Latent Dirichlet Allocation for Domain-Specific Language Modeling in Underwater Biology

In this study, I constructed and preprocessed a specialized text corpus to support the development of domain-specific word embeddings for underwater biology, with a focus on coral reef ecosystems. Wikipedia articles pertaining to key reef-building coral taxa—Acropora, Black Coral, Montipora, Pocillopora, Porites, and Scleractinia—were selected to ensure topical relevance and scientific depth. The goal of this corpus construction is to enhance the semantic precision of large language models in marine and ecological contexts.

Raw textual data were tokenized using a regular expression-based tokenizer to extract alphanumeric word units. Standard English stopwords were removed to reduce lexical noise and focus on semantically meaningful content. Tokens were filtered to include only alphabetic characters, and each cleaned article was stored as a processed text unit for downstream analysis. Throughout the preprocessing pipeline, I computed key corpus statistics, including the number of articles, paragraph counts, total token counts, and the vocabulary size (i.e., number of unique tokens).

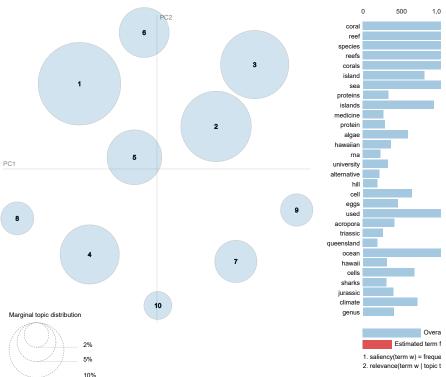
The cleaned and tokenized corpus will serve as input for subsequent vectorization and topic modeling procedures, such as TF-IDF transformation and Latent Dirichlet Allocation (LDA), to uncover latent semantic structures within the data. These representations aim to support the training of fine-tuned language models for marine biology, ultimately contributing to improved performance in domain-specific information retrieval, question answering, and knowledge discovery applications.

```
import os
from nltk import ne_chunk, pos_tag, word_tokenize
from sklearn.feature_extraction.text import TfidfVectorizer, CountVectorizer
from nltk.tokenize import RegexpTokenizer
from sklearn.decomposition import LatentDirichletAllocation
from nltk.corpus import stopwords
import pyLDAvis
from \ pyLDAvis.lda\_model \ import \ prepare \ as \ lda\_prepare
cleaned_text=[]
tokenizer = RegexpTokenizer(r"\w+")
# create English stop words list
englishfile = open("englishvocabulary.txt", encoding="utf8")
engstop = englishfile.read()
engstop = set(stopwords.words("english"))
files=os.listdir('wiki_corpus')
token_set=set()
file_count=0
paragraph count=0
word count=0
center articles="[Acropora, Black Coral, Montipora, Pocillopora, Porites, Scleractinia]"
for file in files:
    with open(f'wiki_corpus/{file}','r') as f:
        raw=f.read().lower()
    tokens=tokenizer.tokenize(raw)
    tokens = [i for i in tokens if not i in engstop and i.isalpha()]
    token set.update(tokens)
    file count+=1
    paragraph count+=raw.count('\n')
    word count+=len(tokens)
    newfile=' '.join(tokens)
    cleaned_text.append(newfile)
print("Articles: ", file_count)
print("Paragraphs:", paragraph_count)
print("Tokens:", word count)
print("Unique tokens:", len(token_set))
print("Central Articles:", center_articles)
→ Articles: 492
     Paragraphs: 12924
     Tokens: 500297
     Unique tokens: 41163
     Central Articles: [Acropora, Black Coral, Montipora, Pocillopora, Porites, Scleractinia]
count=CountVectorizer(lowercase=True,stop words='english', token pattern=r'[A-Za-z][A-Za-z][A-Za-z]+')
docmatrix=count.fit transform(cleaned text)
lda=LatentDirichletAllocation(n_components=n,random_state=42)
lda.fit(docmatrix)
features=count.get_feature_names_out()
topics=[]
for index,topic in enumerate(lda.components ):
    words=[features[i] for i in topic.argsort()[:-w - 1:-1]]
    topics.append(words)
pyLDAvis.enable_notebook()
lda_prepare(lda, docmatrix, count, mds='mmds')
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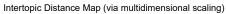
Intertopic Distance Map (via multidimensional scaling)

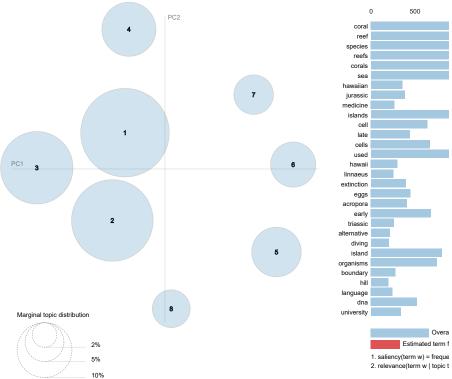


