

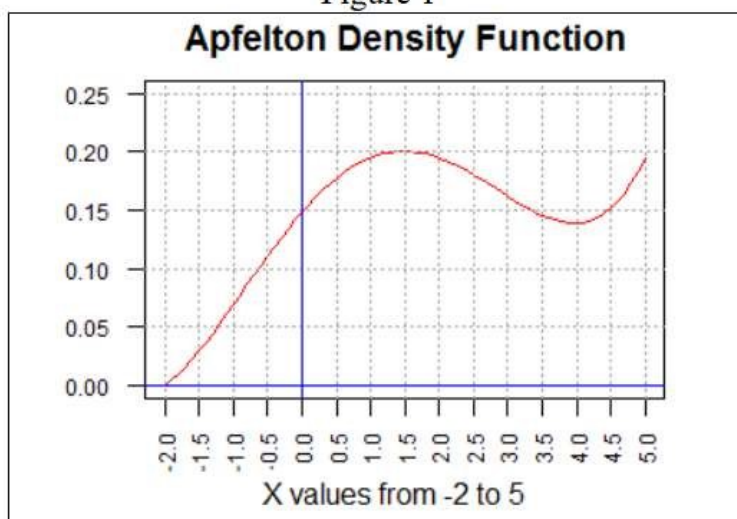
## Topic 11c: Probability: Continuous Cases

Some word of warning is due here. This section presents useless material in the sense the Apfelton and Blumenkopf distributions are a figment of my imagination. The goal here is to demonstrate probability distributions for continuous function. We can see what they are like, how to use them, how to read values from tables, and how to use R functions to get values. In future topics we will encounter real continuous probability distributions and then we can apply all that we learn here to those functions.

Point One: Continuous probability distributions are different from discrete probability distributions. In the discrete case we might ask  $P(X = 2)$  but in a continuous case the probability that  $X$  is equal to any specific value is 0. All we can do is to ask for the probability that  $X$  is less than a value, or  $X$  is greater than a value, or  $X$  is between two values, or  $X$  is less than one value or it is greater than a different value. A small consequence of this distinction is that in these situations  $P(X < 2) = P(X \leq 2)$  since the  $P(X=2)=0$ .

Here is a graph of the Apfelton probability distribution.

Figure 1



The really important points here are that the function is defined for  $x$  values in the interval from -2 to 5 and that the area under the curve, and above the  $x$ -axis, from -2 to 5 is exactly 1. Therefore, for any value, say 3, between -2 and 5, inclusive, we can take the area under the curve and to the left of the line  $y=3$  and make that area the  $P(X < 3)$ .

Look at the Apfelton distribution link on the web page.

With the probability table available we can solve all four types of problems:

$$P(X < 3) =$$

$$P(-1.45 < X < 1.63) =$$

$$P(X > 3.4) =$$

$$P(X \leq 0.54 \text{ or } X \geq 2.04) =$$

Now, instead of using the table, use the `papfelton()` function that we can load into our R environment.

```
3 # first load the papfelton() function
4 #
5 # this will actually loaded both papfelton() and
6 # qapfelton()
7 source("../apfelton.R")
```

Global Environment	
Functions	
<code>papfelton</code>	function (v, lower.tail = TRUE)
<code>qapfelton</code>	function (q, lower.tail = TRUE)

```
9 papfelton( 3 ) # solves P(X<3)
10 1 - papfelton( 3.4 ) # solves P(X>3.4)
11 # or, using the alternative approach
12 papfelton( 3.4, lower.tail=FALSE) # also solves P(X>3.4)
13 papfelton( 1.63 ) - papfelton( -1.45 ) #solves P(-1.45<X<1.63)
14 papfelton(0.54) + (1-papfelton(2.04)) #solves P(X<0.54 or X>2.04)
15 # or, using the alternative approach
16 papfelton(0.54)+papfelton(2.04,lower.tail=FALSE)

> # Now solve the problems
> papfelton( 3 ) # solves P(X<3)
[1] 0.6957328
> 1 - papfelton( 3.4 ) # solves P(X>3.4)
[1] 0.2421646
> # or, using the alternative approach
> papfelton( 3.4, lower.tail=FALSE) # also solves P(X>3.4)
[1] 0.2421646
> papfelton( 1.63 ) - papfelton( -1.45 ) #solves P(-1.45<X<1.63)
[1] 0.4340971
> papfelton(0.54) + (1-papfelton(2.04)) #solves P(X<0.54 or X>2.04)
[1] 0.7073376
> # or, using the alternative approach
> papfelton(0.54)+papfelton(2.04,lower.tail=FALSE)
[1] 0.7073376
```

