Casualty Actuarial Society E-Forum, Summer 2019



The CAS E-Forum, Summer 2019

The Summer 2019 edition of the CAS *E-Forum* is a cooperative effort between the CAS *E-Forum* Committee and various CAS committees, task forces, working parties and special interest sections. This *E-Forum* contains one report from the CAS Insurance On Demand Working Party and one independent research paper.

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IS THERE A DEMAND FOR INSURANCE ON DEMAND?

BY STEPHANIE GOULD RABIN IN COLLABORATION WITH JIM WEISS, SCOTT SWANAY, & THE CAS INSURANCE ON DEMAND WORKING PARTY

According to a survey of over 1,000 consumers, consumers:

- Have only a **MODERATE UNDERSTANDING** of insurance pricing and coverage.
- Are more **TRANSACTIONAL** in their review of insurance need.
- **HAVE NEVER HEARD** of the current insurance on demand companies (95% of survey takers).

With this information at hand, one critical question comes to mind: Is there a demand for insurance on demand?

A Working Party on Insurance On Demand, sponsored by the Casualty Actuarial Society (CAS), issued a survey in December 2017 to over 1,000 consumers (demographics are available at the conclusion of this article) to answer some questions around awareness / readiness of the consumer and potential design features for on-demand insurance in the property & casualty realm. Three key aspects of product development were explored in this survey:

- **PEOPLE:** First and foremost, we need to know what the consumer wants and can bear before we build any solutions. How much do our consumers know and understand about their insurance? How do consumers want to interact with their insurance carriers? What kind of insurance do they care most about?
- **PRODUCT:** Today's technology and economy open up the door to more coverages needed by the average consumer. What coverage features should an on-demand product include? Should any of the coverages be unbundled from the rest? Should coverage be tailor made? How long should policies be?
- **PRICE:** Technology allows us to try to reflect or capture the core risk drivers of insurance. And the potential changes in coverage features affords a whole new way to envision pricing for an on-demand product. So, what pricing features would a consumer be willing to have?

This article will look at a number of responses to this consumer survey to help provide more understanding from the consumer point of view.

DEFINITION

Having an understanding of what represents insurance on demand (IOD) is a critical first building block for any research. The actual definition of Insurance On Demand has seen several variations from very narrow to very broad in scope. Based on insurance industry experiences and upon review of the current IOD companies operating today, the CAS IOD Working party agreed to the following definition:

Insurance On Demand is Insurance where the product exhibits client-focus through nontraditional modifications to coverages, pricing, and/or administration.

These modifications, which are often achieved by leveraging modern-day technology, aim to facilitate the purchase of insurance or offer changes in terms that are desired by the policyholder relative to traditional insurance coverages.

- CAS Insurance On Demand Working Party

This definition is still quite broad but encompasses two key pillars: nontraditional features and modernized insurance technology.

CURRENT STATE OF INSURANCE: CONSUMERS SPEAK OUT

Before understanding what consumers want, it's critical to understand the current state of insurance from the consumers' perspective: What is top of mind? What is important? What do they know?

On-demand insurance is attractive to people penalized by traditional insurance products, that is, consumers with low usage that would otherwise have to pay for more coverage than they need. ??

-Matteo Carbone, IoT Insurance Observatory (interview with Josh Taub, FCAS) Those of us in the insurance sector have a rich understanding of the business of insurance and can make many of our insurance decisions with relative ease. What the average consumer knows and understands is critical to determining what their insurance product should look like in terms of access, price, and coverage. The less knowledge, experience, and practice a consumer has, the more likely they will need to put effort into setting up their insurance. And with more effort required, behavioral economic theory notes they will need a prevailing reason to change, which

is consistent what has been noted by Matteo Carbone.

In a survey question asking consumers to rate the level of importance of each listed line of business, top of mind for the majority of consumers is automobile and homeowners insurance (property and casualty lines only). While other lines were noted, consumers, regardless of

industry, age or gender, prioritized these key lines of business. Consumers also noted Life and Health as top lines (despite the study being focused on Property & Casualty). With Homeowners, Auto, as well as Life and Health, consumer responses are often centered on more traditional U.S. insurance lines of business.





Central to exploring consumer readiness for insurance on demand is the frequency with which consumers review their insurance needs. Consumers in this survey were asked how often they reviewed their insurance policy. Nearly 20% of consumers will only review their insurance policies and coverage when reminded. So, one-fifth of consumers likely renew insurance passively by not reviewing terms or price. In fact, only 15% of consumers review their insurance policies and coverage off cycle (i.e., quarterly or monthly). With such limited review, the purchase of insurance today is largely transactional in nature (dealt with at specific periods of time rather than continuously), limiting access points for something like an on-demand structure.

In an interview with Josh Taub, FCAS, a member of the CAS Insurance On Demand Working Party, Scott Walchek from Trōv (an intermediary using an on-demand approach to insurance for personal items) noted that "...consumers are leaving coverage on for much longer than expected, even months at a time."



Consumers self-reported that they had only a moderate understanding of insurance pricing and coverage. With a moderate understanding, combined with a transactional approach to reviewing insurance needs, the

challenge of attracting people to something like an on-demand insurance product becomes much greater.