```
n,w =8, 10
 count = Count \\ Vectorizer (lower case = True, stop\_words = 'english', token\_pattern = r'[A-Za-z][A-Za-z] \\ [A-Za-z] \\ 
 docmatrix=count.fit_transform(cleaned_text)
 lda=LatentDirichletAllocation(n_components=n,random_state=42)
lda.fit(docmatrix)
 features=count.get_feature_names_out()
 topics=[]
 for index, topic in enumerate(lda.components_):
                     words=[features[i] for i in topic.argsort()[:-w - 1:-1]]
                      topics.append(words)
pyLDAvis.enable_notebook()
lda_prepare(lda, docmatrix, count, mds='mmds')
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 $\lambda = 1$





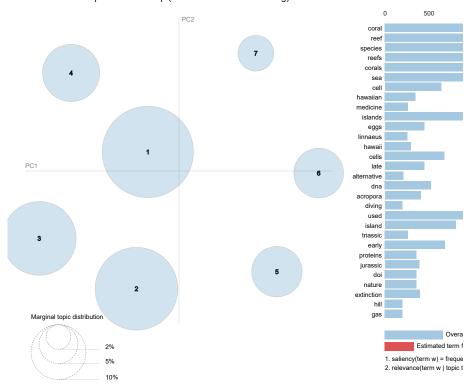
```
n,w =7, 10
count = Count \\ Vectorizer (lower case = True, stop\_words = 'english', token\_pattern = r'[A-Za-z][A-Za-z] \\ [A-Za-z] \\ 
 docmatrix=count.fit_transform(cleaned_text)
 lda=LatentDirichletAllocation(n_components=n,random_state=42)
lda.fit(docmatrix)
 features=count.get_feature_names_out()
 topics=[]
 for index, topic in enumerate(lda.components_):
                     words=[features[i] for i in topic.argsort()[:-w - 1:-1]]
                      topics.append(words)
pyLDAvis.enable_notebook()
lda_prepare(lda, docmatrix, count, mds='mmds')
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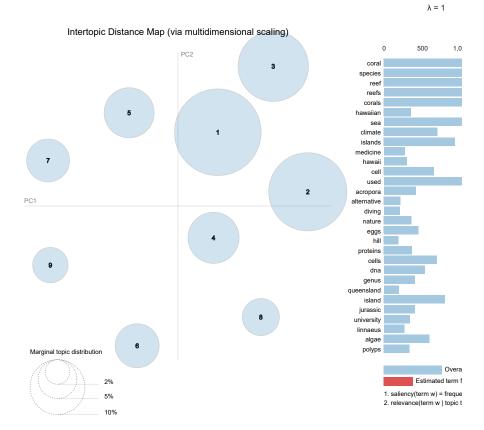
 $\lambda = 1$

Intertopic Distance Map (via multidimensional scaling)



