

## Investigation of Fatality Probability Function Associated with Injury Severity and Age

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**Abstract** The goal of this study is to develop a fatality probability function associated with injury severity and age, which could provide a useful method to estimate the fatality rate accurately. Seven types of logistic regression models were taken into consideration. The results of the estimations from the 7 logistic regression models were compared to select the best fit regression model by using the data from US accident statistics (NASS-CDS: National Automotive Sampling System Crashworthiness Data System) in year 2001. In addition, the constant coefficients of the model best fit to the year 2001 data were replaced by regression functions as a function of year to develop a model incorporating the effect of the year change. The following results were found: 1) the best fit regression model was a function of maximum AIS (Abbreviated Injury Scale) of each body region and age; 2) the accuracy of the regression model was improved by applying regression functions of the coefficients as a function of the year; and 3) the regression functions of the coefficients show that the fatality rate decreases with each progressing year.

**Keywords** Aging, Fatality Probability Function, Frailty, Injury Severity, Regression Model

### I. INTRODUCTION

In order to estimate the fatality rate in traffic accidents accurately, it is necessary to take into account not only the fragility, which means the probability of injury in a given exposure, but also the frailty, which means the probability of fatality for a given injury. For example, the Statistical Database for road traffic accidents from UNECE (United Nations Economic Commission for Europe) [1] showed that the fatality rate defined as the ratio of fatalities to the sum of fatalities and injuries in the elderly older than 65 years was 5.4%, while that of persons from 25 to 64 years old was 2.6%. Additionally, elderly people tend to result in a higher probability of injury in a given exposure compared to younger people [2]. Furthermore, elderly people tend to result in a higher probability of fatality for a given injury compared to younger people [3-4]. From these situations, it is crucial to establish a methodology to accurately predict probability of fatality that takes into account the age of victims.

One way to predict probability of injuries for a given exposure is a computer simulation using a model for a victim of a traffic accident subjected to a given crash environment. Although it is a useful tool to study the injury mechanism in investigating the effect of passive and active safety technology to reduce fatalities in traffic accidents, it can only predict the probability of injury based on the fragility of the victim. Therefore, estimation of fatality rate based on the estimated probability of injuries from the computer simulation requires clarification of the probability of fatality for given injuries sustained by a victim when calculating the fatality rate using the result of the computer simulation. While there have been many studies focusing on the mechanism and probability of injury (e.g. Petitjean et al. [2] and Takahashi et al. [5]), only a few have focused on a method to predict the probability of fatality with a given injury to be used with the results of the computer simulation.

In order to predict the probability of fatality for given injuries using the result of a computer simulation, the information related to the injury severity, injured body region and age are needed. Earlier studies have investigated the probability of fatality for given injuries. Goertz et al. [6] showed the relationship between the probability of fatality and known maximum AIS (Abbreviated Injury Scale), but they did not consider age and the body region in which the injury with the maximum AIS occurred. Baker et al. [3] showed the relationship between the probability of fatality and ISS (Injury Severity Score) for each age group (0 to 49 years, 50 to 69

years, 70+ years), but did not consider the body region in which the injury with the maximum AIS occurred. Additionally, their results may not be relevant to the current situation of traffic accidents because they conducted the study 40 years ago. Boyd et al. [7] developed the regression function for the probability of survival associated with RTS (Revised Trauma Score) [8], ISS and age. However, most human computer simulation models (e.g. Takahashi et al. [5] and Dokko et al. [9]) cannot predict vital signs, such as blood pressure and respiration rate, which are necessary for determining RTS. Therefore, a methodology for estimating the probability of fatality based on the injury severity and age to be used with the result of a computer simulation is still lacking.

The objective of this study is to develop a fatality probability function associated with injury severity and age, which could estimate the fatality rate accurately. Seven types of logistic regression models associated with injury severity and age were taken into consideration to find the best fit regression model by using the data from US accident statistics (NASS-CDS: National Automotive Sampling System Crashworthiness Data System) [10] in year 2001. The numbers of fatalities in other years were then estimated by applying the best fit regression model and its coefficients determined against the data for 2001 to each of the datasets from year 2002 to 2010, and were compared to the actual numbers of fatalities for the corresponding years, to clarify the applicability of the model to the dataset in different years. In addition, the constant coefficients of the model best fit to the year 2001 data were replaced by regression functions as a function of year to develop a model incorporating the effect of the year change.

## II. METHODS

### *Variables for Regression Model*

In order to investigate fatality probability functions as a function of injury severity and age, thirteen variables were defined based on the relevant variables from NASS-CDS [11] and body region code (first digit of AIS code [12]). Table 1 summarizes the thirteen variables (summary of thirteen variables can be found in the APPENDIX). The data including “unknown” were excluded from this study.