Consumers were asked whether they had considered purchasing insurance from Airsurety, Cuvva, Lemonade, Metromile, Slice, Sure, Swyfft, Trōv, or Verifly. In fact, consumers have barely heard of the current leading players in the on-demand insurance space. This may be partly due to



marketing challenges as the internet of things grows and typical advertisement techniques become less relevant. Millennials have a slightly better awareness of the current available ondemand insurance solutions. That said, it is still an overwhelming percentage of even Millennials who are unaware of the additional options available. And of those who have heard of the newer companies, several still have not considered moving to these companies; cited reasons include relationships built with their existing carriers, limited track record, and need for additional information.

With all of the above painting quite a grim picture for insurance on demand, what could help to move the dial? Certainly more awareness and access could help. But, what is that prevailing reason for change? The majority of those few who considered the new On-Demand players commented that they did so due to PRICE OF INSURANCE. This is consistent with the majority of consumer responses to a more generalized question on requested improvements in the industry.

PEOPLE: HOW DO WE REACH THEM?

One of the largest hurdles for the On-Demand Insurance arena is access to consumers. As noted above, consumers have barely heard of the current players in the market. And right now, approximately 63% of consumers are using insurance agents to access their insurance (based on a

survey question asking how insurance was purchased). Some online purchasing was noted by consumers, but that alone did not preclude the use of agents with an online platform.

The insurance agent has historically been the source of knowledge and understanding of insurance and risk management needs. In a forced choice question, consumers had to choose their preference when



setting up insurance of either speaking to a person or using an App/Web. Consumers largely preferred to speak to a person versus leveraging the web or an application. So, combining this response with behavioral economics principles, easier transference of knowledge and understanding of individual risk management is likely the key to unlock consumers' willingness to leverage technology to access alternative options for insurance. This would require having technology that could replace what people value in their agent today.

PRODUCT: BUILDING SOMETHING TO RESPOND TO NEEDS

While consumers seem to be more transactional in nature for their insurance needs, there was a larger portion who desired a tailored policy over more of an all-inclusive single policy.... even



more so for the Millennial generation. This interest in a less traditional product (recalling consumers have homeowners, auto, and health insurance at top of mind) may reflect a true need for some specific protections that may or may not be currently available. Additional consumer research would be necessary to assess the reason.

In fact, this could partly play into perception of price...*if I can*

tailor make my policy, then I pay only for what I need. Recall that price does matter most in helping to drive a consumer to change insurance carriers.

Consumers were asked if they would unbundle their current coverages and only *42% OF CONSUMERS WERE WILLING TO SPLIT COVERAGES* out from their current policies. Interestingly, 60% of Business owners, who also self-reported a slightly higher understanding of coverage and price, were willing to unbundle their coverage. Aspects of coverage that consumers were willing to split out include (in order of popularity):

- 1. Accidents
- 2. Personal Liability
- 3. Car Rentals
- 4. Weather-related perils (Flood, Hurricane, Earthquake, Tornado/Hail)
- 5. Specific Items in a home or business

I don't know if I understand insurance to answer this question ³³

-Consumer Survey Response when asked "What coverage(s) would you like to see split out as a separate policy from your current policies?" While unbundling may be a solution attractive to some consumers, the insurance sector should consider both the role of the insurance sector in making people whole again and the awareness of risk need by the consumer. Should an unbundled approach (where consumers pick and choose what they want to pay for) create gaps in coverage, the consumer would ultimately be the one to suffer.

PRICE: THE PREVAILING REASON FOR CHANGE

As noted in the current state of the market, Consumers cited price as one of the prevalent factors in driving them to change insurance carriers. With the growth in ready access to more data and the enhanced abilities for machine learning, predictive analytics, telematics and more, insurance

technology could help to drive changes in pricing to more accurately capture the risk being covered.

Consumers overwhelmingly indicate that they prefer a fixed price over a per use price that may fluctuate. This desire is quite consistent with the timing with which consumers review their policies. If they are not looking at their insurance regularly, having variation in price throughout the



insurance term would be less preferable.



While variable pricing may not be a desired solution by consumers, there is a growing trend among consumers in their willingness to share information in support of pricing and/or coverage support. In fact, Millennials are much more willing to share information than the older generations. Information Millennials were willing to share included: Location, Driving Behaviors, Exercise, Languages, Body Temperature, and Sleep Activity.

While this willingness to share exists, those not willing to share consistently noted the same reason in their comments: concern over data privacy. So, should the alternative pricing solution include a sufficient level of comfort on privacy, a large portion of

those unwilling may become more willing to share information for using in more nonstandard pricing and coverage support.

WHAT'S NEXT?

The consumer responses open up many more questions than answers. And in an age where information flows quickly, many of these responses may shift over time.

Further, while a majority of responses to the survey demonstrated statistically significant difference among the generations, broadening the analysis both longitudinally (over time) and via number of consumers for better breakouts, may help to highlight if there is a trend on the horizon versus a difference purely due to experience.

In the meantime, there are certainly some signs of demand for insurance on demand, and organizations that best address consumer needs may own a valuable call option on the future.

DEMOGRAPHICS



The following provide some detailed demographics for this survey.

