
A NEW PROBABILITY DISTRIBUTION AND ITS APPLICATION IN MODERN PHYSICS

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ABSTRACT

In this paper we present a new symmetric probability distribution with its properties and we show that it is not a uniform distribution using some standard proofs test like Kolmogorov-Smirnov test and also we may show that it is derived from a new another special function by adjusting it using mean and deviation as two parameters , And in the second section we show that PDF present a wave function using rescaled plasma dispersion function such that we define it as a position of massive particle for such charged quantum system .

Keywords Probability distribution · Special relativity · Energy-momentum · quantum mechanics

1 Introduction

In this paper[1] We have studied a new special function behave more like error function [35] which it is defined as :

$$I(a) = \int_0^a \left(\exp(-x^2 \operatorname{erf}(x)) \right) dx$$

such that we showed a little bit its application in probability theory[13] , we are used here the same kind of that function to derive a new probability distribution which it is defined by the following formula :

$$f(h) = h^2 \exp(-h^2 \operatorname{erf}(h^2))$$

such that $\int_{-\infty}^{\infty} f(h) = 0.9895356577960071620125859226391185075044631958008$,And its Taylor series around $h = 0$ of order 12 is given by :

$$T = \frac{h^3}{3} - \frac{2h^7}{7\sqrt{\pi}} + \frac{(42 + 14\sqrt{\pi})h^{11}}{231\pi} + O(h^{13})$$

Now one can look to other analytical properties of that function regarding the new special function mentioned [1] , We are ready now to define our PDF by adjusting that function using two parameters: μ, σ

1.1 Definition of new probability distribution:

let $k = \sqrt[4]{\frac{2\pi}{1+\frac{\mu^2}{4}}}$; $h = k(z - \sigma)$; $1.01058kh^2 \exp(-h^2 \operatorname{erf}(h^2))$, The PDF can be written as :

$$F(z, \mu, \sigma) = 1.0105750026505362 \times \frac{\sqrt[4]{2\pi}}{\sqrt{1+\mu^2/4}} \frac{(z-\sigma)^2 \sqrt{2\pi}}{\sqrt{1+\mu^2/4}} \exp\left(-\frac{(z-\sigma)^2 \sqrt{2\pi}}{\sqrt{1+\mu^2/4}} \operatorname{erf}\left(\frac{(z-\sigma)^2 \sqrt{2\pi}}{\sqrt{1+\mu^2/4}}\right)\right)$$

The latter PDF is integrand to 1 for σ , and μ are arbitrary real numbers and the constant of Normalisation is 1.0105750026505362, We could then determine the central moments for a specific value of [27] (say $\mu = 1/20$) such that the odd moments are 0 implies the skewness is also 0 look to Figure 1

$$\left(\int_{-\infty}^{\infty} (z - 1)^2 F(z, \mu, \sigma) dz \right) = 0.565411$$

,with $(z, 1, \frac{1}{20})$ and

$$\left(\int_{-\infty}^{\infty} (z - 1)^4 F(z, \mu, \sigma) dz \right) = 0.545414$$

,with $(z, 1, \frac{1}{20})$ and

$$\left(\int_{-\infty}^{\infty} (z - 1)^4 F(z, \mu, \sigma) dz \right) = 0.751608$$

,with $(z, 1, \frac{1}{20})$, But the i -th central moment for any particular value of μ using Mathematica code will be and given by this mathematica Code :

```
centralMoment[i_, μ_] :=
  If[OddQ[i], 0,
    1.0105750026505362 (2 π / (1 + μ^2 / 4)) ^ (-i / 4) ×
    NIntegrate[h ^ (2 + i) Exp[-h^2 Erf[h^2]], {h, -∞, ∞}]];
  Evaluate[centralMoment[i_, μ_]]
```

Figure 1: Evaluation of i -th central moment for any particular value of

And Here is the variance (See code in figure 2) since it depend only to μ :

```
FullSimplify[centralMoment[2, μ], μ > 0]
0.282617 Sqrt[4 + μ ^ 2]
```