TABLE 1  
DEFINITION OF THIRTEEN VARIABLES

Variables	Range	Definition
<i>DEATH</i>	0 or 1	0: “Time to Death” defined in [11] was equal to 0, 1: other
<i>AGE</i>	0 to 99	“Age of Occupant” defined in [11]
<i>ISS</i>	1 to 75	“Injury Severity Score” defined in [11]
<i>MAIS</i>	1 to 6	“Maximum Known Occupant AIS” defined in [11]
<i>MAIS<sub>1</sub></i>	0 to 6	Maximum AIS in body region 1 (first digit of AIS code [12])
<i>MAIS<sub>2</sub></i>	0 to 6	Maximum AIS in body region 2 (first digit of AIS code [12])
<i>MAIS<sub>3</sub></i>	0 to 6	Maximum AIS in body region 3 (first digit of AIS code [12])
<i>MAIS<sub>4</sub></i>	0 to 6	Maximum AIS in body region 4 (first digit of AIS code [12])
<i>MAIS<sub>5</sub></i>	0 to 6	Maximum AIS in body region 5 (first digit of AIS code [12])
<i>MAIS<sub>6</sub></i>	0 to 6	Maximum AIS in body region 6 (first digit of AIS code [12])
<i>MAIS<sub>7</sub></i>	0 to 6	Maximum AIS in body region 7 (first digit of AIS code [12])
<i>MAIS<sub>8</sub></i>	0 to 6	Maximum AIS in body region 8 (first digit of AIS code [12])
<i>MAIS<sub>9</sub></i>	0 to 6	Maximum AIS in body region 9 (first digit of AIS code [12])

### *Investigation of the Best Fit Regression Model*

Figure 1 and Figure 2 show the fatality rate from NASS-CDS as a function of the age group and the body region for each MAIS (Maximum AIS), respectively. From these two graphs, the following were found: 1) the fatality rates for MAIS 6 and MAIS 1 are almost constant in all age groups and body regions; 2) although the

fatality rate for MAIS 2 in the thorax is high compared to that for other body regions, it is almost constant in all age groups; 3) the fatality rates for MAIS 5 and MAIS 3 increase with age and are different between body regions; and 4) although the fatality rate for MAIS 4 is almost constant in all body regions, it does not increase continuously with age. In this study, therefore, 7 types of logistic regression models were taken into consideration in order to confirm the effect of the information about age and injured body region, respectively. They were: 1) 2 types of simplified regression models, which only include MAIS or ISS; 2) 1 type of regression model, which only includes the information about the injured body region for MAIS; 3) 2 types of regression models, which only include the information about age with MAIS and ISS, respectively; and 4) 2 types of regression models, which include the information about age and injured body region for MAIS. Table 2 shows the 7 types of logistic regression models.

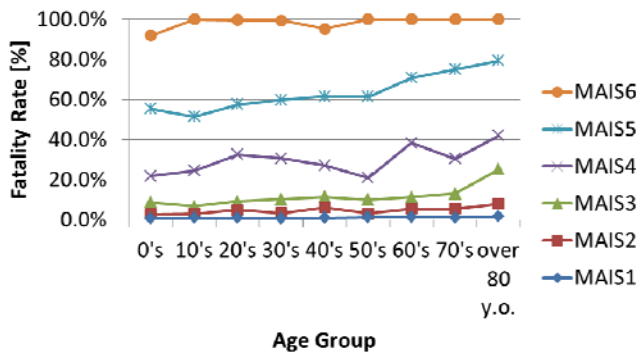


Fig. 1. Fatality rate for each MAIS by age group (data from NASS-CDS from year 2001 to 2007).

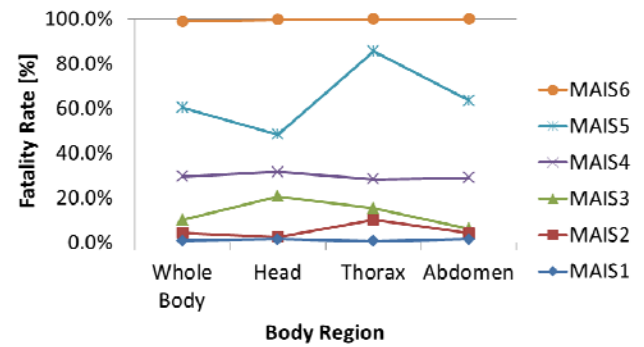


Fig. 2. Fatality rate for each MAIS by body region (data from NASS-CDS from year 2001 to 2007).