Region of United States

East North Central	18%	
East South Central	5%	
Middle Atlantic	14%	
Mountain	8%	
New England	7%	
Pacific	10%	
South Atlantic	21%	
West North Central	8%	
West South Central	8%	
N/A	1%	

46%

Advertising & Marketing	3%
Agriculture	2%
Airlines & Aerospace (including Defense)	1%
Automotive	2%
Business Support & Logistics	4%
Construction, Machinery, and Homes	5%
Education	14%
Entertainment & Leisure	3%
Finance & Financial Services	5%
Food & Beverages	2%
Government	8%
Healthcare & Pharmaceuticals	18%
Insurance	3%
Manufacturing	6%
Nonprofit	6%
Real Estate	2%
Retail & Consumer Durables	5%
Telecommunications, Technology, Internet & Electronics	7%
Transportation & Delivery	2%
Utilities, Energy, and Extraction	3%



R. Stephen Pulis, ACAS, MAAA

Abstract: The following paper is a study of Atlantic windstorms. It reveals that the shape of the distribution of maximum wind speeds is both trending higher and changing its skewed shape. Current estimates of the maximum storm strength in 100 years, etc. are understated as there is an increase in the frequency of more severe wind speeds.

Keywords: Windstorm; Hazard Distribution; Skewness; Weibull modeling.

INTRODUCTION

The recent catastrophic storms raise questions as to the reasonable probability of a 100-year, 500year or 1,000-year storm. A normal distribution indicates that the value at 2.433 standard deviations above the mean includes 99% below this value or represents the equivalent of 1 in 100. Insurance claims, however, do not have negative claims, and therefore the distributions are skewed to the positive side. The hypothesis of this paper is that the distribution of storm frequency and severity has changed over time and these shifts are not uniform.

In this paper, the windstorms in the North Atlantic are investigated. There is a wealth of data available, but, as always, the actuary wishes there was more complete and consistent data. The National Hurricane Center (NHC) is part of the National Oceanic & Atmospheric Administration (NOAA). The NHC has published detailed Atlantic windstorm data¹ since 1851 and is the source of the experience reviewed in this paper.

Experience is selected based on the maximum wind speed measured in 5 knots increments (one knot is approximately 1.15 mph). The reported storms are suspected to be understated for lower wind speeds for older years. From inception, storms with a maximum wind speed of 40 knots or more were reported. Storms with a maximum wind speed of 35 knot storms were not included until 1887, and lesser storms are not included until 1968. This paper only includes storms with at least 40 knot winds. The following graph is for years 1858 through 2017.



The experience is aggregated into 10-year intervals ending with 2008-2017. This results in an average of 51 (38 to 72) major storms (70+ knots) per decade. Higher average storm counts per decade exist for lower minimum wind speeds: 91 for 40+, 78 for 50+, and 63 for 60+.

National Hurricane Center Data Atlantic Windstorm Counts

	Years fr	om/to														
Max	1858	1868	1878	1888	1898	1908	1918	1928	1938	1948	1958	1968	1978	1988	1998	2008
Knots	1867	1877	1887	1897	1907	1917	1927	1937	1947	1957	1967	1977	1987	1997	2007	2017
30	0	0	0	0	0	0	0	0	0	0	0	36	57	23	17	8
35	0	0	1	3	2	1	4	6	6	3	3	0	2	5	10	5
40	1	1	3	1	6	5	8	10	3	5	3	4	2	10	5	10
45	0	0	1	3	14	9	4	9	10	8	10	9	12	13	16	12
50	9	14	10	15	11	7	7	13	10	11	9	5	10	7	19	12
55	0	0	0	2	2	1	3	8	7	3	1	9	6	7	8	17
60	10	4	7	3	8	6	2	10	7	13	5	17	8	7	13	6
65	0	0	0	4	0	5	2	5	1	3	4	9	8	9	10	2
70	18	20	18	6	13	3	8	6	11	10	7	7	7	8	7	5
75	0	0	3	2	2	2	3	3	0	1	5	8	8	8	7	9
80	4	4	11	3	3	2	4	6	3	4	5	5	4	0	7	4
85	0	0	7	16	5	7	2	6	11	4	4	0	4	4	5	5
90	26	15	13	2	4	3	4	5	2	4	1	7	4	3	6	3
95	0	0	0	4	3	4	5	3	4	7	3	3	0	5	2	2
100	0	10	4	3	0	3	1	1	2	6	4	6	3	7	4	3
105	0	0	3	6	6	9	5	1	5	8	2	2	2	2	5	3
110	3	1	5	4	0	0	2	3	1	6	6	4	2	1	3	3
115	0	0	0	2	2	1	1	2	2	6	3	0	5	1	1	4
120	1	0	3	0	0	0	2	4	2	2	2	1	1	3	4	3
125	0	0	0	0	1	2	0	1	3	3	3	0	1	4	6	3
130	0	0	2	0	1	2	3	1	0	2	2	1	0	1	3	0
135	0	0	0	0	0	0	0	2	0	1	2	0	0	0	3	3
140	0	0	0	0	0	0	0	4	1	1	5	1	0	1	1	0
145	0	0	0	0	0	0	1	0	0	0	0	0	0	0	3	0
150	0	0	0	0	0	0	0	1	0	1	1	2	1	1	3	0
155	0	0	0	0	0	0	0	0	0	0	0	0	0	0	1	0
160	0	0	0	0	0	0	0	1	0	0	0	0	0	1	1	0
165	0	0	0	0	0	0	0	0	0	0	0	0	1	0	0	0
Total	72	69	91	79	83	72	71	111	91	112	90	136	148	131	170	122

