

## A

accuracy  
    confusion matrices, 120  
    output verification, 192  
adaptive pretraining, 320  
agents, 218-223  
    agentic RAG, 256  
    ReAct in LangChain, 221-223  
    step-by-step reasoning, 219-221  
AI (artificial intelligence)  
    accelerated development of, 3  
    defined, 4  
ALBERT, 115  
align\_labels function, 350  
all-MiniLM-L6-v2 model, 309  
all-mpnet-base-v2 model, 337  
Annoy, 239  
Anthropic Claude, 29  
APIs (application programming interfaces), 30  
    Cohere, xv, 230  
    external, 134  
    generating embeddings, 123  
    OpenAI, xv, 133  
artificial intelligence (see AI)  
ArXiv, 138  
attention, 11-18  
    overview of, 11-14  
    Transformer architecture, 15-18  
        attention calculation, 91-93  
        attention layer, 79, 86, 88, 106  
        Flash Attention, 100  
        grouped-query attention, 98-100  
        local attention, 96  
        multi-query attention, 98-100

    optimizing attention, 98  
    self-attention and relevance scoring,  
        93-94  
    sparse attention, 96  
attention heads, 91-93, 107  
audience, in text-generation prompts, 178  
Augmented SBERT, 311-315  
autoregressive architecture, 12, 15, 22, 76, 97

## B

bag-of-words model, 6-7  
    embeddings, 10  
    topic modeling, 148-150, 156  
benchmarks, in generative model evaluation,  
    374  
BERT (Bidirectional Encoder Representations  
    from Transformers), 18-20  
    adoption by search engines, 225  
    BERT-like models, 115  
    comparing to other trained tokenizers, 47  
    fine-tuning pretrained BERT models,  
        325-328  
    masked language modeling, 311  
    Transformer blocks versus, 97  
BERTopic, 148-155  
    algorithmic variants, 152  
    modularity of, 151  
    representation blocks, 156-163  
        KeyBERTInspired, 158  
        maximal marginal relevance, 159  
        text generation, 160-163  
BERTScore, 374  
bias and fairness, 28

Bidirectional Encoder Representations from Transformers (see BERT)  
bitsandbytes package, 369  
BLEU, 374  
BLIP-2 (Bootstrapping Language-Image Pre-training for Unified Vision-Language Understanding and Generation 2)  
  chat-based prompting, 283-286  
  image captioning, 280  
  preprocessing text, 280  
  Q-Former, 273-277  
BM25 algorithm, 233  
BPE (byte pair encoding), 43, 55  
byte tokens, 45

## C

c-TF-IDF, 149, 158, 161  
capitalization, 56  
captioning, 280-283  
centroid-based algorithms, 143  
chains, 182-184, 202-209  
  chain-of-thought, 185-188  
  chaining single prompt, 203-205  
  sequential chaining of multiple prompts, 206-209  
character tokens, 45  
chat tokens, 54  
chat-based prompting, 283-286  
Chatbot Arena, 376  
ChatGPT, 202, 273, 383, 386  
  release of, 3  
  text classification, 132-135  
chatgpt\_generation function, 133  
chat\_history input variable, 211  
classification reports, 119-120  
classification step, embedding model, 121  
CLIP, 265-268  
  connecting text and images, 265  
  generating multimodal embeddings, 265-268  
  OpenCLIP, 268-272  
closed-source LLMs, 29  
[CLS] token, 18, 48, 60, 270, 294, 318, 351  
cluster model, 142-144  
clustering (see text clustering)  
CNNs (convolutional neural networks), 260  
Cohere, 252  
  Command R+, 30, 256  
  creating accounts, xv, 230

  generating embeddings, 123  
  query rewriting, 255  
  Rerank endpoint, 241  
completion models, 22  
compute\_metrics function, 351  
confusion matrices, 119  
CoNLL-2003 dataset, 347  
constrained sampling, 194-197  
context  
  attention and, 88  
  prompt engineering, 178  
  training datasets, 64  
context length  
  completion models, 22  
  token processing limits, 81  
context window, in completion models, 22  
contrastive learning  
  text embedding models, 291-293, 296  
  word2vec algorithm and, 64-67  
Contrastive Tension (CT), 316  
conversation buffer memory, 217  
  overview of, 210-212  
  windowed, 212-214  
conversation summary memory, 214-217  
convert\_ids\_to\_tokens function, 269  
convolutional neural networks (CNNs), 260  
cosine similarity, 125, 302-304  
CountVectorizer, 150  
cross-encoders, 244  
cross-entropy loss, 305  
CT (Contrastive Tension), 316