Figure 2: mathematica code for evaluation of variance

As we said above that PDF is symmetric about σ which mean that the Mean Value is zero , here is the plot of that PDF for $\sigma = 0, \mu = 1/20$ see figure 3 :

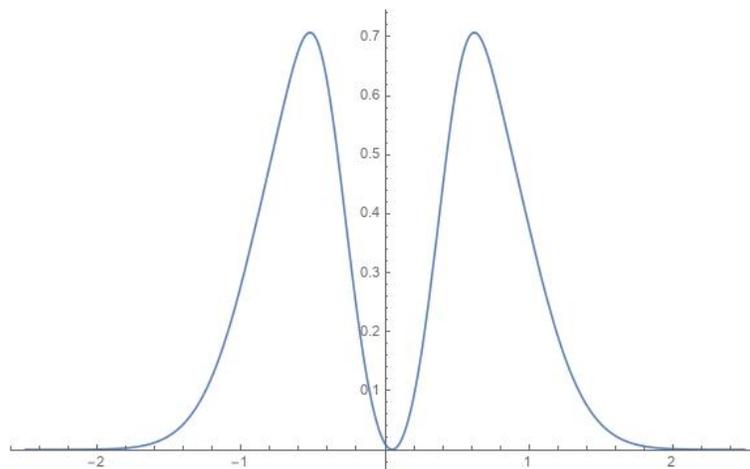


Figure 3: Plot of the new Probability distribution for $\sigma = 0, \mu = 1/20$

We may look to the plot of the new probability distribution for fixed μ and some values of σ , We take as a good example $\mu = 10$, and $\sigma = 0.5, 0.75, 1.5, 2$ as shown below in Figure 4 and we noted that the probability values decay from 0.3 to about 0.22 when we increased the value of μ (Say $\mu = 20$), The numerical evidence show us [32] that Probability distribution depend more to the values of μ than σ . See Figures (4, 5)

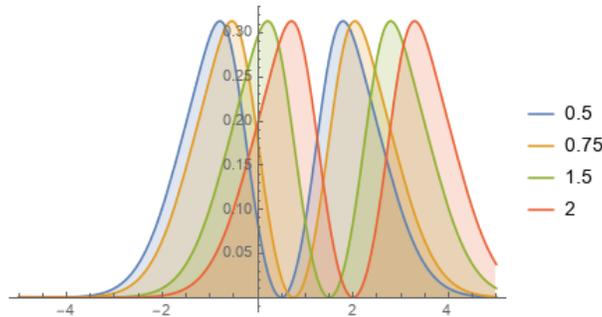


Figure 4: Plot of the new Probability distribution for $\mu = 10$, ($\sigma = 0.5, 0.75, 1.5, 2$)

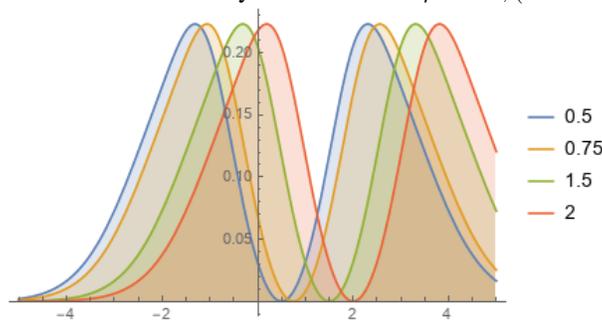


Figure 5: Plot of the new Probability distribution for $\mu = 20$, ($\sigma = 0.5, 0.75, 1.5, 2$)

We may conclude our first section by Histogram plot and test table for normality of that new probability distribution[18] for ($\sigma = 1, 2$), $\mu = 10$ and its Histogram for some given data for that Probability distribution, then here is the CDF (Cumulative distribution function) see Figure 6 for ($\sigma = 1, 2$), $\mu = 10$).

