sup> Over the counter.

Table 4	
Multiple logistic regression	analysis.

Variable <sup>a</sup>	Total (n)	Case(n)	Crude ROR <sup>b</sup> (95% CI <sup>c</sup> )	Adjusted ROR <sup>b</sup> (95% CI <sup>c</sup> )	Likelihood ratio test	
					P-value	
Reporting year	-	-	_	$1.04 (1.03 - 1.1)^{d}$	<0.0001 <sup>e</sup>	
<10 years	27478	87	2.9 (2.4–3.7)	3.5 (2.7-4.5)	$< 0.0001^{e}$	
10–19 years	21635	41	1.7 (1.2–2.3)	2.2 (1.6-3.1)	<0.0001 <sup>e</sup>	
20-29 years	26810	47	1.5 (1.2–2.1)	2.1 (1.5-2.9)	<0.0001 <sup>e</sup>	
30-39 years	42030	64	1.3 (1.04–1.7)	1.8 (1.4–2.4)	$< 0.0001^{e}$	
40-49 years	59534	110	1.7 (1.4–2.1)	2.2 (1.7-2.7)	$< 0.0001^{e}$	
50-59 years	93280	134	1.3 (1.1–1.5)	1.8 (1.4–2.2)	$< 0.0001^{e}$	
80–89 years	99377	71	0.6 (0.5–0.7)	0.9 (0.7–1.1)	0.2161	
Sex (female)	385950	465	1.2 (1.04–1.4)	1.1 (0.99–1.3)	0.0633	
Kawasaki disease	861	14	15.9 (9.3-27.1)	7.7 (4.1–14.4)	<0.0001 <sup>e</sup>	
Aspirin	44224	40	0.9 (0.6–1.2)	0.8 (0.6–1.2)	0.2306	
Clarithromycin	10895	75	7.1 (5.6–9.0)	6.6 (5.2–8.5)	<0.0001 <sup>e</sup>	

<sup>a</sup> Significant variables selected with stepwise method.

<sup>b</sup> Reporting Odds Ratio.

<sup>c</sup> Confidence Interval.

<sup>d</sup> Unit odds ratio when the continuous variable changes by only one unit.

 $^{e}~P < 0.05.$ 

than males. In the present study, AGEP occurred in all age groups and was reported more frequently in females than in males. This finding is consistent with those of other studies [18–20]. In the EuroSCAR study, a multinational case-control study, the mean age was 56 years [19]. In a study using the World Health Organization pharmacovigilance database, the mean age was 57.3 years, with females being older than males [20]. In the present study, the overall mean age was 52.1 years, which is slightly lower than that reported in previous studies. The higher mean age for females was consistent with that of a previous study (49.6 years for males and 54.3 years for females).

A more detailed examination revealed that males reported a higher percentage of AGEP cases in the <10 years and teenage age groups than in other age groups. The results of multiple logistic regression analysis indicated that age <10 years influenced AGEP development. However, a higher percentage of males aged 10–50 years reported AGEP than those aged 60–90 years. Unlike for males,

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Time-to-onset profiles of acute generalized eruptive pustulosis of top 20 most frequently used drugs in the Japanese Adverse Event Report database.