## D

data outliers, 143  
DataCollator class, 326, 342  
datamapplot package, 163  
DBSCAN (Density-Based Spatial Clustering), 143  
DeBERTa, 59, 115  
decoder-only models (see generative models)  
decoding strategy, 79-81, 106  
dense retrieval, 226, 228-240  
  caveats of, 234  
  example of, 230-234  
  fine-tuning embedding models for, 239  
  nearest neighbor search versus vector data-bases, 238  
  text chunking, 235-237  
density-based algorithms, 143

Density-Based Spatial Clustering (DBSCAN),  
143

dimensionality reduction model, 140-142

DistilBERT, 115, 120

domain adaptation, 320

do\_sample parameter, 34

DPO (Direct Preference Optimization),  
384-389

fine-tuning, 388

model quantization, 386

templating alignment data, 386

training configuration, 387

DSPy, 200

## E

easy negatives, 308

Elo rating system, 377

embeddings, 8-10, 37, 57-70

dense retrieval, 226, 228-240

embedding models, defined, 114, 290

multimodality, 263-272

CLIP, 265-268

OpenCLIP, 268-272

overview of, 8-10, 289-291

positional embeddings, 102-104

recommendation systems, 67-70

text classification tasks that leverage,  
120-126

supervised classification, 121-123

zero-shot classification, 123-126

text clustering pipeline, 139-146

cluster model, 142-144

dimensionality reduction model,  
140-142

embedding model, 139

inspecting clusters, 144-146

text embedding models, 61-63, 289-321

contrastive learning, 291-293

creating, 296-308

fine-tuning, 309-315

SBERT, 293-296

unsupervised learning, 316

token embeddings, 57-61

creating contextualized word embed-  
dings, 58-61

tokenizer's vocabulary and, 57

types of, 10

word embeddings, 63-67

pretrained, 63

word2vec algorithm and contrastive  
training, 64-67

encoder-decoder models, 128

encoder-only models (see representation mod-  
els)

ethics, validating output, 192

exponential backoff, 134

## F

F1 score, confusion matrices, 120

FAISS, 239

Falcon, 168

feature extraction step, embedding model, 121

feedforward layer, 86

feedforward neural networks, 15, 87, 106, 326

few-shot classification, 333-339

fine-tuning for classification, 337-339

SetFit, 333-336

few-shot prompting, 181-182, 192-194

find\_topics() function, BERTopic, 154

fine-tuning

embedding models for dense retrieval, 239

generative models, 192, 355-389

evaluating, 373-377

preference tuning, 356, 378-389

supervised fine-tuning, 356-373

training steps, 355-357

overview of, 26

representation models, 323-354

few-shot classification, 333-339

masked language modeling, 340-345

named-entity recognition, 345-353

supervised classification, 323-332

T5 model, 129

text embedding models, 309-315

Augmented SBERT, 311-315

supervised, 309-311

Flan-T5 model, 50, 130-131

Flash Attention, 100, 107

forward pass, 106

components of, 76-79

defined, 74

foundation models, 23

fp16 parameter, 299

freezing layers, 298, 328-332

frozen (nontrainable) models, 114, 121, 324

## G

Ġ symbol, 280

Galactica, 52

General Language Understanding Evaluation (GLUE) benchmark, 297, 300, 359, 374

generated\_text variable, 281, 283

generation\_output variable, 42

generative models, 20-22

- evaluating, 373-377
  - automated evaluation, 376
  - benchmarks, 374
  - human evaluation, 376
  - leaderboards, 376
  - word-level metrics, 374
- fine-tuning, 355-389
  - evaluation, 373-377
  - preference tuning, 356, 378-389
  - supervised fine-tuning, 356-373
  - training steps, 355-357
- reasoning, 184-191
  - self-consistency, 188
  - tree-of-thought, 189-191
- representation models versus, 20
- text classification, 127-135
  - ChatGPT, 132-135
  - T5, 128-131

generative pre-trained transformers (see GPTs)