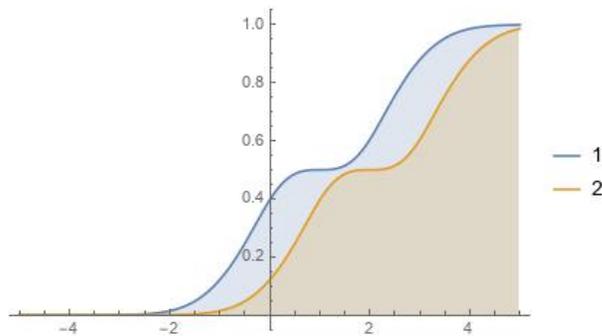


Figure 6: Plot of Cumulative distribution function for a new probability distribution, ($\sigma = 1, 2$), $\mu = 10$

Here is the mathematica code[33] with plot for Histogram corresponding to the new probability distribution for some given data with ($\sigma = 1, \mu = 10$)

Now for normality test [16],[36], We may remember together the following definition ¹ In statistics, normality tests[34] are used to determine if a data set is well-modeled by a normal distribution and to compute how likely it is for a

¹The Faddeeva function or Kramp function is a scaled complex complementary error function[?], It is related to the Fresnel integral, to Dawson's integral, and to the Voigt function. The function arises in various physical problems in describing electromagnetic response in complicated media

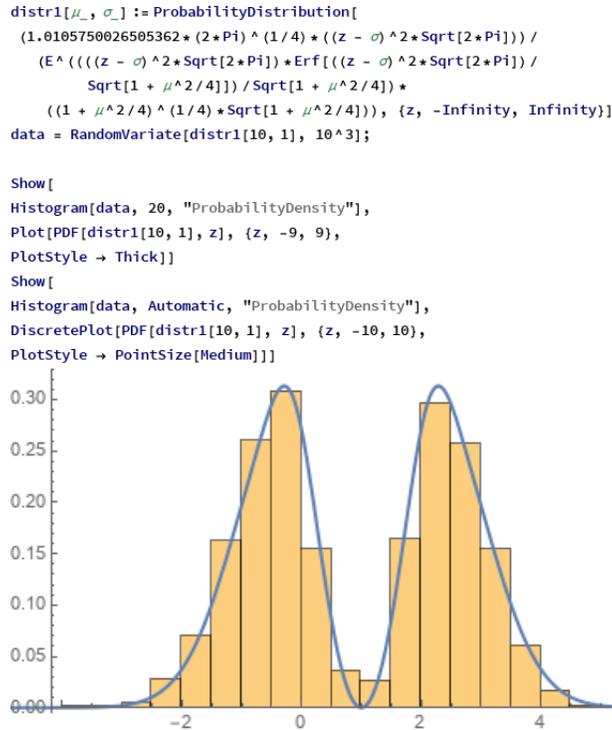


Figure 7: Histogram plot corresponding to the new probability distribution for some given data with $(\sigma = 1, \mu = 10)$

```
data = RandomVariate[distr[100, 20], 10000];
DistributionFitTest[data]
 $\mathcal{H}$  =
DistributionFitTest[data, Automatic, "HypothesisTestData"];
 $\mathcal{H}$ ["TestDataTable", All]
```

	Statistic	P-Value
Anderson-Darling	289.862	0.
Baringhaus-Henze	594.105	0.
Cramér-von Mises	56.8307	0.
Jarque-Bera ALM	713.686	0.
Kolmogorov-Smirnov	0.12989	0.
Kuiper	0.254768	0.
Mardia Combined	713.686	0.
Mardia Kurtosis	-26.6754	9.07601×10^{-157}
Mardia Skewness	1.69223	0.193307
Pearson χ^2	4490.37	0.
Watson U^2	56.8139	0.

Figure 8: Table for Normality test using many proofs with $(\sigma = 100, \mu = 20)$

random variable underlying the data set to be normally distributed. We say that the random variable is normally[37] is distributed iff The P-value is greater than 0.05

The Importance of Testing for Normality: Many statistical procedures[34] such as estimation and hypothesis testing have the underlying assumption that the sampled data come from a normal distribution. This requires either an effective test of whether the assumption of normality holds or a valid argument showing that non-normality does not invalidate the procedure. Tests of normality are used to formally assess the assumption of the underlying distribution.