10 Year		Storms pe	er Year		Average Maximum Wind Speed			
Period	40+	50+	60+	70+	40+	50+	60+	70+
1858-67	7.2	7.1	6.2	5.2	75.8	76.3	80.2	84.0
1868-77	6.9	6.8	5.4	5.0	74.8	75.3	81.9	83.6
1878-87	9.0	8.6	7.6	6.9	78.9	80.6	84.7	87.2
1888-97	7.6	7.2	5.5	4.8	76.4	78.2	86.7	90.2
1898-07	8.1	6.1	4.8	4.0	67.8	75.8	82.6	87.1
1908-17	7.1	5.7	4.9	3.8	75.1	82.9	88.2	95.7
1918-27	6.7	5.5	4.5	4.1	76.3	83.8	91.0	93.8
1928-37	10.5	8.6	6.5	5.0	74.6	81.7	91.3	100.2
1938-47	8.5	7.2	5.5	4.7	73.7	79.1	87.5	92.0
1948-57	10.9	9.6	8.2	6.6	80.7	85.8	91.8	99.2
1958-67	8.7	7.4	6.4	5.5	83.0	89.9	96.1	101.6
1968-77	9.9	8.6	7.2	4.7	72.7	77.2	81.8	92.4
1978-87	8.9	7.5	5.9	4.3	72.2	77.4	84.3	92.4
1988-97	10.1	7.8	6.4	4.9	73.4	82.4	88.9	96.9
1998-07	14.0	12.0	9.4	7.2	78.9	84.8	93.9	103.7
2008-17	10.9	8.7	5.8	5.0	71.0	78.2	90.8	95.5
Average								
1858-2017	9.1	7.8	6.3	5.1	75.3	80.6	87.6	93.5
1938-2017	10.2	8.6	6.9	5.4	75.7	81.8	89.4	96.7

The count of storms as far back as 1851 is expected to be reasonably accurate up to 70+ knots. There is a general increase in storm frequencies for each higher level of storm severity. The greatest increases are for the lower level maximum winds. Page 4 displays a straight line fit to storm frequency (Y=A+BX).

Atlantic Storms Frequency





10-Year	Change i	in	Number	of Storms	per	Year
	on the set			01 0001110	- P	

Group	Α	В	R	Change	% Ch.
40+	6.4800	0.3082	0.735	0.308	2.78%
50+	6.2075	0.1881	0.544	0.188	2.08%
60+	5.3500	0.1103	0.394	0.110	1.57%
70+	4.8525	0.0306	0.145	0.031	0.58%

Average Maximum Wind Velocity

Group	Α	В	R	Change	% Ch.
40+	75.7818	-0.0587	0.074	-0.059	-0.08%
50+	78.0371	0.2924	0.339	0.292	0.35%
60+	83.0113	0.5303	0.536	0.530	0.58%
70+	85.5732	0.9263	0.729	0.926	0.93%

The lack of sophisticated equipment to measure extreme maximum wind speeds raises questions as to the accuracy of the maximum winds for the early years. The advent of modern aircraft and the high level of military interest in the measuring and tracking North Atlantic weather for ships and planes during WWII is taken as assurance that data since 1938 is reasonable accurate. Therefore, statistics on average speed, its standard deviation and skewness is calculated from 1938 through 2017. Because of the volatility of the experience, rolling 20-year periods are analyzed.

For each of the four selected minimum wind speeds (40+, 50+, 60+ & 70+), four measures are calculated: a) average storms per year, b) average maximum wind speed, c) standard deviation of the average maximum wind speed, and d) skewness of the average maximum wind speed.

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Atlantic	W	indstorm	D	Distri	butions

20 Year		Storms 1	ber Year		Average Maximum Wind Speed			
Period	40+	50+	60+	70+	40+	50+	60+	70+
1938-57	9.7	8.4	6.9	5.7	77.65	82.95	90.04	96.24
1948-67	9.8	8.5	7.3	6.1	81.76	87.62	93.66	100.33
1958-77	9.3	8.0	6.8	5.1	77.55	83.06	88.53	97.40
1968-87	9.4	8.1	6.6	4.5	72.47	77.27	82.94	92.44
1978-97	9.5	7.7	6.2	4.6	72.82	79.93	86.71	94.84
1988-07	12.1	9.9	7.9	6.1	76.58	83.81	91.90	100.95
1998-17	14.1	11.5	8.5	7.0	75.66	83.03	93.76	100.90

20 Year	5	Standard	Deviation	1		Skev	vness	
Period	40+	50+	60+	70+	40+	50+	60+	70+
1938-57	26.8	24.9	22.1	19.3	0.433	0.383	0.335	0.364
1948-67	29.4	27.1	24.5	21.5	0.362	0.307	0.294	0.272
1958-77	28.2	26.5	25.0	22.8	0.754	0.760	0.736	0.620
1968-87	24.8	23.6	22.6	21.2	1.130	1.215	1.244	1.194
1978-97	27.1	25.5	23.9	22.4	0.994	1.001	1.022	0.966
1988-07	30.7	29.2	27.2	24.8	0.847	0.768	0.681	0.561
1998-17	30.8	29.5	26.9	24.6	0.853	0.737	0.587	0.545

On page 6, the graph of average storm counts per year displays the fluctuations from period to period. Immediately following it, is a graph displaying the trend line fitted to each storm group. The storm frequency increases for each group with the largest increase at the lower velocities and increasing at lesser amounts for each higher velocity group. This phenomenon is the result of there being larger numbers of storms with lesser velocities that trend up just enough to pierce the next higher velocity level. The last 80 years helps to show the leverage (ratio of number of storms in lower level to next higher level) that a skewed distribution has. If a uniform increase is assumed, a greater relative increase in storm counts is experienced at the lower velocity layers than it is at the higher velocity layers.