Generative Pseudo-Labeling (GPL), 316

Gensim library, 63

get\_topic function, 153

get\_topic\_info() method, 152

GGUF model, 195, 200

GitHub, xvi

GloVe, 294

GLUE (General Language Understanding Evaluation) benchmark, 297, 300, 359, 374

gold datasets, 312

Goodhart's Law, 377

Google Colab, xii, xiv, 29, 33, 85, 372

Google Gemini, 250

Google Search, 225

GPL (Generative Pseudo-Labeling), 316

GPT2Tokenizer, 280

GPTs (generative pre-trained transformers), 20, 167

- (see also text generation)
- GPT-1, 20
- GPT-2, 3, 20, 46, 49
- GPT-3, 20, 87, 96, 364
- GPT-3.5, 23, 132, 205, 221
- GPT-4, 23, 29, 38, 50, 87

GPUs

- Flash Attention, 100
- requirements, xiv, 28
- SRAM and HBM, 100

grammar, 192, 194-197

greedy decoding, 80

grounded generation, 250-252

grouped-query attention, 98-100, 107

GSM8k, 374

Guardrails, 194

Guidance, 194

## H

hallucination

- avoiding in in instruction-based prompting, 177
- text generation models, 225

hard negatives, 308

harmful content, generating, 28

Haystack, 200

HBM (high bandwidth memory), 100

HDBSCAN (Hierarchical Density-Based Spatial Clustering of Applications with Noise), 143, 153

HellaSwag, 374

high bandwidth memory (HBM), 100

Hugging Face, 32, 112

- creating accounts, xv
- evaluate package, 351
- tokenizers, 56

human evaluation, 145, 376

HumanEval, 375

hybrid search, 235, 242

## I

Idefics 2, 277

images (see multimodality)

in-context learning, 180-182

indexes, 232

InfoNCE, 304

input\_ids variable, 40

instruction-based prompting, 175-177

intellectual property, 28

intuition-first philosophy, xi

invoke function, 204

## J

Jupyter, 85

## K

k-means algorithm, 143, 151, 153  
KeyBERTInspired, 158, 160  
keyword search  
    reranking, 242-243  
    verifying semantic search with, 233  
kv (keys and values) cache, 83-85

## L

LangChain, 200  
    (see also chains)  
    loading quantized models with, 200-202  
    ReAct in, 221-223  
Language AI (Language Artificial Intelligence), 3-24  
    defining, 4  
    recent history of, 5-24  
        attention, 11-18  
        bag-of-words model, 6-7  
        embeddings, 8-10  
        generative models, 20-22  
        representation models, 18-20  
        Year of Generative AI, 23-24  
language modeling, 355  
language modeling head (LM head), 76-79  
large language models (see LLMs)  
latent Dirichlet allocation, 147  
LayerNorm, 101  
leaderboards, in generative model evaluation, 376  
learning\_rate parameter, 371  
Llama, 168  
Llama 2, xv, 29, 53, 98, 168, 273  
llama-cpp-python library, 195  
LLaVA, 277  
LLM-as-a-judge, 257  
LLMs (large language models), 3-35  
    code examples and exercises, xvi  
    embeddings, 37, 57-70  
        recommendation systems, 67-70  
        text embeddings, 61-63  
        token embeddings, 57-61  
        word embeddings, 63-67  
    fine-tuning generative models, 355-389  
        evaluation, 373-377  
        preference tuning, 356, 378-389  
        supervised fine-tuning, 356-373  
        training steps, 355-357  
    fine-tuning representation models, 323-354