TABLE 2  
CANDIDATE FOR THE BEST FIT REGRESSION MODEL

ID	Regression Model
A	$p = \frac{\exp(aMAIS + b)}{1 + \exp(aMAIS + b)}$
B	$p = \frac{\exp(aISS + b)}{1 + \exp(aISS + b)}$
C	$p = \frac{\exp(\sum_{i=1}^9 a_i MAIS_i + b)}{1 + \exp(\sum_{i=1}^9 a_i MAIS_i + b)}$
D	$p = \frac{\exp(aMAIS + bAGE + c)}{1 + \exp(aMAIS + bAGE + c)}$
E	$p = \frac{\exp(aISS + bAGE + c)}{1 + \exp(aISS + bAGE + c)}$
F	$p = \frac{\exp(\sum_{i=1}^9 a_i MAIS_i + bAGE + c)}{1 + \exp(\sum_{i=1}^9 a_i MAIS_i + bAGE + c)}$
G	$p = \frac{\exp(\sum_{i=1}^9 a_i MAIS_i^2 + bAGE^2 + c)}{1 + \exp(\sum_{i=1}^9 a_i MAIS_i^2 + bAGE^2 + c)}$

where  $p$  is probability of fatality, and  $a$ ,  $b$ ,  $c$  and  $a_i$  are the coefficients of the regression model.

The coefficients of each regression model were optimized to the data from NASS-CDS in year 2001 by the maximum likelihood estimation. The regression model with the best fit was chosen, based on the Akaike

Information Criterion (AIC) and Root Mean Square Error (RMSE). The AIC assesses the likelihood of the model and takes into account the number of variables used in the model. The lowest AIC and RMSE indicate the best fit model. The AIC and RMSE are defined by the following equations:

$$AIC = -2 \times \text{LogLikelihood} + 2 \times \text{Number of Variables}, \quad (1)$$

$$RMSE = \sqrt{\frac{\sum (DEATH - \text{Probability of Fatality})^2}{n}}, \quad (2)$$

where  $n$  is the number of occupants in the dataset of NASS-CDS. Additionally, a chi-square test was conducted only for the best fit regression model to confirm the conformity degree of the prediction.

Furthermore, the numbers of fatalities estimated by applying the best fit regression model determined from the year 2001 data to each dataset from year 2002 to 2010 were compared to those of actual fatalities to clarify the applicability of the model to the dataset in different years.

### ***Investigation of Coefficients of Regression Model***

Since the result of the comparison between the actual and estimated fatalities showed the low applicability of the model to the dataset in different years, regression functions for replacing the constant coefficients of the model best fit to the year 2001 data were investigated. The constant coefficients of the best fit regression model determined from the year 2001 data were optimized for each dataset of NASS-CDS from year 2002 to 2010. Regression functions as a function of the year for replacing each constant coefficient were then determined by using the coefficients optimized for each year. Furthermore, RMSE and the number of fatalities were compared between 2 estimations. One was estimated from the regression model best fit to the year 2001 with the coefficients from the year 2001 data by applying the regression model best fit to the year 2001 to the year 2011 data. The other was estimated from the regression model best fit to the year 2001 with the coefficients calculated from the regression functions for replacing the constant coefficients by applying the regression model best fit to the year 2001 to the year 2011 data.

## **III. RESULTS**

### ***Investigation of the Best Fit Regression Model***

Figure 3 shows the comparison of AIC and RMSE for each regression model from the result of optimization. AIC and RMSE of the regression model G were lowest among all of the regression models (AIC: 1332.7 and RMSE: 0.177). From the results, the best fit regression model is G. Additionally, Figure 4 shows the comparison of the actual fatalities and the estimated fatalities by regression model G which confirms that the distribution of the estimated fatalities is similar to that of the actual fatalities. Furthermore, as a result of chi-square test, there was no significant difference between the distribution of the actual fatalities and that of the estimated fatalities calculated from the best fit regression model G ( $p=0.67$ ), which means the conformity degree of distribution of fatalities by the best fit regression model is high.