Wind Vel	o <u>city</u>	Number	
<u>From</u>	<u>to</u>	of storms	<u>Leverage</u>
30	39	175	1.33
40	49	132	0.94
50	59	141	1.16
59	69	122	1.13
70	79	108	

Within each period, the average wind velocity is calculated. Page 6 displays the average maximum wind speed over the 7 periods. Trend lines in the average maximum wind velocity are displayed in the graph immediately following. There is an increasing trend in the average maximum wind speed with the greatest increases for the higher wind velocity categories. The exception to this is for the layer of storms with the lowest wind speeds of 40 or more. It is the increase of minimal wind speeds that brings down the average slightly.

Is the distribution of storms simply a transfer of the shape of the distribution to an increased mean for more recent years? The standard deviation is calculated within each period. Page 8 displays the graph of the standard deviations. As the range of velocities is tightened, the standard deviation is reduced, but they generally move in the same upward direction. The graph of the trend lines fit to the standard deviations shows increasing trends that are almost parallel lines, with the higher wind groups at lower standard deviations as the wind speed increases.

Atlantic Storms Frequency



Years



Average Maximum Wind Velocity of Atlantic Storms

















A Wikipedia article,³ "Air Pressure and Wind," says the Net Force of a wind is determined by five factors: Net Force = PGF +G +Co +F +Ce.

- 1. PGF = Pressure Gradient Force, which is based on the distance and the difference between pressure levels.
- 2. G = Gravity.
- 3. Co = Coriolis Force, which causes greater turning forces as the distance from the equator increases.
- 4. F = Friction, which is caused by the winds interaction with the earth's surface.
- 5. Ce = Centrifugal Force.

Barometric	Number	Maximum	Barometric	Number	Maximum	Barometric	Number	Maximum
Pressure	Storms	<u>Velocity</u>	Pressure	Storms	Velocity	Pressure	Storms	<u>Velocity</u>
882	1	160.0	944	4	106.3	979	14	83.2
888	1	160.0	945	5	111.0	980	18	74.7
897	1	150.0	946	6	111.7	981	10	81.5
899	1	165.0	947	1	115.0	982	10	73.5
900	1	150.0	948	4	123.8	983	6	70.8
905	3	150.0	949	3	110.0	984	9	75.6
908	1	150.0	950	10	111.5	985	19	70.5
910	1	145.0	951	1	105.0	986	16	65.9
914	2	155.0	952	3	110.0	987	13	67.3
915	1	145.0	953	3	115.0	988	6	67.5
916	1	130.0	954	4	106.3	989	12	63.8
918	1	140.0	955	6	107.5	990	22	62.3
920	2	132.5	956	4	108.8	991	5	63.0
921	1	135.0	957	3	101.7	992	14	55.4
922	1	150.0	958	8	102.5	993	15	63.3
923	1	140.0	959	3	103.3	994	24	55.0
924	2	142.5	960	10	98.5	995	13	63.8
926	1	150.0	961	2	100.0	996	15	54.7
927	2	125.0	962	8	99.4	997	22	53.4
928	1	130.0	963	6	93.3	998	18	60.0
929	5	134.0	964	8	100.0	999	19	50.8
930	2	130.0	965	13	85.0	1000	31	51.1
931	2	142.5	966	5	84.0	1001	21	50.0
932	2	132.5	967	4	95.0	1002	21	46.7
933	2	130.0	968	9	88.3	1003	11	46.8
934	5	124.0	969	7	91.4	1004	18	46.4
935	6	120.8	970	11	91.4	1005	21	46.4
936	1	125.0	971	4	90.0	1006	12	45.8
937	2	130.0	972	5	84.0	1007	12	47.1
938	7	122.1	973	10	83.5	1008	1	45.0
939	3	121.7	974	9	84.4	1009	3	65.0
940	7	120.0	975	10	86.0	1011	1	40.0
941	4	132.5	976	8	79.4			
942	5	115.0	977	12	78.8	# Storms	750	
943	4	122.5	978	6	77.5			

Comparison of Barometric Pressure with Wind Velocity

The record barometric pressures⁴ are: 870 mb from a Pacific typhoon TIP on 10/17/1979, and 1084 mb in Siberia on 12/31/1968. A straight line fit between the measured barometric pressure and the average maximum wind velocity produced the following coefficients and correlation:

Coefficients	Pressure	Pressure	Projected Wind
а	1177.880	1000	53
b	-1.125	950	109
Correlation	-0.924	900	165

The change in barometric pressure causes a trivial change in sea level³; a decrease of 1 hPa results in a sea level rise of 1 cm. The shift between pressure levels is the most significant contributor to severe wind velocities. The longer the wind continues, the greater the impact on creation of waves.