    few-shot classification, 333-339  
    masked language modeling, 340-345  
    named-entity recognition, 345-353  
    supervised classification, 323-332  
generating text, 32-34  
generative models, 20-22  
hardware and software requirements, xiv, 28  
high-level view, 38  
history of Language AI, 5-24  
interfacing with, 29-32  
    closed-source models, 29  
    open models, 30-32  
intuition-first philosophy, xi  
moving definition of, 25  
multimodality, 259-286  
    embedding models, 263-272  
    text generation models, 273-286  
    Vision Transformer, 260-262  
prompt engineering, 167-198  
    chain prompting, 182-184  
    in-context learning, 180-182  
    instruction-based prompting, 175-177  
    output verification, 191-197  
    potential complexity of prompts, 177-179  
    prompt components, 173-175  
    reasoning with generative models, 184-191  
    text generation models, 167-172  
representation models, 18-20  
responsible development and usage of, 28  
retrieval-augmented generation, 227, 249-257  
    agentic RAG, 256  
    converting search system to, 250  
    evaluating results, 257  
    grounded generation, 252  
    multi-hop RAG, 256  
    multi-query RAG, 255  
    query rewriting, 255  
    query routing, 256  
    with local models, 252, 254  
semantic search, 225-249  
    dense retrieval, 226, 228-240  
    reranking, 226, 240-244  
    retrieval evaluation metrics, 244-249  
text classification, 111-135  
    with generative models, 127-135  
    movie reviews, 112, 113

- with representation models, 113-126
- text clustering, 137-146
- text embedding models, 289-321
  - contrastive learning, 291-293
  - creating, 296-308
  - fine-tuning, 309-315
  - SBERT, 293-296
  - unsupervised learning, 316
- text generation, 199-224
  - agents, 218-223
  - chains, 202-209
  - memory of conversations, 209-217
  - model I/O, 200-202
- tokens and tokenizers, 37-61
  - comparing trained tokenizers, 46-54
  - downloading and running LLMs, 39-42
  - input preparation, 38
  - text breakdown, 43
  - token embeddings, 57-61
  - tokenization schemes, 44-46
  - tokenizer properties, 55-56
- topic modeling, 138, 146-163
- training paradigm of, 25-26
- Transformer architecture, 73-107
  - decoding strategy, 79-81
  - forward pass components, 76-79
  - inputs and outputs of, 74-76
  - keys and values cache, 83-85
  - parallel token processing and context size, 81-83
  - recent improvements to, 95-105
  - Transformer blocks, 85-94
  - utility of, 27
- LM head (language modeling head), 76-79
- LMQL, 194
- local attention, 96
- LoRA (low-rank adaptation), 361-364, 387
  - (see also QLoRA)
- lora\_alpha parameter, 370
- loss functions, 301-308
  - cosine similarity loss, 302-304
  - multiple negatives ranking loss, 304-308
- lr\_scheduler\_type parameter, 371

## M

- Mamba, 24
- MAP (mean average precision), 244-249
- MarginMSE loss, 302
- masked language modeling (MLM), 340-345
  - mask\_token [MASK], 48
- Massive Text Embedding Benchmark (MTEB), 116, 140, 253, 300
- matplotlib library, 145
- maximal marginal relevance (MMR), 159
- max\_new\_tokens parameter, 34
- McCarthy, John, 4
- mean average precision (MAP)
- memory of conversations, 209-217
  - conversation buffer, 210-212
  - conversation summary, 214-217
  - windowed conversation buffer, 212-214
- Meta Llama model, 30
- Microsoft Bing, 225, 241
- Microsoft Bing AI, 250
- Microsoft Phi model, 30
- microsoft/mpnet-base model, 297
- min\_cluster\_size parameter, 144
- min\_dist parameter, 142
- MIRACL, 243
- Mistral, 30, 168, 277
- MLM (masked language modeling), 340-345
- MMLU, 374, 376
- MMR (maximal marginal relevance), 159
- MNLI (Multi-Genre Natural Language Inference) corpus, 297, 305
- MNR (multiple negatives ranking) loss, 304-308
- model I/O, 200-202
- monoBERT, 244
- MTEB (Massive Text Embedding Benchmark), 116, 140, 253, 300
- Multi-Genre Natural Language Inference (MNLI) corpus, 297, 305
- multi-hop RAG, 256
- multi-query attention, 98-100
- multi-query RAG, 255
- multilevel perceptrons, 79
  - (see also feedforward neural networks)
- multimodality, 259-286
  - defined, 259
  - embedding models, 263-272
    - CLIP, 265-268
    - OpenCLIP, 268-272
  - text generation models, 273-286
    - BLIP-2, 273-277
    - chat-based prompting, 283-286
    - image captioning, 280-283
    - preprocessing images, 278