So far we have seen problems where we are given a value, **x**, and then we are asked to find the probability such as getting that value or less, **P(X ≤ x)**. We can solve this using the table of probabilities, or by using `papfelton()`.

How can we solve a problem such as "Find the **x** value such that **P(X ≤ x) = 0.35**?" We can find that answer by using the table "backwards."

Find **y** such that **P(X ≤ y) = 0.35**. Answer: **y = \_\_\_\_\_** Between 1.16 and 1.17.  
Probably close to 1.166.

Find **y** such that **P(X ≥ y) = 0.17**. To do this we have to remember that if **P(X ≥ y) = 0.17** then **P(X < y)** must be **1.0 - 0.17 = 0.83**. Then we can use the table to find **y** such that **P(X < y) = 0.83**. That **y** is the answer to the original problem.

**y = \_\_\_\_\_** Between 3.90 and 3.91, almost certainly close to 3.905.

We could use `papfelton()` to find these answers by doing a successive approximation. That would be very inefficient for us, though not bad for a computer. Fortunately, we have a different function, `qapfelton()`, that will do this "backwards" problem for us. Here is how we can solve the two problems using `qapfelton()`.



```

18 # now solve the "backwards" problem if finding a
19 # value, y, such that P(X<=y)=0.35. We use
20 # qapfelton() to do this.
21 qapfelton( 0.35 )
22 # find y such that P(X>=y) = 0.17.
23 # We could do this as
24 qapfelton( 1-0.17 )
25 # or we could use the lower.tail=FALSE parameter
26 qapfelton( 0.17, lower.tail=FALSE)

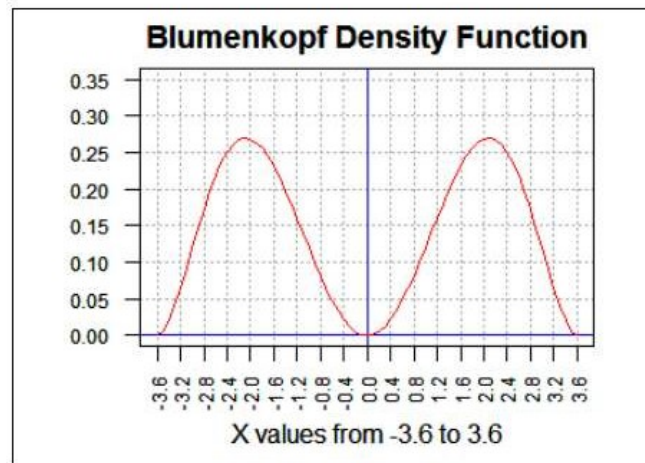
```

```

> # now solve the "backwards" problem if finding a
> # value, y, such that P(X<=y)=0.35. We use
> # qapfelton() to do this.
> qapfelton( 0.35 )
[1] 1.166635
> # find y such that P(X>=y) = 0.17.
> # We could do this as
> qapfelton( 1-0.17 )
[1] 3.904935
> # or we could use the lower.tail=FALSE parameter
> qapfelton( 0.17, lower.tail=FALSE)
[1] 3.904935

```

As part of this introduction we move onto the Blumenkopf probability distribution. It has the following graph.



