## EDAN95

# Applied Machine Learning http://cs.lth.se/edan95/

Lecture 9: Autoencoders and Generative Learning

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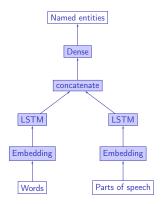
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December 3, 2018



## The Functional Model

So far, we have used the Sequential model to build networks These models correspond to pipelines with one input and one output



To build graphs, we need to use the functional model.

# Comparing the Models

### Sequential:

```
seq_model = Sequential()
seq_model.add(layers.Dense(32, activation='relu',
  input_shape=(64,)))
seq_model.add(layers.Dense(32, activation='relu'))
seq_model.add(layers.Dense(10, activation='softmax'))
Functional:
input_tensor = Input(shape=(64,))
x = layers.Dense(32, activation='relu')(input_tensor)
x = layers.Dense(32, activation='relu')(x)
output_tensor = layers.Dense(10, activation='softmax')(x)
model = Model(input_tensor, output_tensor)
```

## Building a Multi Input Model

To build a multi input, we need the functional model and at a certain point, merge the branches with layers.concatenate() function

## Code Example

We will now build a NER tagger that uses two inputs: the words and parts of speech and we will compare it to a sequential model Jupyter Notebooks: 5.2-multiinput.ipynb and 5.3-monoinput.ipynb

# Multiple Outputs

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