- preprocessing text, 279
- Vision Transformer, 260-262
- multiple negatives ranking (MNR) loss, 304-308

## N

- named-entity recognition (see NER)
- natural language inference (NLI), 296
- natural language processing (NLP), 4, 293
- nDCG (normalized discounted cumulative gain), 243, 249
- nearest neighbor search
  - pretrained word embeddings, 63
  - recommendation system embeddings, 68
  - vector databases versus, 238
- negative sampling, 65
- NER (named-entity recognition), 345-353, 360
  - fine-tuning for, 352
  - preparing data for, 347-351
- neural networks, 8
- NLI (natural language inference), 296
- NLP (natural language processing), 4, 293
- noise-contrastive estimation, 65
- nonplayable characters (NPCs), 4
- nontrainable (frozen) models, 114, 121, 324
- normalization, Transformer block, 101
- normalized discounted cumulative gain (nDCG), 243, 249
- NPCs (nonplayable characters), 4
- NTXentLoss, 304
- nucleus sampling, 171
- NumPy, 238
- num\_train\_epochs parameter, 299, 371
- NVIDIA GPUs, xiv, 33
- n\_components parameter, 142

## O

- Odds Ratio Preference Optimization (ORPO), 389
- one-shot prompting, 182
  - chain-of-thought versus, 186
  - in-context learning, 181
- Open LLM Leaderboard, 202, 376
- open-source LLMs, 30-32
- OpenAI, 132
  - (see also ChatGPT; GPTs)
  - creating accounts, xv, 133
  - generating embeddings, 123
- OpenCLIP, 268-272

- optim parameter, 372
- ORPO (Odds Ratio Preference Optimization), 389
- output verification, 191-197
  - constrained sampling, 194-197
  - providing examples, 192

## P

- pad\_token [PAD], 47
- parallel processing, 91, 106
- parallel prompts, 184
- PCA (Principal Component Analysis), 141
- PEFT (parameter-efficient fine-tuning), 359-367
  - adapters, 359-361
  - compression, 364-367
  - LoRA, 361-364
- peft library, 370
- peft\_config parameter, 371
- Perplexity, 250
- persona, in text-generation prompts, 178
- per\_device\_eval\_batch\_size argument, 299
- per\_device\_train\_batch\_size argument, 299
- Phi-3
  - comparing to other trained tokenizers, 53
  - forward pass, 79
  - loading quantized models, 202
  - prompt template, 204
  - quantization, 201
- Phi-3-mini, 32, 168
- Pinecone, 239
- positional embeddings, 102-104
- PPO (Proximal Policy Optimization), 383
- precision predictions, confusion matrices, 119
- predictions, task-specific model, 118
- preference tuning, 132, 356, 378-389
  - Direct Preference Optimization, 384-389
    - fine-tuning, 388
    - model quantization, 386
    - templating alignment data, 386
    - training configuration, 387
  - reward models, 379-383
    - inputs and outputs of, 380
    - training, 380-383
- pretraining, defined, 26
- primacy effect, 177
- Principal Component Analysis (PCA), 141
- projection matrices, 92
- prompt engineering, 127, 167-198

- chain prompting, 182-184
- in-context learning, 180-182
- instruction-based prompting, 175-177
- output verification, 191-197
  - constrained sampling, 194-197
  - providing examples, 192
- potential complexity of prompts, 177-179
- prompt components, 173-175
- reasoning with generative models, 184-191
- text generation models, 167-172
  - choosing, 167
  - controlling output, 170-172
  - loading, 168-169
- Proximal Policy Optimization (PPO), 383
- Python, learning about, xii

## Q

- Q-Former (Querying Transformer), 274-277
- Q8 model, 201
- QLoRA (quantized low-rank adaptation), 367-373
  - fine-tuning, 372
  - LoRA configuration, 370
  - merging weights, 373
  - model quantization, 369
  - templating instruction data, 367
  - training configuration, 371
- quantization, 201, 364-367
- quantization\_config parameter, 371
- quantized low-rank adaptation (see QLoRA)
- Querying Transformer (Q-Former), 274-277