Much statistical research has been concerned with evaluating the magnitude of the effect of violations of the normality assumption on the true significance level of a test or the efficiency of a parameter estimate. Geary (1947) showed that for comparing two variances, having a symmetric non-normal underlying distribution can seriously affect the true significance level of the test. For a value of 1.5 for the kurtosis of the alternative distribution, the actual significance level of the test is 0.000089, as compared to the nominal level

and we may show its application to test normality [22] of our probability distribution [16] such that we may show that the random variable of that new probability distribution is not normally distributed using many proofs of test ([3],[17]) as shown in Figure 8 table ([16],[15]). For the entropy of that new probability distribution one can compute it using the following simple mathematica Code :

```
[ (* Sample size *)
```

```
n = 97
```

```
(* Take random sample *)
```

```
x = RandomVariate[distr[10,2],n]
```

```
(* Calculate entropy *)
```

```
Entropy[x] = Log 97 ]
```

Note: for many example we have tried by mathematica always the Entropy[38] of that probability distribution is Log_2 .

2 Application to Modern physics

2.1 A new probability distribution by means of Plasma dispersion relation

In the first we may try to write our PDF in terms of plasma dispersion function or Faddeeva function [2] using the fact that : $k = \sqrt{\frac{2\pi}{1+\frac{\mu^2}{4}}}$; $x = k(z - \sigma)$; $f(z, \mu, \sigma) = 1.0105750026505kx^2 \exp(-x^2 \text{erf}(x^2))$, We try to write $\text{erf}(x^2)$ in terms of Faddeeva function using the following steps : let $w(-ix)$ be the Faddeeva function defined as :

$$w(-ix) = \exp(x^2)(1 + \text{erf}(x)) \quad (1)$$

From (1) we can get :

$$\text{erf}(x^2) = \exp(-x^4)w(-ix^2) - 1 \quad (2)$$

We use the definition of Fried and Conte for the rescaled function[35] $Z(x) = i\sqrt{\pi}w(x)$ implies that $Z(-ix^2) = i\sqrt{\pi}w(-ix^2)$, Now Multiplying the sides of equation (2) by the factor $-x^2$ using the fact that $Z(-ix^2) = i\sqrt{\pi}w(-ix^2)$, we can get the following equation

$$-x^2 \text{erf}(x^2) = i \frac{x^2}{\sqrt{\pi}} (\exp(-x^4)Z(-ix^2) - i\sqrt{\pi}) \quad (3)$$

²The wave function is the most fundamental concept of quantum mechanics. It was first introduced into the theory by analogy (Schrödinger 1926); the behavior of microscopic particles likes wave, and thus a wave function is used to describe them. Schrödinger originally regarded the wave function as a description of real physical wave. But this view met serious objections and was soon replaced by Born's probability interpretation (Born 1926), which becomes the standard interpretation of the wave function today. According to this interpretation, the wave function is a probability amplitude, and the square of its absolute value represents the probability density for a particle to be measured in certain locations

Now one can raise the Exp power of the two both sides of (3) and multiplying the obtained equation by the factor kx^2 one can get the final formula which it is the definition of our new PDF in terms of rescaled Faddeeva function

$$kx^2 \exp(-x^2 \operatorname{erf}(x^2)) = kx^2 \exp\left(i \frac{x^2}{\sqrt{\pi}} (\exp(-x^4) Z(-ix^2) - i\sqrt{\pi})\right) \quad (4)$$

with $k = \sqrt{\frac{2\pi}{1+\frac{\mu^2}{4}}}$; $x = k(z - \sigma)$ hence we are defined the PDF by means of plasma dispersion relation exactly we have got a new wave function [4] in 3D defined in terms of Energy relativistic and position X of a Massive particle at time t and C for a charged quantum system [25], [5] by:

$$\psi(X, t) = kE_0 \exp(i(CX - i\sqrt{\pi}E_0)) \quad (5)$$

such that : $C = \frac{\sqrt{2}E_0}{\exp(E_0^2)}$, $X(x) = Z(-ix^2)$, Here the position X takes values of rescaled plasma dispersion function ,