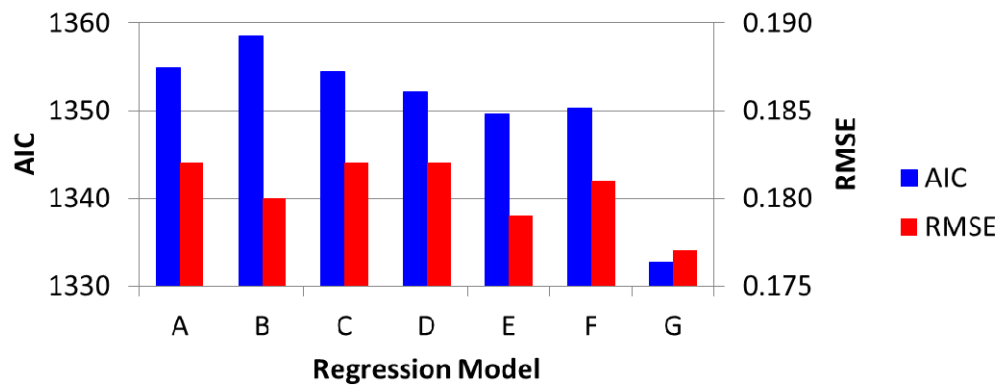


Fig. 3. Comparison of AIC and RMSE among regression models

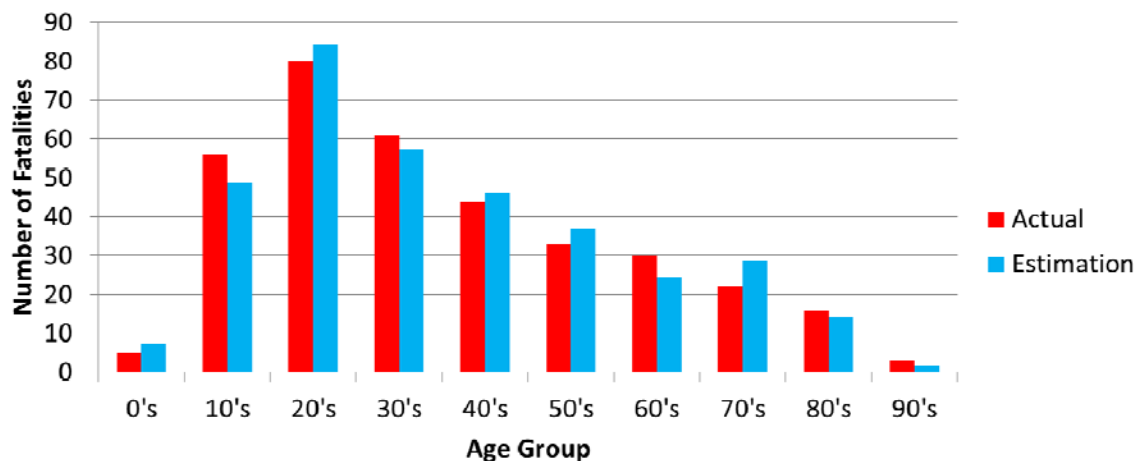


Fig. 4. Comparison of actual fatalities and fatalities estimated by regression model G

Table 3 and Figure 5 show the comparison of the actual fatalities and the fatalities estimated by applying the best fit regression model determined from the year 2001 data to each dataset from year 2002 to 2010. From Figure 5, the degree of overestimation of the estimated fatalities to the actual fatalities increased with each progressing year.

TABLE 3

COMPARISON OF ACTUAL FATALITIES AND FATALITIES ESTIMATED BY APPLYING THE REGRESSION MODEL G TO EACH OF DATASET FROM YEAR 2002 TO 2010

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010
Actual	350	378	392	393	413	505	415	320	186	151
Estimated	350.0	385.2	402.7	427.2	417.3	530.2	474.9	417.0	239.4	183.0
Estimated/ Actual	1.00	1.02	1.03	1.09	1.01	1.05	1.14	1.30	1.29	1.21

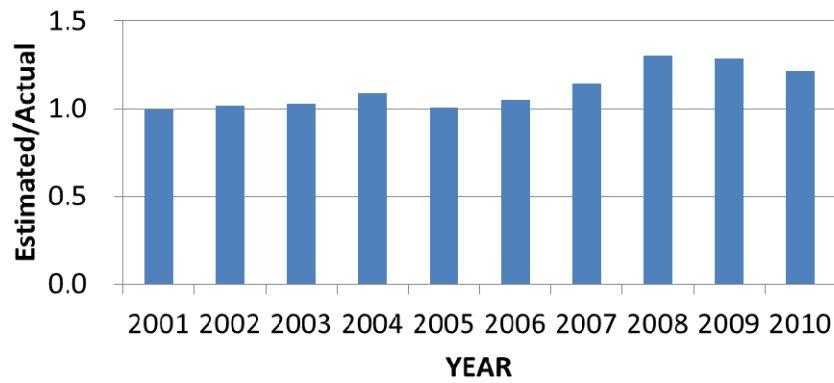


Fig. 5. Ratio of estimated fatalities calculated from the regression model G to actual fatalities for each year

#### Investigation of Coefficients of Regression Model

Table 4 shows the regression function related to the year for the coefficients of the regression model G, as well as the coefficients for the year 2011 determined from the regression function, and the constant coefficients of the regression model G optimized for the year 2001. Table 4 shows that the coefficients  $a_1$ ,  $a_5$ ,  $a_7$ ,  $a_9$  and  $b$  had a positive correlation while other coefficients had a negative correlation with each progressing year.