Much of the damage resulting from windstorms are causes by the water surge. A Wikipedia article⁶, "Wind Wave," identifies 5 elements that influence the height of waves.

- 1. Wind speed relative to wave speed.
- 2. Uninterrupted distance of open water.
- 3. Width of area affected by fetch.
- 4. Wind duration, and
- 5. Water depth.

Given open ocean water conditions, the article indicates the height of the waves (trough to crest) that can develop based on continuous wind velocity are:

Wind Velocity in Knots	Hours Duration	Wave Height in Feet
30	23	13
40	42	28
50	69	49

The table on page 12 shows the number of storms and their maximum velocity by the duration in days that the storm was tracked. The maximum winds were not maintained for this whole time, but high wind velocities are correlated with the duration of the storm. Generally, the longer the storm endures, the greater the waves size.

Line Fit to Wind Velocity from Days of Duration

Coefficients	<u># Days</u>	_	<u>Days</u>	Projected Wind
а	43.412		5	63
b	3.906		10	82
			15	102
Correlation	0.582		20	122

The impact on shore, however, is a combination of the initial wave size and the structure of the shoreline. Waves break when the wave height exceeds 80% of the water depth.

A missing hazard from this analysis is the amount of rain produced by windstorms. It seems a priori that longer duration storms will produce greater levels of rain, and subsequent resulting damages.

	Maximum Wind Velocity											
Duration							100-	110-	120-		Number	Average
in Days	40-49	50-59	60-69	70-79	80-89	90-99	109	119	129	130+	Storms	Velocity
2	6	6	2								14	50.0
3	20	15	16	4	2						57	54.0
4	39	35	16	11		1				1	103	53.3
5	16	29	9	17	6	4	2	1	1		85	62.2
6	17	23	15	20	10	4	4	5	1	2	101	68.0
7	12	14	10	9	9	7	5	4	1	2	73	72.0
8	12	7	16	16	9	5	5	7	4	1	82	75.2
9	8	4	10	5	7	6	9	5	3	4	61	84.0
10	5	7	7	10	6	7	10	3	2	2	59	81.1
11	2	1	8	4	2	6	7	5	1	4	40	89.6
12	1	1	5	3	3	4	3	4	1	6	31	97.1
13	2	1	3	2	8	1	6	1	9	6	39	99.4
14	1	1	2	4	1	1	4	6	5	3	28	101.6
15	1	1	2		3	5	4		4	5	25	101.8
16				4	1	4	5	2	1	2	19	98.4
17					2	2	2	1	3	3	13	111.5
18				2			1	2	4	3	12	115.0
19								1	1	1	3	130.0
20			1				1	1	1		4	100.0
21			1							1	2	97.5
23				1				1	_	2	4	116.3
25					1						1	80.0
30						1					1	95.0
Total	142	145	123	112	70	58	68	49	42	48	857	75.8
Ave #												
Days	5.6	5.6	7.3	7.9	9.3	10.4	11.0	11.0	13.0	13.1		

Knowing that the basic distribution is skewed to the right, how has its skewness changed? The skewness (3rd moment divided by the cubed standard deviation) is calculated for each storm group within each period. The graph on page 9 shows a degree of volatility by group, but with a definite increasing trend. There is generally movement in the same direction and magnitude for each wind group for each period. The higher wind speed groups have lower skewness. The trend lines on page 9 show increases with larger increases for lower wind speed groups. This result is due to the disproportional increase in the smaller storms.

The distribution of storm severity closely fits a Weibull distribution. The Weibull model has desirable characteristics for modeling windstorms:

- 1. It is a continuous function.
- 2. It is a right tailed distribution.
- 3. Has a zero value for $x \le 0$.
- 4. As x increases, the probability density function is asymptotic to the x axis.

Range 0<= X <=+ infinity

Scale parameter b>0, Shape parameter c>0 a smaller c, extends the density distribution

Mean: $b*\hat{\Gamma}[(c+1)/c]$

Distribution function: $1 - \exp[-(x/b)^c]$

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Probability density function: $(cx^{c-1}/b^c) \exp[-(x/b)^c]$

Hazard function: cx^{c-1}/b. The Appendix gives an expanded description⁷ of the Weibull Distribution.

To verify the shift in the distribution, Weibull curves are fit to the storm severity distributions for four periods: 1938-1967, 1968-1997, 1998-2017 and the total 1858-2017 (see pages 14 thru 17). The Weibull distribution is a continuous distribution, but the windstorm data is discrete, so pages 14 thru 17 include a range +or -.5 around each x value to recognize this difference. The coefficient of correlation R is at least 0.997 for the Distribution Functions for each period. Since the Weibull distribution extends to infinity, there is always the possibility for larger values, thus a correlation of R=1.0 for sample data to a Weibull is not possible except due to rounding.