Now we may show how we have got the function defined in (5), Recall that $x = k(z - \sigma)$, let $z = \sqrt{E} \geq 0$ be the energy of massive particle at time t and let σ be the time t and $\mu = \frac{2p}{m_0c}$ such that c is the speed of light in a vacuum and m_0 is the rest massive and p is the momentum , here K is represented by means of Lorentz factor , using the formula of μ we would get $k = \frac{2\pi^{\frac{1}{4}}}{\sqrt{\gamma}}$, γ is the Lorentz factor [6] which is rarely used although it does appear in

Maxwell-Jüttner distribution [7],[30]) and it is defined by this formula : $\gamma = \sqrt{1 + \left(\frac{p}{m_0c}\right)^2}$, For instance take $t = 0$,

we have $x = kz^2 = k^2z^2$ using the assumption $z = \sqrt{E}$ and $k = \frac{2\pi^{\frac{1}{4}}}{\sqrt{\gamma}}$ gives : $x^2 = \sqrt{2\pi} \frac{E}{\gamma}$ using the fact the law of relativistic mass $mc^2 = m_0\gamma c^2$ then we can deduce directly that $x^2 = \sqrt{2\pi}E_0$ with $E_0 = m_0c^2$,The substitution of x^2 by $\sqrt{2\pi}E_0$ in (4) gives immediately the wave function [28] defined in (5).

The function defined in (5) present a probability distribution for massive particle for charged quantum system [32],[23] such that We get the probability for the observable [24],[29] to be lie in the range $[X, \delta X + X]$ at random time t with initial energy E_0 . We can interpret the histogram plot defined in figure 7 as The number of times that a particle was measured [26] to be in the range $[X, \delta X + X]$

3 Conclusion:

We have showed using our new PDF that for the massive particle with known initial energy[18] at time t_0 and momentum p such that the particule position defined by rescaled plasma function we have always a non -uniform symmetric distribution for energy -relativistic .

Data Availability:The data supported this research is The plasma dispersion function and mathematica wolfram language computation and Also Normality test .

There is no conflict of interest for Author of this research

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References

- [1] A New special function and its application in probability In *International journal of mathematics and mathematical science, Department of Mathematics ,University Batna2,Algeria November 01, 2018*
- [2] Plasma Dispersion Function In *Ramo-Wooldridge Division, Thompson Ramo Wooldridge Inc., Canoga Park, California,Space Technology Laboratories, Inc., Los Angeles, California, Science direct*
- [3] Kolmogorov-Smirnov Test In *International Encyclopedia of Statistical Science,02 December 2014*
- [4] Consistent wave equations for families of massive particles with any spin In *International Journal of Theoretical Physics ,September 1979*
- [5] Meaning of the wave function In *Unit for HPS Centre for Time, SOPHI, University of Sydney. arxiv , preprint /1001/1001.5085.pdf*
- [6] The lorentz transformation In *First Online: 13 August 2013,Undergraduate Lecture Notes in Physics, Springer 2013*