## R

- r parameter, 370
- RAG (retrieval-augmented generation), 57, 63, 227, 249-257
  - agentic RAG, 256
  - basic pipeline, 249
  - converting search system to, 250
  - evaluating results, 257
  - grounded generation, 252
  - with local models, 252-254
  - multi-hop RAG, 256
  - multi-query RAG, 255
  - query rewriting, 255
  - query routing, 256
- Ragas, 257
- random\_state parameter, 142
- rate limit errors, 134

- ReAct
  - in LangChain, 221-223
  - step-by-step reasoning, 219
- reasoning
  - with generative models, 184-191
    - chain-of-thought, 185-188
    - self-consistency, 188
    - tree-of-thought, 189-191
  - step-by-step, 219, 221
- recall predictions, confusion matrices, 119
- recency effect, 177
- recommendation systems, 67-70
- recurrent neural networks (RNNs), 11
- reduce\_outliers() function, 153
- regulation, 28
- relevance scoring, 90, 92-94, 104
- repository, xiv
- representation models, 18-20
  - defined, 7
  - fine-tuning for classification, 323-354
    - few-shot classification, 333-339
    - masked language modeling, 340-345
    - named-entity recognition, 345-353
    - supervised classification, 323-332
  - generative models versus, 20
  - text classification, 113-126
    - classification tasks that leverage embeddings, 120-126
    - model selection, 115-116
    - task-specific models, 116
- representation\_model parameter, 161
- reranking, 226, 240-244
  - BERTopic, 156
  - example of, 241-243
  - function of reranking models, 244
  - sentence transformers, 243
- response validation, in chain prompting, 184
- retrieval evaluation metrics, 244-249
  - scoring multiple queries with mean average precision, 248
  - scoring single queries with average precision, 247
- retrieval-augmented generation (see RAG)
- return\_full\_text parameter, 33
- reward models, 379-383
  - inputs and outputs of, 380
  - training, 380-383
- RMSNorm, 101
- RNNs (recurrent neural networks), 11



RoBERTa, 46, 115  
RoPE (rotary positional embeddings), 102-104  
Rorschach test, 282  
Rotten Tomatoes dataset, 112, 325  
ROUGE, 374  
RWKV, 24

## S

SBERT, 293-296, 311-315  
self-attention, 15, 93-94  
self-consistency, 188  
semantic search, 225-249  
    defined, 225  
    dense retrieval, 226, 228-240  
        caveats of, 234  
        example of, 230-234  
        fine-tuning embedding models for, 239  
        nearest neighbor search versus vector  
        databases, 238  
        text chunking, 235-237  
    reranking, 226, 240-244  
        example of, 241-243  
        function of reranking models, 244  
        sentence transformers, 243  
    retrieval evaluation metrics, 244-249  
        scoring multiple queries with mean average precision, 248  
        scoring single queries with average precision, 247  
Semantic Textual Similarity Benchmark (STSB), 298  
semi-hard negatives, 308  
sentence-transformers, 62, 122, 151, 243, 272, 293-295, 309, 333  
SentencePiece, 50  
[SEP] token, 47, 60, 351  
sequence-to-sequence models, 127, 128  
SetFit, 323, 333-336  
SFT (supervised fine-tuning), 356-373  
    full fine-tuning, 357  
    parameter-efficient fine-tuning, 359-367  
        adapters, 359-361  
        compression, 364-367  
        LoRA, 361-364  
    QLoRA, 367-373  
        fine-tuning, 372  
        LoRA configuration, 370  
        merging weights, 373  
        model quantization, 369

    templating instruction data, 367  
    training configuration, 371  
shared memory (SRAM), 100  
shortlisting, 242  
silver datasets, 312  
SimCSE (Simple Contrastive Learning of Sentence Embeddings), 316  
skip-gram, 65  
softmax loss function, 301  
song recommendation systems, 67-70  
sparse attention, 96  
special tokens, 47, 55  
specificity, in instruction-based prompting, 176  
SRAM (shared memory), 100  
StableLM, 168  
StarCoder2, 51  
step-by-step reasoning, 219-221  
structured output, validating, 191  
STSB (Semantic Textual Similarity Benchmark), 298  
subword tokens, 44  
supervised classification, 121-123  
    fine-tuning representation models for, 323-332  
    freezing layers, 328-332  
    pretrained BERT models, 325-328  
supervised fine-tuning (see SFT)  
system 1 and 2 thinking processes, 185