Weilbell	Para	meters				Correlation	Coefficient R
Years	С	В	Mean	Std. Dev.	Variance	Density	Distribution
1858-2017	1.407	5.117	0.925	1.212	1.470	0.986	0.998
1938-1967	1.338	6.172	1.075	1.410	1.987	0.954	0.999
1968-1997	1.480	4.100	0.767	1.006	1.012	0.951	0.998
1998-2017	1.186	4.181	0.662	0.868	0.753	0.948	0.997

It is interesting to note the number of storms with 130 of higher winds; 1938-1967 had 14 over 30 years; 1968-1997 had 10 over its 30 years, and 1998-2017 had 18 in only 20 years. The Weibull distribution for 1968-1997 indicates 1% of the storms will have winds or 145 knots or higher (11.5 table index). The Weibull distribution of 1998-2017 indicates 3.41% of the storms will have 145 knots or higher wind speeds. The Weibull distribution for 1998-2017 experience indicates 1% of the storms will reach 188 knots.

This study has considered storm frequency, maximum wind velocity, barometric pressure, and duration and wave propagation. An expanded study should include:

- 1. Width of the storm with decreasing wind velocities as distance increases from the center.
- 2. Length of the path traveled.
- 3. Location of storm, such as sea depth and water temperature, and
- 4. Precipitation.

To price windstorm/hurricane insurance, the exposure measure has to be based on current information. The number of structures continues to increase the exposure. The values of properties and construction costs have changed. Local regulations may have required improved structural standards. Ideally, a pricing structure will recognize the level of damage a structure can endure before damage occurs and the likelihood of levels of those winds.

CONCLUSION

In conclusion, the experience displays long-term upward trends in the occurrence of significant windstorms in the Atlantic. There is also a shift in the shape of the distribution that warrants consideration in predicting the likelihood of the maximum windstorm within a defined probability. Although precipitation experience is not considered, wave surge will increase with increasing winds. The greatest increases in storm frequencies is for the lesser severities but increases in severe storms make the likelihood that recognition of the 100-year storm is currently being understated and the potential damages under appreciated.

Data =	<u>1858-2017</u>		Weibull
Mean =	73.372	mean =	0.925
Variance =	4264.000	variance =	1.470
Std Dev =	65.299	std dev =	1.212
# Points =	13	C (shape)= B (scale)=	1.407 5.117

		Probability Density		Weibull		
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	207	0.142	0.103	0.128	0.140
50	2	243	0.167	0.140	0.144	0.143
60	3	188	0.129	0.143	0.138	0.131
70	4	215	0.148	0.131	0.123	0.113
80	5	149	0.102	0.113	0.103	0.094
90	6	147	0.101	0.094	0.084	0.075
100	7	116	0.080	0.075	0.066	0.058
110	8	74	0.051	0.058	0.051	0.044
120	9	55	0.038	0.044	0.038	0.032
130	10	29	0.020	0.032	0.028	0.024
140	11	18	0.012	0.024	0.020	0.017
150	12	11	0.008	0.017	0.014	0.012
<u>160</u>	13	4	0.003	0.012	0.010	0.008
Total		1456				

		Cum	Cum		Weibull	
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	207	0.142	0.037	0.096	0.163
50	2	450	0.309	0.163	0.234	0.306
60	3	638	0.438	0.306	0.376	0.443
70	4	853	0.586	0.443	0.507	0.566
80	5	1002	0.688	0.566	0.620	0.669
90	6	1149	0.789	0.669	0.714	0.753
100	7	1265	0.869	0.753	0.789	0.820
110	8	1339	0.920	0.820	0.847	0.870
120	9	1394	0.957	0.870	0.891	0.908
130	10	1423	0.977	0.908	0.923	0.936
140	11	1441	0.990	0.936	0.947	0.956
150	12	1452	0.997	0.956	0.964	0.970
160	13	1456	1.000	0.970	0.976	0.980
Co	pefficient of			Density	0.986	
C	orrelation R			Distribution	0.998	

	Data = Mean = Variance = Std Dev = # Points =	<u>1938-1967</u> 77.117 2421.762 49.211 12		mean = variance = std dev = C (shape)= B (scale)=	Weibull 1.075 1.987 1.410 1.3383 6.1715	
		Probabili	ty Density		Weibull	
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	39	0.139	0.090	0.107	0.116
50	2	41	0.146	0.116	0.119	0.119
60	3	33	0.117	0.119	0.116	0.112
70	4	34	0.121	0.112	0.107	0.101
80	5	31	0.110	0.101	0.095	0.089
90	6	21	0.075	0.089	0.082	0.076
100	7	27	0.096	0.076	0.069	0.063
110	8	24	0.085	0.063	0.057	0.052
120	9	15	0.053	0.052	0.047	0.042
130	10	7	0.025	0.042	0.038	0.034
140	11	7	0.025	0.034	0.030	0.027
150	12	2	0.007	0.027	0.024	0.021
<u>160</u>				0.021	0.019	0.016
Total		281				

		Cum	Cum		Weibull	
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	39	0.139	0.034	0.084	0.140
50	2	80	0.285	0.140	0.199	0.258
60	3	113	0.402	0.258	0.317	0.374
70	4	147	0.523	0.374	0.429	0.481
80	5	178	0.633	0.481	0.530	0.576
90	6	199	0.708	0.576	0.618	0.658
100	7	226	0.804	0.658	0.694	0.727
110	8	250	0.890	0.727	0.757	0.785
120	9	265	0.943	0.785	0.809	0.832
130	10	272	0.968	0.832	0.852	0.870
140	11	279	0.993	0.870	0.886	0.900
150	12	281	1.000	0.900	0.912	0.924
160				0.924	0.933	0.942
Co	oefficient of			Density	0.962	
C	orrelation R			Distribution	0.998	