TABLE 4  
REGRESSION FUNCTION FOR COEFFICIENTS OF REGRESSION MODEL G, ESTIMATED COEFFICIENTS FOR YEAR 2011 FROM REGRESSION FUNCTION AND CONSTANT COEFFICIENTS OF REGRESSION MODEL G OPTIMIZED FOR YEAR 2001

Coefficients	Regression function for replacing the constant coefficients		Estimated for 2011 from the regression function	Constant coefficients optimized for the year 2001
	Regression function	Correlation Coefficient		
$a_1$	$1.97E-03 \times YEAR - 3.83E+00$	0.48	1.35E-01	1.19E-01
$a_2$	$-1.87E-03 \times YEAR + 3.76E+00$	-0.13	3.22E-03	4.30E-02
$a_3$	$-6.44E-02 \times YEAR + 1.30E+02$	-0.63	8.85E-02	1.01E+00
$a_4$	$-1.95E-03 \times YEAR + 4.06E+00$	-0.27	1.41E-01	1.50E-01
$a_5$	$3.61E-03 \times YEAR - 7.16E+00$	0.37	9.55E-02	1.07E-01
$a_6$	$-5.26E-03 \times YEAR + 1.06E+01$	-0.66	1.09E-02	9.07E-02
$a_7$	$5.56E-04 \times YEAR - 1.10E+00$	0.04	2.14E-02	2.33E-02
$a_8$	$-1.70E-03 \times YEAR + 3.46E+00$	-0.27	5.15E-02	7.83E-02
$a_9$	$4.36E-03 \times YEAR - 8.55E+00$	0.30	2.16E-01	1.58E-01
$b$	$8.28E-06 \times YEAR - 1.95E-03$	0.52	2.07E-04	1.26E-04
$c$	$-5.90E-02 \times YEAR + 1.14E+02$	-0.57	-5.16E+00	-4.81E+00

where  $a_i$ ,  $b$ ,  $c$  and are the coefficients of the regression model G shown in below.

$$p = \frac{\exp(\sum_{i=1}^9 a_i MAIS_i^2 + b AGE^2 + c)}{1 + \exp(\sum_{i=1}^9 a_i MAIS_i^2 + b AGE^2 + c)}$$

Figure 6 shows the comparison of RMSE from the estimation of the total number of fatalities in 2011 using the regression model G between the constant coefficients optimized for the year 2001 and the coefficients estimated for 2011 from the regression functions (all of the coefficients are shown in Table 4). Additionally,

Figure 7 shows the comparison of the ratio of the total number of fatalities in 2011 estimated by applying the regression model G to the actual total number of fatalities in 2011 between the constant coefficients optimized for the year 2001 and the coefficients estimated for 2011 from the regression functions (all of the coefficients are shown in Table 4). Furthermore, Figure 8 shows the comparison of the actual number of fatalities in 2011, the estimated number of fatalities in 2011 calculated by applying the regression model G with the constant coefficients optimized for the year 2001, and the coefficients estimated for 2011 from the regression functions (all of the coefficients are shown in Table 4) by age group. Figure 6 and Figure 7 show that RMSE and the ratio of the estimated total number of fatalities to the actual total number of fatalities were improved from 0.160 to 0.156 and from 1.20 to 0.92, respectively, when the regression functions for the coefficients were taken into consideration. Additionally, from Figure 8, it was found that the representativeness of the number of fatalities improved especially in 20s, 30s and 50s.