## T

T5 (Text-to-Text Transfer Transformer), 128-131  
target\_modules parameter, 370  
task-specific models, 113  
temperature parameter, 171-172, 188  
Tesla T4, 372  
test splits, 113, 118  
text chunking, 230, 235-237  
    approaches for, 237  
    multiple vectors per document, 236  
    one vector per document, 236  
text classification, 111-135  
    with generative models, 127-135  
        ChatGPT, 132-135  
        T5, 128-131  
    movie reviews, 112-113  
    with representation models, 113-126  
        classification tasks that leverage embeddings, 120-126

- model selection, 115-116
  - task-specific models, 116-120
- text clustering, 137-146
  - CLIP embedding model and, 265
  - common pipeline for, 139-146
    - cluster model, 142-144
    - dimensionality reduction model, 140-142
    - embedding model, 139
    - inspecting clusters, 144-146
- text embedding models, 61-63, 289-321
  - contrastive learning, 291-293
  - creating, 296-308
    - evaluating, 300
    - generating contrastive examples, 296
    - loss functions, 301-308
    - training, 297-300
  - fine-tuning, 309
    - Augmented SBERT, 311-315
    - supervised, 309-311
  - SBERT, 293-296
  - unsupervised learning, 316
- text generation, 32-34, 199-224
  - agents, 218-223
    - ReAct in LangChain, 221-223
    - step-by-step reasoning, 219-221
  - chains, 202-209
    - chaining single prompt, 203-205
    - sequential chaining of multiple prompts, 206-209
  - memory of conversations, 209-217
    - conversation buffer, 210-212
    - conversation summary, 214-217
    - windowed conversation buffer, 212-214
  - model I/O, 200-202
  - multimodality, 273-286
    - BLIP-2, 273-277
    - chat-based prompting, 283-286
    - image captioning, 280-283
    - preprocessing images, 278
    - preprocessing text, 279
  - prompt engineering, 167-172
    - choosing models, 167
    - controlling output, 170-172
    - loading models, 168-169
  - topic modeling, 160-163
- text-in-text-out model, 74
- Text-to-Text Transfer Transformer (T5), 128-131
- thenlper/gte-small model, 140
- %%timeit magic command, 85
- TinyLlama, 367, 385
- tokenization-free encoding, 45
- tokens and tokenizers, 33, 37-61
  - bag-of-words model, 6
  - comparing trained tokenizers, 46-54
    - BERT base model (cased), 48
    - BERT base model (uncased), 47
    - Flan-T5, 50
    - Galactica, 52
    - GPT-2, 49
    - GPT-4, 50
    - Phi-3 and Llama 2, 53
    - StarCoder2, 51
  - decoding strategy, 79-81
  - downloading and running LLMs, 39-42
  - forward pass, 76-77
  - input preparation, 38
  - masked language modeling, 129
  - parallel token processing and context size, 81-83
  - special tokens, 47
  - task-specific representation model, 117, 127
  - text breakdown, 43
  - text-focused versus code-focused models, 56
  - token embeddings, 57-61, 77, 106
    - creating contextualized word embeddings, 58-61
    - tokenizer's vocabulary and, 57
  - token spans, 129
  - tokenization schemes, 44-46
    - byte tokens, 45
    - character tokens, 45
    - subword tokens, 44
    - word tokens, 44
  - tokenizer properties, 55-56
    - datasets, 56
    - methods, 55
    - parameters, 55
    - white space characters, 50
  - tone of voice, in text-generation prompts, 178
  - topic modeling, 138, 146-163
    - BERTopic, 148-155
    - representation blocks, 156-163
  - top\_k parameter, 172
  - top\_p parameter, 171, 188
  - train splits, 113
  - TrainingArguments class, 327

