

Language modeling
Using transformers

MLM ← Causal LM (More on these at the very end)

BERT,
RoBERTa

predictive, comprehension

look at both
preceding & succeeding
text

Thus, uses bidirectional
full context

BART, PLBART, T5
GPT, Llama, CodeLlama

generative

predict the next token
in a sequence based
only on the tokens
that came before it

Thus, only uses uni-directional
context (left-to-right)
hence
(autoregressive)
(more on this
later)

both use the transformer
model.

- evolved from RNN-based seq. to seq. model

- key feature → ^{new} attention mechanisms

attention → the importance of each
component in a sequence
relative to other components
in that seq.

also referred to
as "self-attention" from
the perspective of a token
in a sequence.

formalized by Vaswani, it to have
3 components Q, K, V, & computed
as,

$$\text{Attention}(Q, K, V) = \text{softmax}\left(\frac{QK^T}{\sqrt{d_k}}\right)V$$

dim'd. dimensionality
of the key

In transformers.

You have multi-head attention

↓
runs several attentions
in parallel,

↓
each can focus on different
relationships in text,

Syntax, semantics

Other components of transformers,

FFNs built by FCs

Residual conns. (a type of skip conn.)

↓
any bypassing
conn.

↓

additive skip

(lets each layer make
a small edit to the
representation instead
of rewriting from scratch)

employs ~~self-supervision~~ [For later]

(also exists in RNNs)

can

IMP: transformers have either or both
of 2 architectural components
that determines its main objective,
encoders & decoders.

uses bi-directional context

encoder-only

BERT

MLM

encoder-decoder
(autoencoder)

BART

decoder-only

GPT,
Llama

Causal LMs

uses left-to-right
context

Masked lang. modeling^(MLM) & causal lang. modeling^(CLM)
are language modeling techniques which
distinguished by the pre-training
objectives they optimize, which in
turn determine how context is used
during prediction.

/See more/

Also note the denoising
process in BART.