Drugs Male and Female				Male				Female				
	Cases <sup>a</sup> (n)	Median (interquartile range, day)	Scale parameter, α (95% confidence interval)	Shape parameter, β (95% confidence interval)	Cases <sup>a</sup> (n)	Median (interquartile range, day)	Scale parameter, α (95% confidence interval)	Shape parameter, β (95% confidence interval)	Cases <sup>a</sup> (n)	Median (interquartile range, day)	Scale parameter, α (95% confidence interval)	Shape parameter, β (95% confidence interval)
Total	572	2.0 (1.0-7.0)	6.3 (5.6–7.0)	0.9 (0.8–0.9)	250	2.0 (1.0-6.0)	6.2 (5.2–7.3)	0.9 (0.8–1.0)	322	2.0 (1.0-7.0)	6.3 (5.5–7.3)	0.8 (0.8–0.9)
Acetaminophen	23	2.0 (1.0-3.0)	2.5 (1.5-3.9)	1.2 (0.8–1.7)	8	2.0 (1.0-3.5)	2.1 (1.1-3.6)	2.0 (0.9-3.7)	15	2.0 (1.0-3.0)	2.7 (1.4-4.9)	1.1 (0.7–1.7)
Amoxicillin hydrate	20	1.0 (1.0–2.8)	2.3 (1.6–3.3)	1.6 (1.1–2.2)	10	1.0 (0.0–1.0)	1.6 (0.9–2.8)	1.6 (0.9–2.5)	10	2.0 (1.0-3.5)	2.9 (1.9–4.4)	1.8 (1.01–2.9)
Terbinafine hydrochloride	15	17.0 (8.0–23.0)	17.0 (12.3–23.0)	1.9 (1.2–2.8)	7	9.0 (8.0–20.0)	6.3 (2.3–16.2)	1.04 (0.5–1.9)	8	21.0 (9.5–29.0)	22.1 (15.6–30.8)	2.7 (1.3–4.8)
Clarithromycin	10	1.0 (0.0-2.3)	2.6 (1.4-4.6)	1.5 (0.8-2.5)	4	1.5 (0.3-2.0)	1.8 (1.2-2.8)	4.6 (1.3–11.4)	6	1.0 (0.0-3.8)	3.0 (1.1–7.9)	1.4 (0.6–2.7)
Loxoprofen sodium hydrate	16	1.0 (0.0–3.8)	3.8 (2.1–6.6)	1.2 (0.7–1.8)	6	2.5 (0.8–5.3)	4.2 (1.9–9.1)	1.5 (0.7–2.6)	10	1.0 (0.0–2.8)	3.4 (1.2–8.8)	1.02 (0.5–1.8)
Ampicillin sodium/ Sulbactam sodium	15	2.0 (1.0–3.0)	2.9 (1.9–4.4)	1.4 (0.9–2.0)	5	1.0 (0.5–1.5)	1.4 (0.9–2.2)	3.0 (1.2–5.6)	10	2.5 (1.0–5.3)	3.5 (1.9–6.2)	1.6 (0.8–2.6)
L-Carbocisteine	13	2.0 (1.0-4.0)	2.9 (1.9–4.3)	1.7 (1.001–2.6)	9	2.0 (1.0–5.0)	3.1 (1.8–5.1)	1.6 (0.9–2.7)	4	1.5 (0.3–2.8)	2.3 (1.2–4.3)	2.7 (0.9–6.1)
Cefcapene pivoxil hydrochloride hydrate	14	2.0 (0.8–4.3)	4.2 (2.3–7.3)	1.2 (0.7–1.8)	5	0.0 (0.0–2.5)	2.8 (0.6–14.0)	1.7 (0.4–4.7)	9	2.0 (2.0–6.5)	3.6 (1.5–8.2)	0.96 (0.5–1.5)
Combination cold re	medy (OTC <sup>D</sup>	Drugs)										
Celecoxib	12	2.0 (0.0–9.8)	6.9 (4.0–11.7)	1.6 (0.8–2.8)	3	6.0 (0.0–9.0)	8.1 (5.2–13.0)	5.9 (1.3–16.0)	9	2.0 (0.0–10.0)	6.3 (2.8–13.6)	1.3 (0.6–2.4)
Cefdinir	7	5.0 (1.0–5.0)	8.0 (2.5–24.1)	0.9 (0.4–1.5)	4	3.0 (0.3–5.0)	4.1 (1.6–10.6)	2.0 (0.6–4.9)	3	5.0 (2.0–34.0)	17.6 (1.3–271.7)	1.01 (0.2–2.7)
Fexofenadine hydrochloride	3	5.0 (5.0–85.0)	-	-	2	5.0 (5.0–5.0)	-	-	1	85.0 (85.0–85.0)	-	-
Diclofenac sodium	9	2.0 (1.5–21.0)	8.6 (2.2–30.9)	0.7 (0.4–1.2)	2	9.0 (2.0–16.0)	-	-	7	2.0 (1.0–26.0)	9.0 (1.4–50.3)	0.7 (0.3–1.2)
Diltiazem hydrochloride	5	4.0 (2.5–9.0)	4.3 (1.2–15.4)	1.03 (0.4–1.9)	2	4.0 (4.0–4.0)	-	-	3	3.0 (2.0–14.0)	6.4 (0.4–113.3)	0.97 (0.2–2.6)
Tazobactam sodium/ Piperacillin sodium	6	2.0 (0.8–5.3)	4.1 (1.4–11.6)	1.1 (0.5–1.9)	3	3.0 (0.0–3.0)	_	_	3	1.0 (1.0–12.0)	_	_
Cefditoren pivoxil	9	3.0 (1.0-3.0)	3.5 (2.0-6.0)	1.5 (0.8-2.3)	2	2.0 (1.0-3.0)	-	_	7	3.0 (1.0-3.0)	3.9 (2.0–7.4)	1.5 (0.8-2.5)
Aspirin	4	0.5 (0.0-3.3)	2.8 (0.6-14.0)	1.7 (0.4-4.7)	3	0.0 (0.0-1.0)	_	_	1	4.0 (4.0-4.0)	_	-
Tosufloxacin tosilate hydrate	2	4.5 (1.0-8.0)	-	-	0	-	-	-	2	4.5 (1.0-8.0)	-	-
Cefazolin sodium	7	1.0 (1.0–13.0)	5.1 (1.6–14.8)	0.9 (0.4–1.6)	0	-	-	-	7	1.0 (1.0–13.0)	5.1 (1.6–14.8)	0.9 (0.4–1.6)
Ceftriaxone sodium hydrate	7	5.0 (0.0–11.0)	11.6 (6.7–19.8)	2.5 (0.98–4.9)	4	5.5 (0.0–15.5)	15.2 (9.4–25.2)	5.5 (1.2–14.9)	3	5.0 (0.0–8.0)	7.1 (4.2–12.2)	5.1 (1.1–13.8)

