

# Understanding the MGF's Foundation:

## Why the Existence Condition is Key

A Note for Probability Students

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### The Role of the Existence Condition

We've defined the Moment Generating Function (MGF) for a random variable  $X$  as  $M_X(t) = \mathbb{E}[e^{tX}]$ . A crucial caveat accompanies this definition: it's only valid if this expectation actually yields a finite number. The standard safeguard is the *existence condition*: we require that there exists some positive number  $\delta > 0$  for which  $\mathbb{E}[e^{\delta|X|}]$  is finite.

Why is this specific condition sufficient? Let's walk through the argument to see how it guarantees that  $M_X(t)$  is well-defined (i.e., finite) for all values of  $t$  within the important interval  $(-\delta, \delta)$  centered around zero. This interval is precisely where the MGF needs to be well-behaved for us to extract moments via differentiation.

*Demonstration of Sufficiency.* Our objective is to rigorously show that if we assume  $\mathbb{E}[e^{\delta|X|}] < \infty$  for some  $\delta > 0$ , then it logically follows that  $\mathbb{E}[e^{tX}]$  must be finite for any  $t$  satisfying  $|t| < \delta$ .

Let's select an arbitrary real number  $t$  such that it falls within the specified interval, meaning  $|t| < \delta$ . Our task is to confirm that the value  $\mathbb{E}[e^{tX}]$  is finite.

The core idea is to relate the quantity  $e^{tX}$  (whose expectation we're interested in) to  $e^{\delta|X|}$  (whose expectation we know is finite). We can construct a chain of simple inequalities to bridge this gap:

1. **Relating  $tX$  to  $|tX|$ :** For any real number  $a$ , it's always true that  $a \leq |a|$ . Applying this fundamental property to the exponent  $tX$ , we have:

$$tX \leq |tX|$$

2. **Using Properties of Absolute Value:** The absolute value of a product is the product of the absolute values:  $|tX| = |t||X|$ . Substituting this gives:

$$tX \leq |t||X|$$

3. **Incorporating the Condition  $|t| < \delta$ :** Here's the crucial step where our assumption about  $t$  enters. Since  $|X|$  is non-negative, and we know  $0 \leq |t| < \delta$ , multiplying  $|X|$  by  $|t|$  results in a value less than or equal to multiplying it by  $\delta$ :

$$|t||X| \leq \delta|X|$$

(We use  $\leq$  rather than  $<$  to correctly handle the case where  $X = 0$ ).

4. **Combining the Inequalities:** Linking these steps together provides the key relationship:

$$tX \leq |t||X| \leq \delta|X|$$

Therefore, we have established that  $tX \leq \delta|X|$ .

Now, we leverage the fact that the exponential function  $f(u) = e^u$  is *monotonically increasing*. This means that if  $a \leq b$ , then  $e^a \leq e^b$ . Applying this to our inequality  $tX \leq \delta|X|$ , we obtain:

$$e^{tX} \leq e^{\delta|X|}$$

This inequality holds true for every possible outcome of the random variable  $X$ .

The final step involves the expectation operator,  $\mathbb{E}[\cdot]$ . A fundamental property of expectation is that it preserves inequalities for non-negative random variables. Since both  $e^{tX}$  and  $e^{\delta|X|}$  are always positive, we can take the expectation of both sides of the inequality:

$$\mathbb{E}[e^{tX}] \leq \mathbb{E}[e^{\delta|X|}]$$

By our initial assumption (the existence condition), we know that  $\mathbb{E}[e^{\delta|X|}]$  is a finite number. Let's denote this finite value by  $K$ . Substituting this into our inequality yields:

$$\mathbb{E}[e^{tX}] \leq K < \infty$$

Since the expectation  $\mathbb{E}[e^{tX}]$  is bounded above by the finite number  $K$ , it must necessarily be finite itself.

**Conclusion:** We have successfully demonstrated that the condition  $\mathbb{E}[e^{\delta|X|}] < \infty$  for some  $\delta > 0$  is indeed sufficient to guarantee that the MGF  $M_X(t) = \mathbb{E}[e^{tX}]$  exists as a finite value for all  $t$  in the open interval  $(-\delta, \delta)$ . This foundational result ensures the MGF is well-defined in the neighborhood of  $t = 0$ , allowing us to confidently proceed with using its derivatives to generate moments.  $\square$