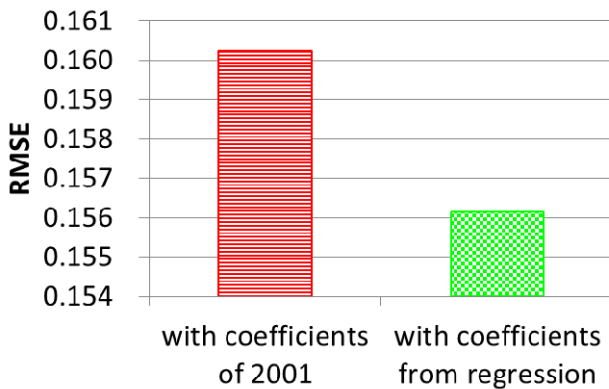


Fig. 6. Comparison of RMSE for year 2011

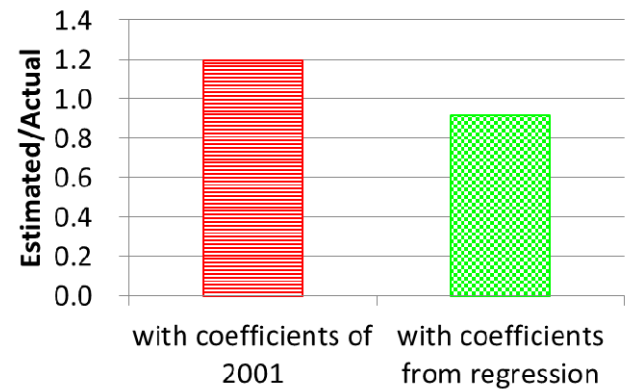


Fig. 7. Comparison of ratio of estimated to actual total number of fatalities in 2011

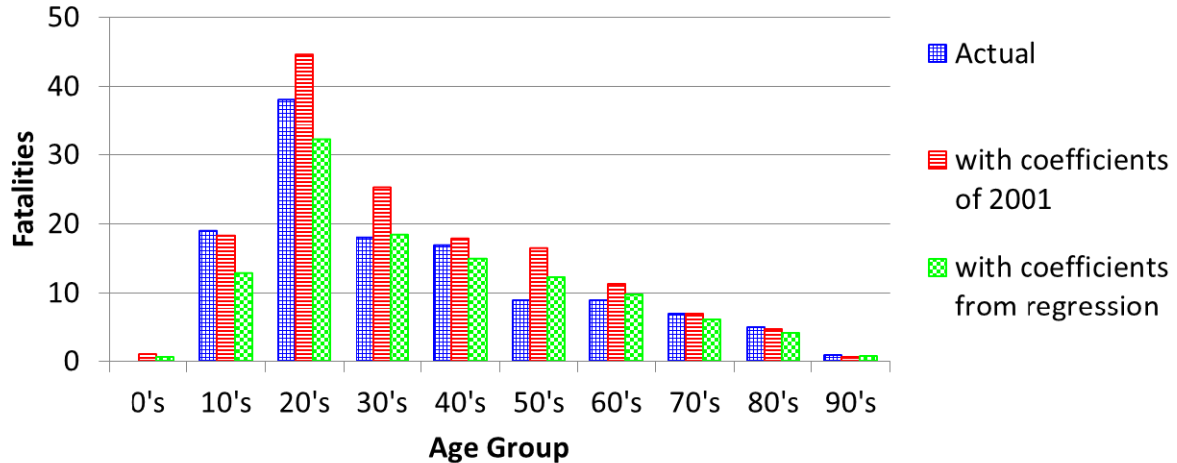


Fig. 8. Comparison of number of fatalities in 2011 by age group

#### IV. DISCUSSION

As shown in Figure 3, comparing the regression models A and D, B and E, and C and F, the differences of which were age, AIC and RMSE of the regression models which included age were lower than those which did not include age. Additionally, comparing the regression models F and G, the difference of which was the degree of variables, AIC and RMSE of the regression model G were lower than those of F. Furthermore, comparing the regression models D and F, and E and G, the differences of which were the consideration of the injured body region, AIC and RMSE of the regression models which considered the injured body region were lower than those which did not consider the injured body region. These results suggest that it is important to consider the influence of age, degree of variables and injured body region when estimating the probability of fatality accurately. This can explain why the best fit regression model is the regression model G in Table 2.

As shown in Table 4 and Figures 6 through 8, the regression model G with the estimated coefficients from the regression functions for 2011 shown in Table 4 can predict the fatalities more accurately than that with the constant coefficients optimized to the year 2001. This result suggests that the coefficients of the regression model are significantly dependent upon each progressing year.

From Figure 7 and Table 4, it can be said that the coefficients for the regression model G change with year progress and that the probability of fatality decreases even if injury severity and age do not change. In order to investigate the factors for the change of the coefficients of the regression model, the year change of the fatality rate in hospitals was checked. Zimmerman et al. [13] showed the year change of the fatality rate in intensive care units in the U.S. Figure 9 shows the year change of fatality rate.