<sup>a</sup> The records with complete adverse event occurrence and prescription star date were used for the time-to-onset analysis. <sup>b</sup> Over the counter.

the reporting rate for females aged <10 years was not higher than that for other age groups. The highest reporting rate was observed among females in their 40s. The highest reporting rate among males was observed in those aged <10 years. Some studies have suggested that AGEP occurs more frequently in individuals aged >25 years [18]. Some studies have reported that AGEP is rare in children [2]. Thus, the results of our study differ from those of previous studies involving males with AGEP.

## 4.2. Time-to-onset analysis

The median time of drug onset was within 5 days for all drugs, except terbinafine hydrochloride, cefdinir, fexofenadine hydrochloride and ceftriaxone sodium hydrate. (Table 5). Thus, the AGEP reported in the JADER database may represent a case of drug sensitization. However, the lack of detailed information precludes further discussion.

# 4.3. Association of AGEP with prescription drugs

The following drugs have been cited as causative agents of AGEP in other studies: antibiotics (beta-lactams, macrolides, aminoglycosides, quinolones), sulfonamides, antifungals, antimalarials, acetaminophen, NSAIDs, calcium channel blockers, and antiepileptic drugs [2,5,21]. In the EuroSCAR study, the use of pristinamycin (an antistaphylococcal drug used in Europe), aminopenicillin (ampicillin/amoxicillin), quinolone antibiotics, (hydroxy)chloroquine, sulfonamide antibiotics, terbinafine, and diltiazem is strongly associated with AGEP [19]. Many drugs reported in the present study are consistent with those listed above.

Clarithromycin was reported in nearly 17% of males aged <10 years but in  $\leq$ 10% of males of other generations. Clarithromycin is listed as a suspected drug of AGEP in the Ministry of Health, Labour and Welfare's "Serious Adverse Drug Reaction Manual" [22]. In this study, clarithromycin use was most frequently reported in males aged <10 years. Therefore, when prescribing clarithromycin to male patients aged <10 years, clinicians should consider the risk of AGEP. NSAIDs have been reported as drugs suspected to induce AGEP [2]. Therefore, when aspirin is used to treat Kawasaki disease, caution should be exercised against the development of AGEP. Since clarithromycin is not used for the treatment of Kawasaki disease, there is no relationship between the two, and each is considered an independent risk factor for AGEP. Famotidine is prescribed to prevent gastrointestinal disturbances when prednisolone is used to treat Kawasaki disease. Multiple regression analysis suggested that famotidine use had little effect on the occurrence of AGEP.

Previous studies have shown that certain drugs may be associated with severe drug eruptions and human leukocyte antigen (HLA) gene polymorphisms. Carbamazepine is one such drug that has been associated with HLA-B\*15:02, HLA-A\*31:01, and HLA-B\*15:11 [23]. Although no clear mechanism has been elucidated, it is believed that the interaction of major histocompatibility complex molecules encoded by specific HLA alleles with carbamazepine or carbamazepine-derived molecules causes them to be recognized by T cells as "non-self" and trigger an immune response [24]. A previous study suggested an association between HLA-A\*02:07 and clarithromycin-induced skin damage in Chinese patients [25]. This bonding is believed to cause T cells to recognize the drug as "non-self" similar to the case of carbamazepine, triggering an immune response that leads to drug rash. These findings may partially explain the high number of reports describing AGEP in males aged <10 years.

