# Bilder för Statistik vid 300 frågor.

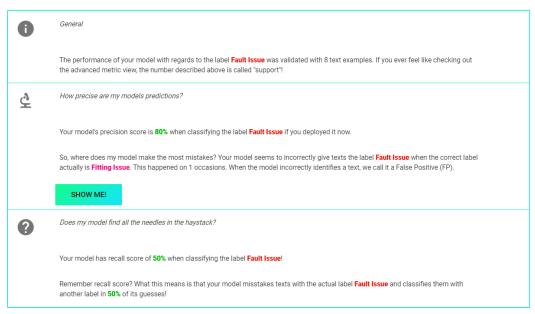
# LG3-107

# **Export overview**



| Data Export Table     |        |           |       |
|-----------------------|--------|-----------|-------|
| Labels ↑              | Manual | Predicted | Total |
| Fault Issue           | 56     | 543       | 599   |
| Fitting Issue         | 72     | 853       | 925   |
| Order info & Progress | 57     | 328       | 385   |
| Sales                 | 41     | 429       | 470   |
| Total                 | 300    | 2596      | 2896  |

# Fault Issue



## Order info & Progress



Ganaral

The performance of your model with regards to the label **Order info & Progress** was validated with 3 text examples. If you ever feel like checking out the advanced metric view, the number described above is called "support"!



How precise are my models predictions?

Your model's precision score is 100% when classifying the label Order info & Progress if you deployed it now.

So, where does my model make the most mistakes? Your model seems to incorrectly give texts the label **Order info & Progress** when the correct label actually is **Sales**. This happened on 0 occasions. When the model incorrectly identifies a text, we call it a False Positive (FP).



Does my model find all the needles in the haystack?

Your model has recall score of 67% when classifying the label Order info & Progress!

Remember recall score? What this means is that your model misstakes texts with the actual label **Order info & Progress** and classifies them with another label in **33%** of its guesses!

#### Model overview



#### How accurate is my model?

Currently the model predicts the right label 76% of the time. I have validated your results on 37 text examples.



# What's the precision of my model?

The model's precision is 80%, which means that your model is guessing the wrong label of a text 20% of the time on average (per label).

Does it sound confusing? Let us try with an actual example!

Your models precision score on the label Order info & Progress is 100%.

What this means is that when your model classifies a text as **Order info & Progress**, it is be correct in **100%** of its guesses! Your model wrongfully labels texts as **Order info & Progress** in **0%** of its predictions.



TIP! If it is important that your model is very precise, you should pay attention to this metric when checking out the rest of the labels



# Does my model find all the needles in the haystack?

Your model has a recall score of **75%**. It means that it captures **75%** of all labels on average! Let us again explain this concept with an actual example.

Your model misstakes texts with the label **Warranty & Post delivery** for another label in **0%** of its guesses. Thus, your label **Warranty & Post delivery** has a recall score of **100%**!



TIP! If it is important that your model has a good recall score, you should pay attention to this metric when checking out the rest of the labels

## **Warranty & Post delivery**



General

The performance of your model with regards to the label **Warranty & Post delivery** was validated with 7 text examples. If you ever feel like checking out the advanced metric view, the number described above is called "support"!



How precise are my models predictions?

Your model's precision score is 78% when classifying the label Warranty & Post delivery if you deployed it now.

So, where does my model make the most mistakes? Your model seems to incorrectly give texts the label **Warranty & Post delivery** when the correct label actually is **Fault Issue**. This happened on 2 occasions. When the model incorrectly identifies a text, we call it a False Positive (FP).

# SHOW ME!



Does my model find all the needles in the haystack?

Your model has recall score of 100% when classifying the label Warranty & Post delivery!

Remember recall score? What this means is that your model misstakes texts with the actual label **Warranty & Post delivery** and classifies them with another label in **0%** of its guesses!

## Fitting Issue



General

The performance of your model with regards to the label **Fitting Issue** was validated with 12 text examples. If you ever feel like checking out the advanced metric view, the number described above is called "support"!



How precise are my models predictions?

Your model's precision score is **69%** when classifying the label **Fitting Issue** if you deployed it now.

So, where does my model make the most mistakes? Your model seems to incorrectly give texts the label Fitting Issue when the correct label actually is Fault Issue. This happened on 2 occasions. When the model incorrectly identifies a text, we call it a False Positive (FP).

## SHOW ME!



Does my model find all the needles in the haystack?

Your model has recall score of 75% when classifying the label Fitting Issue!

Remember recall score? What this means is that your model misstakes texts with the actual label **Fitting Issue** and classifies them with another label in **25%** of its guesses!