```
n,w =9, 10
 count = Count \\ Vectorizer (lower case = True, stop\_words = 'english', token\_pattern = r'[A-Za-z][A-Za-z] \\ [A-Za-z] \\ 
 docmatrix=count.fit_transform(cleaned_text)
lda=LatentDirichletAllocation(n_components=n,random_state=42)
lda.fit(docmatrix)
 features=count.get_feature_names_out()
topics=[]
 for index, topic in enumerate(lda.components_):
                     words=[features[i] for i in topic.argsort()[:-w - 1:-1]]
                      topics.append(words)
pyLDAvis.enable_notebook()
lda_prepare(lda, docmatrix, count, mds='mmds')
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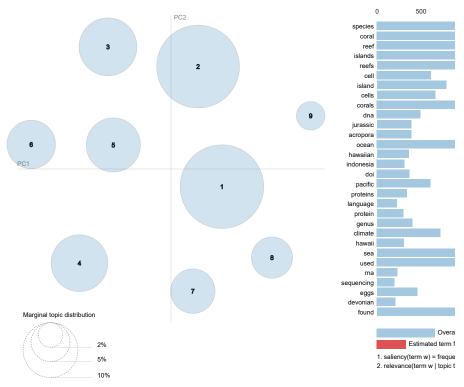


```
n,w =9, 10
count=CountVectorizer(lowercase=True, token_pattern=r'[A-Za-z][A-Za-z][A-Za-z]+')
docmatrix=count.fit_transform(cleaned_text)
lda=LatentDirichletAllocation(n_components=n,random_state=42)
lda.fit(docmatrix)
features=count.get_feature_names_out()
topics=[]
for index, topic in enumerate(lda.components_):
    words=[features[i] for i in topic.argsort()[:-w - 1:-1]]
    topics.append(words)
pyLDAvis.enable_notebook()
lda_prepare(lda, docmatrix, count, mds='mmds')
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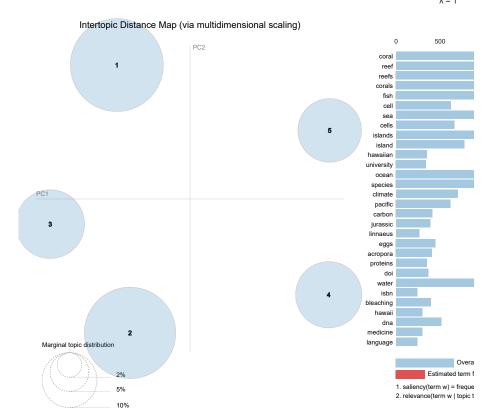
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Intertopic Distance Map (via multidimensional scaling)

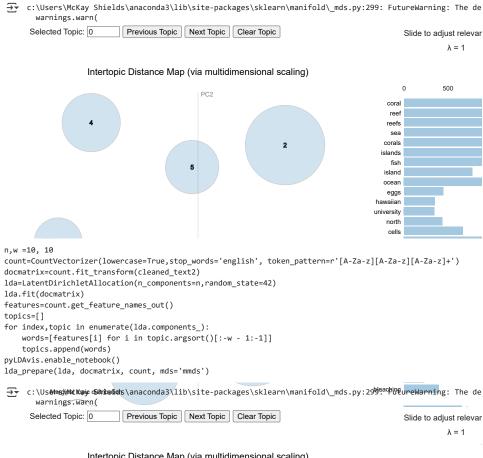


```
cleaned_text2=[]
 tokenizer = RegexpTokenizer(r"\w+")
 # create English stop words list
englishfile = open("englishvocabulary.txt", encoding="utf8")
engstop = englishfile.read()
engstop = set(stopwords.words("english"))
files=os.listdir('wiki_corpus')
for file in files:
               with open(f'wiki_corpus/{file}','r') as f:
                               lines=f.readlines()
                for line in lines:
                               raw=line.lower()
                               tokens=tokenizer.tokenize(raw)
                              tokens = [i for i in tokens if not i in engstop and i.isalpha()]
newfile=' '.join(tokens)
                               cleaned_text2.append(newfile)
n,w =5, 10
count = Count \lor count \lor count = count \lor count \lor count = count \lor count = count \lor coun
docmatrix=count.fit_transform(cleaned_text2)
{\tt lda=LatentDirichletAllocation(n\_components=n, random\_state=42)}
lda.fit(docmatrix)
 features=count.get_feature_names_out()
 topics=[]
 for index,topic in enumerate(lda.components_):
               words=[features[i] for i in topic.argsort()[:-w - 1:-1]]
                topics.append(words)
pyLDAvis.enable_notebook()
lda_prepare(lda, docmatrix, count, mds='mmds')
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```
n,w =7, 10
 count = Count \\ Vectorizer (lower case = True, stop\_words = 'english', token\_pattern = r'[A-Za-z][A-Za-z] \\ [A-Za-z] \\ 
 docmatrix=count.fit_transform(cleaned_text2)
 lda=LatentDirichletAllocation(n_components=n,random_state=42)
lda.fit(docmatrix)
 features=count.get_feature_names_out()
 topics=[]
 for index, topic in enumerate(lda.components_):
                     words=[features[i] for i in topic.argsort()[:-w - 1:-1]]
                      topics.append(words)
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lda_prepare(lda, docmatrix, count, mds='mmds')
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Intertopic Distance Map (via multidimensional scaling)