# S&E Utforskande - träna modellen med totalt 400 frågor & ladda ned dataset med 400 frågor

## Testmål:

- Etikettera, dataset Bikez, totalt 400 frågor med labels enligt <a href="https://lucid.app/lucidchart/9204f0b1-66c8-476f-b066-b6eff6fc3aa2/view?page=VgO4~ivS24CQ#">https://lucid.app/lucidchart/9204f0b1-66c8-476f-b066-b6eff6fc3aa2/view?page=VgO4~ivS24CQ#</a> för grupp 3.
- Ladda ned dataset Bikez med totalt 400 frågor etiketterade

## Notes:

- Bulkarna blir nu efter 300 i större mängd, föut har de kommit 10 åt gången, nu blir de 20 åt gången. Efter några mer bulkar märker vi att vissa fortfarande är 10 åt gången, men flertalet är 20 st
- I en other bulk, är alla korrekt kategoriserade som 'other', kul att se att den har blivit så bra på att kategorisera 'other'

# Resultat:



Data Export Table			
Labels ↑	Manual	Predicted	Total
Fault Issue	106	498	604
Fitting issue	51	395	446
Order info & progress	70	431	501
Other	59	77	136
Sales	59	773	832
Total	406	2455	2861
Warranty & Post Delivery	61	281	342

#### Model overview



## How accurate is my model?

Currently the model predicts the right label 47% of the time. I have validated your results on 38 text examples.



## What's the precision of my model?

The model's precision is 37%, which means that your model is guessing the wrong label of a text 63% 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 Warranty & Post Delivery is 75%.

What this means is that when your model classifies a text as Warranty & Post Delivery, it is be correct in 75% of its guesses! Your model wrongfully labels texts as Warranty & Post Delivery in 25% 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



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

Your model has a recall score of 36%. It means that it captures 36% of all labels on average!
Let us again explain this concept with an actual example.

Your model misstakes texts with the label Sales for another label in 31% of its guesses. Thus, your label Sales has a recall score of 69%!



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