Fig. 9. Fatality rate for 482,601 intensive care units from 2001-2003 to 2010-2012 from the result of Zimmerman et al. [13]

From Figure 9, the fatality rate in intensive care units decreases with the progressing year. This tendency of year change of fatality rate is similar to the result of the year change of the coefficients. Furthermore, Zimmerman et al. [13] cited that the decrease in fatality rate might be attributable to improvements in quality of care. It can be presumed that the fatality rate in a hospital is related to the medical technology and medical infrastructure. Although the change of the coefficients of the regression model is affected by various factors, it is possible that one of the factors which changes the coefficients of the regression model is the improvement of medical technology and medical infrastructure. This suggests that the coefficients of the regression model should be changed when factors such as the country or year change to reflect current medical technology and medical infrastructure.

In this study, since the fatality probability function was developed using the accident data from a single country, additional parameters related to different environments may need to be taken into consideration to develop a more comprehensive regression model.

Additionally, while this study focused on the fatal outcome, there are many non-fatal outcomes, such as disability and impairment. A similar kind of regression model to predict these non-fatal outcomes may be identified in the future by considering such information about disability as described, for example, in the Functional Capacity Index (FCI) [14].

## V. CONCLUSIONS

In this study, seven types of regression models were investigated to identify the best fit fatality probability function associated with injury severity and age relative to US accident statistics (NASS-CDS). Additionally, regression analyses for the coefficients of the regression model were conducted to develop a model incorporating the effect of the year change. The followings results were found:

- The best fit fatality probability function was the maximum AIS of each body region and age.
- The overestimation of the fatalities calculated from the regression model to the actual fatalities increased with each progressive year.
- The accuracy of the regression model was improved by applying the regression functions of the



coefficients as a function of the year.

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## VII. APPENDIX

### APPENDIX 1

#### NUMBERS OF AVAILABLE CASES AND UNKNOWN CASES FOR EACH YEAR

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
Available	5567	6297	6426	7102	5939	6405	6090	6006	3703	3203	2663
Unknown	88	76	87	59	66	92	116	70	60	91	65
Total	5655	6373	6513	7161	6005	6497	6206	6076	3763	3294	2728

### APPENDIX 2

#### NUMBERS OF FATAL CASES AND NON-FATAL CASES FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
Fatal	5217	5919	6034	6709	5526	5900	5675	5686	3517	3053	2540
Non-Fatal	350	378	392	393	413	505	415	320	186	150	123

## APPENDIX 3

## NUMBERS OF CASES BY EACH AGE GROUP FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0s	267	335	278	337	267	276	241	226	141	123	86
10s	1107	1284	1217	1351	1136	1194	1088	1055	518	399	342
20s	1433	1598	1701	1861	1488	1714	1715	1566	1091	840	750
30s	921	1051	995	1147	961	1001	904	903	549	510	421
40s	732	870	921	990	827	839	823	823	494	429	320
50s	456	484	600	620	616	622	582	665	387	394	350
60s	275	292	325	383	317	366	361	416	277	267	216
70s	242	241	249	256	208	254	231	191	147	137	115
80s	123	135	131	142	112	123	131	146	91	95	55
90s	11	7	9	15	7	16	14	15	8	9	8

## APPENDIX 4

## NUMBERS OF CASES BY EACH ISS RANGE FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0-9	4455	5060	5138	5743	4583	4859	4631	4764	3008	2696	2278
10-19	614	646	699	734	708	735	741	635	353	256	187
20-29	212	276	256	268	286	350	295	265	144	120	89
30-39	105	123	141	142	126	166	153	135	75	55	38
40-49	62	73	71	70	82	89	91	69	41	22	23
50-59	45	39	42	58	61	61	75	54	32	20	14
60-69	1	4	7	5	4	11	4	3	4	1	5
70-75	73	76	72	82	89	134	100	81	46	33	29

## APPENDIX 5

NUMBERS OF CASES BY EACH *MAIS* FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
1	3516	4042	4108	4736	3623	3829	3705	3914	2514	2209	1808
2	931	988	1012	972	928	975	877	833	467	479	466
3	633	713	722	780	772	809	799	679	392	277	205
4	233	270	293	305	306	398	367	305	166	126	100
5	181	208	220	228	224	263	242	197	118	80	55
6	73	76	71	81	86	131	100	78	46	32	29

## APPENDIX 6

NUMBERS OF CASES BY EACH *MAIS*<sub>1</sub> FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	3824	4405	4481	5105	4064	4395	4114	4161	2607	2332	1910
1	813	959	912	972	874	932	911	926	573	405	311
2	501	470	518	539	501	453	514	468	260	270	283
3	144	138	153	134	160	187	185	147	71	61	43
4	134	144	172	161	153	205	165	130	78	72	63
5	110	136	150	140	143	158	141	124	84	50	38
6	41	45	40	51	44	75	60	50	30	13	15