- [7] Maxwell-Jüttner distributions in relativistic molecular dynamics In *The European Physical Journal B - Condensed Matter and Complex Systems*, Springer link, published in 12 April 2006
- [8] Probability Distribution In *Encyclopedia of Health Economics*, 2014
- [9] Grinstead and Snell's Introduction to Probability In *Version dated 4 July 2006*
- [10] How to fit a response time distribution In *Published: 01 September 2000*
- [11] Response time distributions in memory scanning In *.Journal of Mathematical Psychology*,37, 526–555, *Published: 1993*
- [12] Mathematical statistics In *Mathematical statistics. San Francisco: Holden-Day, Published: 1977*
- [13] An introduction to probability theory and its applications (Vol. 1) In *Published: New York: Wiley , 1968*
- [14] The distribution of latencies constrains theories of decision time In *A test of the random-walk model using numeric comparison.Australian Journal of Psychology*,50, 149–156.,1998
- [15] The Concise Encyclopedia of Statistics. Springer, New York, NY In (2008)*Kolmogorov–Smirnov Test*
- [16] Table for estimating the goodness of fit of empirical distributions In *Ann. Math. Stat. 19*, 279–281 (6.1) (1948)
- [17] A new Weibull-X family of distributions: properties, characterizations and applications In *Journal of Statistical Distributions and Applications volume 5, Article number: 5 (2018)*
- [18] A new generalized class of distributions: properties and estimation based on type-I censored samples. *Annals of Data Science* In *JAnnals of Data Science. (2018)*,<https://doi.org/10.1007/s40745-018-0160-5>
- [19] Generalized beta-generated distributions. *Computational Statistics and Data Analysis. In MAYFEB Journal of Materials Science. 2, 5–18 (2017)*
- [20] Log-gamma-generated families of distributions. *Statistics, iFirst, In iFirst. (2012)*,<https://doi.org/10.1008/02331888.2012.748775>
- [21] The type I half-logistic family of distributions. *Journal of Statistical Computation and Simulation* In 86(4), 707–728 (2016),<https://doi.org/10.1008/02331888.2012.748775>
- [22] Beta-normal distribution and its applications. *Communications in Statistics Theory and Methods* In 31, 497–512 (2002)
- [23] The Quantization Property of Probability Distributions of the Characteristics of Dynamic Systems Observed in the Presence of Random Disturbances In *Automation and Remote Control volume 64, pages76–94(2003)*
- [24] Observation and Principles of Quantum Mechanics In *Nauk, 1991, vol. 316, no.1, pp. 57-62.*
- [25] Quantum Theory In *Physical Chemistry, 2017*
- [26] Time Evolution of Gaussian Wave Packets, In *MU-NECTEC Collaborative Research Unit on Quantum Information, Department of Physics, Faculty of Science, Mahidol University, Bangkok, Thailand, 10400., May 2016*, <https://arxiv.org/pdf/1601.03827.pdf>
- [27] Finite Trotter Approximation to the Averaged Mean Square In 36, *Article number: 1131 (2009)*
- [28] *Methods of Modern Mathematical Physics. Vol. 2: Fourier Analysis, Self-Adjointness* In *. In: Academic Press, Oct. 1975*
- [29] Localization of electromagnetic waves in random media In *. In Journal of Quantitative Spectroscopy and Radiative Transfer 79–80 (2003). Electromagnetic and Light Scattering by Non-Spherical Particles, pp. 1189–1198. issn: 0022-4073*,[http://dx.doi.org/10.1016/S0022-4073\(02\)00349-7](http://dx.doi.org/10.1016/S0022-4073(02)00349-7).
- [30] Experimental observation of the quantum Hall effect and Berry's phase in graphene In *. In: Nature 438 (2005), pp. 201–204.,doi,http://10.1038/nature04235*
- [31] Propagation of Relativistic Quantum Particles in Spacetime In *. In: Fortschritte der Physik/Progress of Physics 39.7 (1991), pp. 501–530. issn: 1521-3979.,doi,https://doi.org/10.1002/prop.2190390704*
- [32] A practical method for numerical evaluation of solutions of partial differential equations of the heat-conduction type In *. English. In: Advances in Computational Mathematics 6.1 (1996), pp. 207–226. issn: 1019-7168*
- [33] Wolfram Language System In *.Introduced in 2010 (8.0)|Updated in 2015 (10.2*
- [34] Normality Tests In *.First Online: 02 December 2014,Doi: https://doi.org/10.1007/978-3-642-04898-2_423*
- [35] Error functions In *. Retrieved October 8, 2019.*

- [36] On the Distributions of Sums of Symmetric Random Variables and Vectors, In *The Annals of Probability Vol. 14, No. 1 (Jan., 1986), pp. 247-259*
- [37] Uniform Distribution in Statistics In *First Online: 02 December 2014, DOI: https://doi.org/10.1007/978-3-642-04898-2_642*
- [38] Probability distribution and entropy as a measure of uncertainty In *Institut Supérieur des Matériaux et Mécaniques Avancés du Mans, 44 Av. Bartholdi, 72000 Le Mans, France .To appear in J. Phys. A (2008)*