#### 4.4. AGEP and disease in males aged < 10 years

Kawasaki disease primarily occurs in infants under 5 years of age, with boys being at 1.5 times higher risk than girls [26]. In this study, aspirin signaling was detected only in RORs among males, and Kawasaki disease was identified as the reason for aspirin use in males aged <10 years. Clear differences in the incidence of Kawasaki disease have been observed among ethnic groups [26]. The incidence of Kawasaki disease is 2.5 times higher in Asians and Pacific Islanders than in Caucasians, with the Japanese population exhibiting the highest global incidence [26]. According to the ARM analysis, the combination of Kawasaki disease, male sex, and age <10 years showed relatively high *lift* values. Therefore, the incidence of AGEP should be considered when treating males aged <10 years with Kawasaki disease in Japan.

Males under 10 years of age treated with clarithromycin should be cautioned against the development of AGEP. Clinicians should note that aspirin, which is used to treat Kawasaki disease, is a possible causative agent of AGEP. We hope that the data from this study will be useful in the clinical treatment of AGEP.

## 4.5. Limitations

Analyses using SRSs, such as the JADER database, have several notable limitations. Intrinsic problems with SRS data include overreporting, underreporting, missing data, lack of detailed information on the patient's background, lack of a control population or reference group, and the presence of confounding factors. Considering the inherent problems of all SRSs, it should not be used for true risk assessment [6,27]. To date, no widely accepted method has been established to adjust for covariates in studies using datasets from the SRSs. Multiple logistic regression methods can be used to address confounders that affect the reliability of results. The adjusted ROR provides an approximate indicator of signal strength and can be used to formulate hypotheses. In ARM, the researcher determines the parameters (*support, confidence*, and *maxlen*) according to the dataset and research object. In several studies, the values of these parameters varied [10,11]. These parameters are not strict criteria. Further epidemiological studies are required to confirm these findings. The results of the analysis using SRSs should be interpreted with caution, keeping in mind existing clinical results.

In clinical practice, AGEP profiles from real-world post-marketing data have not yet been established. JADER is the primary tool available for pharmacovigilance, as it is the largest and most widely used database globally. We believe the results of the JADER

database analysis to be valid because appropriate analytical methods were used.

## **Ethics** approval

Ethical approval and informed consent were not sought for this study because the study was a database-related observational study without directly involving any research subjects. All results were obtained from data openly available online from the PMDA website (www.pmda.go.jp). All data from the JADER database were fully anonymized by the relevant regulatory authority before we accessed them.

# **Funding statement**

This study was partially supported by the Japan Society for the Promotion of Science (JSPS) KAKENHI (grant numbers: 21K06646 and 21K11100). No additional external funding was received for this study.

## Data availability statement

The datasets used and analyzed during the current study are available from the corresponding author on reasonable request. Data used in this study are available from the following links: https://www.pmda.go.jp/safety/info-services/drugs/adr-info/suspected-adr/0003.html.

## CRediT authorship contribution statement

Moe Yamashita: Writing – review & editing, Writing – original draft, Visualization, Formal analysis, Data curation. Mika Maezawa: Writing – review & editing, Data curation. Wataru Wakabayashi: Writing – review & editing, Data curation. Sakiko Hirofuji: Writing – review & editing, Data curation. Koumi Miyasaka: Writing – review & editing, Data curation. Nanaka Ichihara: Writing – review & editing, Visualization, Formal analysis. Yuka Nokura: Writing – review & editing, Visualization, Formal analysis. Kensuke Matsui: Writing – review & editing, Visualization, Formal analysis. Satoshi Nakao: Writing – review & editing, Data curation. Hiroyuki Tanaka: Writing – review & editing, Writing – original draft, Visualization, Formal analysis. Mitsuhiro Nakamura: Writing – review & editing, Writing – original draft, Visualization, Formal analysis.

## Declaration of competing interest

The authors declare the following financial interests/personal relationships which may be considered as potential competing interests: Mitsuhiro Nakamura reports financial support was provided by Japan Society for the Promotion of Science. If there are other authors, they declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

# Appendix A. Supplementary data

Supplementary data to this article can be found online at https://doi.org/10.1016/j.heliyon.2024.e27800.