## APPENDIX 7

NUMBERS OF CASES BY EACH  $MAIS_2$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	3264	3859	4095	4615	3794	4034	3956	3939	2582	2295	1900
1	2085	2196	2086	2235	1896	2095	1863	1832	976	825	678
2	166	181	183	192	184	217	188	183	101	78	82
3	52	61	61	60	65	59	83	52	44	4	3
4	0	0	1	0	0	0	0	0	0	1	0
5	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0

## APPENDIX 8

NUMBERS OF CASES BY EACH  $MAIS_3$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	5230	5933	6049	6709	5650	6028	5754	5651	3485	3018	2517
1	322	354	354	371	276	346	315	331	207	176	135
2	10	6	15	11	7	20	8	11	4	4	5
3	3	2	6	8	3	9	10	6	4	5	5
4	0	0	2	1	1	0	0	3	0	0	1
5	0	1	0	1	0	0	2	1	0	0	0
6	2	1	0	1	2	2	1	3	3	0	0

## APPENDIX 9

NUMBERS OF CASES BY EACH  $MAIS_4$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	3721	4207	4285	4814	3895	4102	3860	3936	2435	2108	1755
1	1205	1313	1360	1460	1215	1321	1297	1261	843	740	637
2	128	167	155	136	151	159	164	134	82	54	44
3	244	314	326	342	319	346	321	317	157	192	136
4	178	201	200	241	242	336	311	254	132	65	62
5	70	73	77	91	94	113	107	80	45	33	19
6	21	22	23	18	23	28	30	24	9	11	10

## APPENDIX 10

NUMBERS OF CASES BY EACH  $MAIS_5$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	4693	5233	5435	5923	4900	5193	4933	4924	3091	2676	2219
1	553	680	608	746	639	749	730	721	415	374	306
2	196	235	239	277	219	260	239	196	118	90	80
3	55	82	65	86	79	83	78	70	27	26	20
4	47	38	53	44	67	65	69	60	35	28	22
5	19	27	26	25	34	53	37	32	16	9	15
6	4	2	0	1	1	2	4	3	1	0	1

## APPENDIX 11

NUMBERS OF CASES BY EACH  $MAIS_6$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	3953	4508	4461	5018	4117	4474	4118	4062	2498	2261	1874
1	1269	1400	1536	1649	1327	1351	1425	1457	944	710	602
2	229	237	275	261	322	374	319	314	146	153	123
3	67	100	109	113	129	138	161	125	84	54	48
4	6	11	7	13	9	10	17	12	9	4	3
5	28	29	26	27	19	39	30	22	17	14	10
6	15	12	12	21	16	19	20	14	5	7	3

## APPENDIX 12

NUMBERS OF CASES BY EACH  $MAIS_7$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	2762	3183	3287	3643	3060	3295	3096	3037	1863	1595	1349
1	2193	2417	2498	2771	2239	2342	2306	2345	1483	1340	1089
2	453	511	453	476	446	532	459	432	221	222	193
3	159	186	188	212	194	236	229	192	136	46	32
4	0	0	0	0	0	0	0	0	0	0	0
5	0	0	0	0	0	0	0	0	0	0	0
6	0	0	0	0	0	0	0	0	0	0	0

## APPENDIX 13

NUMBERS OF CASES BY EACH  $MAIS_8$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	2718	3102	3326	3668	3048	3206	3181	3131	1890	1721	1452
1	2012	2270	2191	2511	1947	2117	1988	2028	1317	1105	908
2	426	463	453	433	442	501	392	372	227	216	175
3	409	457	446	482	496	563	509	454	261	148	117
4	2	5	9	7	6	13	14	16	7	9	9
5	0	0	1	1	0	5	6	5	1	4	2
6	0	0	0	0	0	0	0	0	0	0	0

## APPENDIX 14

NUMBERS OF CASES BY EACH  $MAIS_9$  FOR EACH YEAR (ONLY AVAILABLE CASES)

Year	2001	2002	2003	2004	2005	2006	2007	2008	2009	2010	2011
0	5430	6161	6282	6977	5815	6250	5948	5861	3593	3099	2569
1	129	123	126	110	110	121	132	141	100	98	87
2	0	0	0	1	0	2	0	0	3	0	1
3	1	1	2	2	2	2	1	0	0	0	1
4	2	0	0	0	1	0	0	0	0	0	0
5	0	5	7	4	2	6	0	1	0	2	0
6	5	7	9	8	9	24	9	3	7	4	5