	Data = Mean = Variance = Std Dev = # Points =	<u>1968-1997</u> 70.308 2487.625 49.876 13		mean = variance = std dev = C (shape)= B (scale)=	Weibull 0.767 1.012 1.006 1.48 4.1	
		Probabili	ty Density		Weibull	
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	50	0.171	0.126	0.162	0.178
50	2	44	0.151	0.178	0.181	0.176
60	3	58	0.199	0.176	0.166	0.152
70	4	46	0.158	0.152	0.136	0.120
80	5	17	0.058	0.120	0.104	0.089
90	6	22	0.075	0.089	0.075	0.062
100	7	22	0.075	0.062	0.051	0.042
110	8	13	0.045	0.042	0.034	0.027
120	9	10	0.034	0.027	0.021	0.017
130	10	2	0.007	0.017	0.013	0.010
140	11	2	0.007	0.010	0.008	0.006
150	12	4	0.014	0.006	0.004	0.003
<u>160</u>	13	2	0.007	0.003	0.003	0.002
Total		292				

		Cum	Cum	Weibull		
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	50	0.171	0.043	0.117	0.202
50	2	94	0.322	0.202	0.292	0.382
60	3	152	0.521	0.382	0.467	0.547
70	4	198	0.678	0.547	0.619	0.683
80	5	215	0.736	0.683	0.739	0.787
90	6	237	0.812	0.787	0.827	0.862
100	7	259	0.887	0.862	0.890	0.913
110	8	272	0.932	0.913	0.932	0.947
120	9	282	0.966	0.947	0.959	0.969
130	10	284	0.973	0.969	0.976	0.982
140	11	286	0.979	0.982	0.987	0.990
150	12	290	0.993	0.990	0.993	0.995
160	13	292	1.000	0.995	0.996	0.997
Coefficient of				Density	0.951	
Correlation R				Distribution	0.998	

	Data = Mean = Variance = Std Dev = # Points =	<u>1998-2017</u> 72.659 2544.542 50.443 13		mean = variance = std dev = C (shape)= B (scale)=	Weibull 0.662 0.753 0.868 1.186 4.181	
		Probabili	ty Density		Weibull	
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	43	0.171	0.176	0.181	0.174
50	2	56	0.222	0.174	0.163	0.150
60	3	31	0.123	0.150	0.136	0.122
70	4	28	0.111	0.122	0.109	0.097
80	5	21	0.083	0.097	0.085	0.075
90	6	13	0.052	0.075	0.065	0.057
100	7	15	0.060	0.057	0.049	0.043
110	8	11	0.044	0.043	0.037	0.032
120	9	16	0.063	0.032	0.027	0.023
130	10	9	0.036	0.023	0.020	0.017
140	11	4	0.016	0.017	0.015	0.012
150	12	4	0.016	0.012	0.011	0.009
<u>160</u>	13	<u>1</u>	0.004	0.009	0.008	0.006
Total		252				

		Cum	Cum	Weibull		
Speed	X=	Counts	Percentage	X5	Curve	X+.5
40	1	43	0.171	0.077	0.167	0.257
50	2	99	0.393	0.257	0.341	0.419
60	3	130	0.516	0.419	0.491	0.555
70	4	158	0.627	0.555	0.613	0.664
80	5	179	0.710	0.664	0.710	0.750
90	6	192	0.762	0.750	0.784	0.815
100	7	207	0.821	0.815	0.842	0.865
110	8	218	0.865	0.865	0.885	0.902
120	9	234	0.929	0.902	0.916	0.929
130	10	243	0.964	0.929	0.940	0.949
140	11	247	0.980	0.949	0.957	0.964
150	12	251	0.996	0.964	0.970	0.974
160	13	252	1.000	0.974	0.979	0.982
Coefficient of				Density	0.948	
Correlation R				Distribution	0.997	





APPENDIX

The Weibull can be defined with 1, 2 or 3 parameters. The 3-parameters probability density function is:

 $PDF(x) = [C(x-A)^{C-1}/B^{C}] \exp\{-[(x-A)/B]^{C}\}$ where the A parameter adjusts the location (± infinity).

Setting A=0 produces the 2-parameter probability density function (PDF) used in this paper and described⁷ below:

 $PDF(x) = (Cx^{C-1}/B^{C}) \exp[-(x/B)^{C}]$

The distribution function (DF) is: $DF(x) = 1 - exp[-(x/B)^{C}]$

Variance = $B^2 (\Gamma[(C+2)/C] - {\Gamma[(C+1)/C]}^2)$

Mode = B $(1-1/C)^{1/C}$ for C ≥ 1 and = 0 for C ≤ 1

C is the shape parameter (C>0). A Weibull distribution with C=1 produces the exponential distribution with mean B. A C value slightly below 3.0 looks somewhat like a normal distribution with a slight tail. The scale parameter(B), adjusts the length of the distribution. The following graphs uses Excel's function, WEIBULL(x,C,B=1, 0 & 1), to illustrate the shift in the distributions by adjusting its shape parameter (C):





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