#### References

- L. Feldmeyer, K. Heidemeyer, N. Yawalkar, Acute generalized exanthematous pustulosis: pathogenesis, genetic background, clinical variants and therapy, Int. J. Mol. Sci. 17 (8) (2016) 1214, https://doi.org/10.3390/ijms17081214.
- [2] G.A. Lizardo-Castro, G.W. Guillén-Mejía, Acute generalized exanthematous pustulosis. Pustulosis exantemática generalizada aguda, Bol. Med. Hosp. Infant. Mex. 79 (4) (2022) 268–273, https://doi.org/10.24875/BMHIM.21000125.
- [3] S. Sidoroff, S. Halevy, J.N.B. Bavinck, L. Vaillant, J.C. Roujeau, Acute generalized exanthematous pustulosis (AGEP) a clinical reaction pattern, J. Cutan. Pathol. 28 (3) (2001) 113–119, https://doi.org/10.1034/j.1600-0560.2001.028003113.x.
- [4] M. Sussman, A. Napodano, S. Huang, A. Are, S. Hsu, K. Motaparthi, Pustular psoriasis and acute generalized exanthematous pustulosis, Medicina 57 (10) (2021) 1004, https://doi.org/10.3390/medicina57101004.
- [5] A. De, S. Das, A. Sarda, D. Pal, P. Biswas, Acute generalised exanthematous pustulosis: an update, Indian J. Dermatol. 63 (1) (2018) 22–29, https://doi.org/ 10.4103/ijd.JJD 581 17.
- [6] E.P. van Puijenbroek, A. Bate, H.G. Leufkens, M. Lindquist, R. Orre, A.C. Egberts, A comparison of measures of disproportionality for signal detection in spontaneous reporting systems for adverse drug reactions, Pharmacoepidemiol. Drug Saf. 11 (1) (2002) 3–10, https://doi.org/10.1002/pds.668.
- [7] E.P. van Puijenbroek, A.C. Egberts, E.R. Heerdink, H.G. Leufkens, Detecting drug-drug interactions using a database for spontaneous adverse drug reactions: an example with diuretics and non-steroidal anti-inflammatory drugs, Eur. J. Clin. Pharmacol. 56 (9–10) (2000) 733–738, https://doi.org/10.1007/ s002280000215.
- [8] A.C. Egberts, R.H. Meyboom, E.P. van Puijenbroek, Use of measures of disproportionality in pharmacovigilance: three Dutch examples, Drug Saf. 25 (6) (2002) 453–458, https://doi.org/10.2165/00002018-200225060-00010.
- [9] O. Sauzet, A. Carvajal, A. Escudero, M. Molokhia, V.R. Cornelius, Illustration of the weibull shape parameter signal detection tool using electronic healthcare record data, Drug Saf. 36 (10) (2013) 995–1006, https://doi.org/10.1007/s40264-013-0061-7.