Here we see that the **Blumenkopf** distribution is defined for  $x$  being between  $-3.6$  and  $3.6$ . The area under the curve and above the  $x$ -axis is **1.0**. Again, the  **$P(X < -2.4)$**  is the area under the curve and to the left of  $-2.4$ . Also, and this is an important distinction, the **Blumenkopf** distribution is **symmetric**. Thus, the  **$P(X < -2.4) = P(X > 2.4)$** . We have a table of values for the **Blumenkopf** distribution (see the web page link). And, fortunately, we can load the **pblumenkopf()** and **qblumenkopf()** functions.

$P(X < -2.4) =$

$P(X > 2.4) =$

$P(-1.335 < X < 0.827) =$

$P(X < 1.2 \text{ or } X > 2.13) =$

```

28     # start using pblumenkopf()
29 source("../blumenkopf.R")
30     # find P( X < -2.4 )
31 pblumenkopf( -2.4 )
32     # find P( X > 2.4 )
33 1 - pblumenkopf( 2.4 ) # the old, short way
34 pblumenkopf( 2.4, lower.tail=FALSE) # the way that is more clear
35     # find P(-1.335 < X < 0.827 )
36 pblumenkopf( 0.827 ) - pblumenkopf( -1.335 )
37     # find P( X <1.2 or X > 2.13 )
38 pblumenkopf( 1.2 ) + (1 - pblumenkopf(2.13) ) #old
39 pblumenkopf( 1.2 ) + pblumenkopf(2.13, lower.tail=FALSE) #new
    >     # start using pblumenkopf()
    > source("../blumenkopf.R")
    >     # find P( X < -2.4 )
    > pblumenkopf( -2.4 )
    [1] 0.143586
    >     # find P( X > 2.4 )
    > 1 - pblumenkopf( 2.4 ) # the old, short way
    [1] 0.143586
    > pblumenkopf( 2.4, lower.tail=FALSE) # the way that is more clear
    [1] 0.143586
    >     # find P(-1.335 < X < 0.827 )
    > pblumenkopf( 0.827 ) - pblumenkopf( -1.335 )
    [1] 0.1184279
    >     # find P( X <1.2 or X > 2.13 )
    > pblumenkopf( 1.2 ) + (1 - pblumenkopf(2.13) ) #old
    [1] 0.7846857
    > pblumenkopf( 1.2 ) + pblumenkopf(2.13, lower.tail=FALSE) #new
    [1] 0.7846857

```

And we can use `qblumenkopf()` to solve the "backwards" problems.

Find  $y$  such that  $P(X < y) = 0.513$ :

Find  $y$  such that  $P(X < -y \text{ or } X > y) = 0.123$ :

Find  $y$  such that  $P(X > y) = 0.823$ :

Find  $y$  such that  $P(-y < X < y) = 0.950$ :

```

40     # Find y such that P(X < y ) = 0.513
41 qblumenkopf( 0.513 )
42     # Find y such that P( X > y ) = 0.823
43 qblumenkopf( 1 - 0.823 ) # old way
44 qblumenkopf( 0.823, lower.tail=FALSE ) # better way
45     # Find y such that P( X < -y or X > y ) = 0.123
46 qblumenkopf( 0.123/2, lower.tail=FALSE)
47     # Find y such that P( -y < X < y ) = 0.950
48 qblumenkopf( (1-0.950)/2, lower.tail=FALSE)
    >     # Find y such that P(X < y ) = 0.513
    > qblumenkopf( 0.513 )
    [1] 0.6620361
    >     # Find y such that P( X > y ) = 0.823
    > qblumenkopf( 1 - 0.823 ) # old way
    [1] -2.269995
    > qblumenkopf( 0.823, lower.tail=FALSE ) # better way
    [1] -2.269995
    >     # Find y such that P( X < -y or X > y ) = 0.123
    > qblumenkopf( 0.123/2, lower.tail=FALSE)
    [1] 2.775806
    >     # Find y such that P( -y < X < y ) = 0.950
    > qblumenkopf( (1-0.950)/2, lower.tail=FALSE)
    [1] 3.025195

```