- transfer learning, 19
- Transformer architecture, 15-18, 73-107
  - attention layer, 79, 86
  - decoding strategy, 79-81
  - feedforward layer, 86
  - forward pass components, 76-79
  - inputs and outputs of, 74-76
  - keys and values cache, 83-85
  - optimizing attention, 98
  - parallel token processing and context size, 81-83
  - recent improvements to, 95-105
    - more efficient attention, 96-100
    - positional embeddings, 102-104
    - Transformer blocks, 101
  - Transformer blocks, 85-94
    - attention calculation, 91-93
    - attention layer, 88
    - attention mechanism, 89-91
    - feedforward neural networks, 87
    - self-attention and relevance scoring, 93-94
  - Vision Transformer, 260-262
- transparency and accountability, 28
- tree-of-thought, 189-191
- TruthfulQA, 374, 376
- TSDAE (Transformer-Based Sequential Denoising Auto-Encoder)
  - for domain adaptation, 320
  - overview of, 316-320

## U

- UltraChat dataset, 368
- UMAP (Uniform Manifold Approximation and Projection), 141
- unigram language model, 50
- unk\_token [UNK], 47
- use\_cache parameter, 84

## V

- valid output, verifying, 192
- validation splits, 113
- vector databases
  - dense retrieval, 229

- nearest neighbor search versus, 238
- retrieval-augmented generation, 253
- video random-access memory (VRAM), xiv, 29
- visualization
  - BERTopic, 155
  - cluster analysis, 144
  - dimensionality reduction and, 145
- ViT (Vision Transformer), 260-262, 274
- vocabulary, of tokenizers, 55, 57, 77, 106
- VRAM (video random-access memory), xiv, 29

## W

- warmup\_ratio parameter, 387
- warmup\_steps argument, 299
- Weaviate, 239
- whitespace characters, 50
- windowed conversation buffer memory, 212-214, 217
- word embeddings, 63-67
  - pretrained, 63
  - word2vec algorithm and contrastive training, 64-67
- word tokens, 44
- word-level metrics, in generative model evaluation, 374
- word2vec algorithm, 8, 10-12
  - contrastive training and, 64-67, 293
  - embedding songs, 67
- WordPiece, 43
  - cased BERT base model, 48
  - uncased BERT base model, 47
- <work> token, 53

## Y

- Year of Generative AI, 23-24

## Z

- zero-shot classification, 123-126
  - CLIP, 265
  - SetFit, 339
- zero-shot prompting
  - chain-of-thought, 187
  - in-context learning, 181

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## Colophon

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The animal on the cover of *Hands-On Large Language Models* is a red kangaroo (*Osphranter rufus*). They are the largest of all kangaroos, with a body length that can get up to a little over 5 feet and a tail as long as 3 feet. They are very fast and can hop to speeds over 35 miles per hour. They can jump 6 feet high and leap a distance of 25 feet in a single bound. The position of their eyes allows them see up to 300 degrees.

Red kangaroos are named after the color of their fur. While the name makes sense for the males—they have short, red-brown fur—females are typically more of a blue-grey color with a tinge of brown throughout. The red color in their fur comes from a red oil excreted from the glands in their skin. Because of their color, Australians refer to male red kangaroos as “big reds.” However, because females are faster than males, they are often called “blue fliers.”

Preferring open, dry areas with some trees for shade, red kangaroos can be found across Australia’s mainland except in the upper north, lower southwest, and east coast regions of the country. Surrounding environmental conditions can affect reproduction. Because of this, females can pause or postpone pregnancy or birth until conditions are better. They often use this ability to delay birth of a new baby (joey) until the previous one has left their pouch.

The cover illustration is by Karen Montgomery, based on an antique line engraving from *Cassell’s Popular Natural History*. The series design is by Edie Freedman, Ellie Volckhausen, and Karen Montgomery. The cover fonts are Gilroy Semibold and Guardian Sans. The text font is Adobe Minion Pro; the heading font is Adobe Myriad Condensed; and the code font is Dalton Maag’s Ubuntu Mono.

The background of the entire page is a vibrant red-to-orange gradient. Overlaid on this are several large, semi-transparent, overlapping circles in various shades of red and orange, creating a dynamic, organic feel.

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