- [10] R. Harpaz, H.S. Chase, C. Friedman, Mining multi-item drug adverse effect associations in spontaneous reporting systems, BMC Bioinf. 11 (2010) S7, https://doi. org/10.1186/1471-2105-11-S9-S7.
- [11] P. Yildirim, Association patterns in open data to explore ciprofloxacin adverse events, Appl. Clin. Inf. 6 (4) (2015) 728–747, https://doi.org/10.4338/ACI-2015-06-RA-0076.
- [12] H. Hatahira, S. Hasegawa, S. Sasaoka, Y. Kato, J. Abe, Y. Motooka, et al., Analysis of fall-related adverse events among older adults using the Japanese Adverse Drug Event Report (JADER) database, J Pharm Health Care Sci 4 (2018) 32, https://doi.org/10.1186/s40780-018-0129-8.
- [13] M. Tanaka, S. Hasegawa, S. Nakao, K. Shimada, R. Mukai, K. Matsumoto, et al., Analysis of drug-induced hearing loss by using a spontaneous reporting system database, PLoS One 14 (10) (2019) e0217951, https://doi.org/10.1371/journal.pone.0217951.
- [14] K. Shimada, S. Hasegawa, S. Nakao, R. Mukai, K. Matsumoto, M. Tanaka, et al., Adverse event profiles of ifosfamide-induced encephalopathy analyzed using the food and drug administration adverse event reporting system and the Japanese adverse drug event report databases, Cancer Chemother. Pharmacol. 84 (5) (2019) 1097–1105, https://doi.org/10.1007/s00280-019-03949-5.
- [15] S. Nakao, S. Hasegawa, K. Shimada, R. Mukai, M. Tanaka, K. Matsumoto, et al., Evaluation of anti-infective-related *Clostridium difficile*-associated colitis using the Japanese adverse drug event report database, Int Med Sci 17 (7) (2020) 921–930, https://doi.org/10.7150/ijms.43789.
- [16] K. Shimada, S. Hasegawa, S. Nakao, R. Mukai, A. Sasaoka, N. Ueda, et al., Adverse reaction profiles of hemorrhagic adverse reactions caused by direct oral anticoagulants analyzed using the Food and Drug Administration Adverse Event Reporting System (FAERS) database and the Japanese Adverse Drug Event Report (JADER) database, Int Med Sci 16 (9) (2019) 1295–1303, https://doi.org/10.7150/ijms.34629.
- [17] R. Agrawal, R. Srikant, Fast algorithms for mining association rules in large databases, Proceedings of the 20th International Conference on Very Large Data Bases 7 (1994) 478–499.
- [18] E. Vallejo-Yagüe, A. Martinez-De la Torre, O.S. Mohamad, S. Sabu, A.M. Burden, Drug triggers and clinic of acute generalized exanthematous pustulosis (AGEP): a literature case series of 297 patients, J. Clin. Med. 11 (2) (2022) 397, https://doi.org/10.3390/jcm11020397.
- [19] A. Sidoroff, A. Dunant, C. Viboud, S. Halevy, J.N. Bouwes Bavinck, L. Naldi, et al., Risk factors for acute generalized exanthematous pustulosis (AGEP)-results of a multinational case-control study (EuroSCAR), Br. J. Dermatol. 157 (5) (2007) 989–996, https://doi.org/10.1111/j.1365-2133.2007.08156.x.
- [20] A. Martinez-De la Torre, E. van Weenen, M. Kraus, S. Weiler, S. Feuerriegel, A.M. Burden, A network analysis of drug combinations associated with acute generalized exanthematous pustulosis (AGEP), J. Clin. Med. 10 (19) (2021) 4486, https://doi.org/10.3390/jcm10194486.
- [21] L. Miteva, M. Kadurina, R.A. Schwartz, Childhood acute generalized exanthematous pustulosis induced by oral ketoconazole, Acta Dermatovenerol. Croat. 18 (4) (2010) 267–270.
- [22] Ministry of Health, Labour and Welfare. Serious adverse drug reaction manual: acute generalized eruptive pustulosis. https://www.pmda.go.jp/files/ 000145283.pdf.
- [23] L. Dean, Carbamazepine therapy and HLA genotype, in: V.M. Pratt, S.A. Scott, M. Pirmohamed, B. Esquivel, B.L. Kattman, A.J. Malheiro (Eds.), Medical Genetics Summaries, 2018. Bethesda (MD.
- [24] U. Amstutz, N.H. Shear, M.J. Rieder, S. Hwang, V. Fung, H. Nakamura, et al., Recommendations for HLA-B\*15:02 and HLA-A\*31:01 genetic testing to reduce the risk of carbamazepine-induced hypersensitivity reactions, Epilepsia 55 (4) (2014) 496–506, https://doi.org/10.1111/epi.12564.
- [25] S.A. Chen, L.R. Zhang, F.P. Yang, et al., HLA-A\*02:07 allele associates with clarithromycin-induced cutaneous adverse drug reactions in Chinese patients, Basic Clin. Pharmacol. Toxicol. 123 (3) (2018) 308-313, https://doi.org/10.1111/bcpt.13011.
- [26] S. Agarwal, D.K. Agrawal, Kawasaki disease: etiopathogenesis and novel treatment strategies, Expert Rev Clin Immunol 13 (3) (2017) 247–258, https://doi.org/ 10.1080/1744666X.2017.1232165.
- [27] E. Poluzzi, E. Raschi, C. Piccinni, F.D. Ponti, Data mining techniques in pharmacovigilance: analysis of the publicly accessible FDA Adverse Event Reporting System (AERS), Data Mining Applications in Engineering and Medicine (2012) 265–302, https://doi.org/10.5772/50095. (Accessed 19 January 2024).