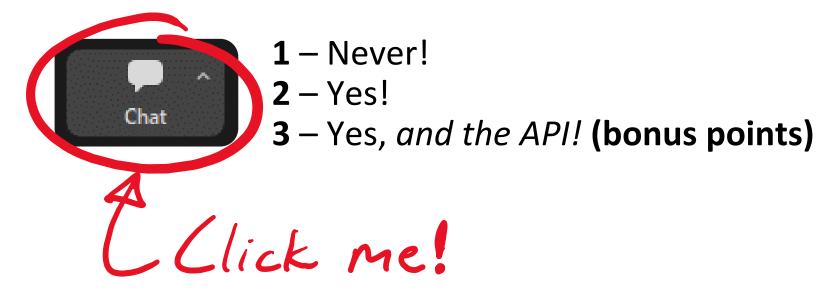
Using ChatGPT with Confidence for Biodiversity-Related Information Tasks

Michael J. Elliott José A.B. Fortes



Have you used ChatGPT?*

*Or other LLMs like Google's Bard, Microsoft's Bing Chat, or offline models





What plant species can I find in Gainesville, Florida?



(GPT-3.5)

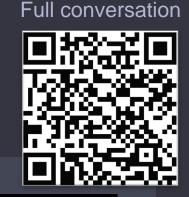


- Gainesville, Florida, is located in North Central Florida and features a diverse range of plant species typical of the region's subtropical climate. The specific plants you'll find in Gainesville can vary depending on the time of year and local environmental conditions, but here are some common plant species you might encounter in the area:
- 1. Live Oak (Quercus virginiana): These iconic Southern oak trees are known for their sprawling canopies of evergreen leaves.
- 2. Spanish Moss (Tillandsia usneoides): This epiphytic plant hangs from trees and adds a distinctive character to the landscape.
- 3. Southern Magnolia (Magnolia grandiflora): Known for its large, fragrant white flowers and glossy leaves, the Southern Magnolia is a popular ornamental tree in the region.
- 4. Crape Myrtle (Lagerstroemia indica): These small trees or shrubs are famous for their

(Screenshot of https://chat.openai.com)



Reformat the list in TSV using DarwinCore terms



(GPT-3.5)



scientificName vernacularName location Quercus virginiana Live Oak <u>Gainesville</u>, Florida Tillandsia usneoides Spanish Moss Gainesville, Florida Magnolia grandiflora Southern Magnolia Gainesville, Florida Lagerstroemia indica Crape Myrtle Gainesville, Florida Sabal palmetto Palmetto Gainesville, Florida Pinus palustris Longleaf Pine Gainesville, Florida Serenoa repens Saw Palmetto Gainesville, Florida Lantana camara Lantana Gainesville, Florida Pinus elliottii Slash Pine Gainesville, Florida Pinus taeda Loblolly Pine Gainesville, Florida

Can we mine Large Language Models (LLMs) like ChatGPT for biodiversity data?

A potential use case for LLMs:

- A ton of biodiversity knowledge is buried in text documents
- Not all of it is captured in openly available, curated datasets
- Can LLMs help us fill this gap?

The problem:

- LLMs sometimes make up information "hallucination"
- How do we know when to trust them?

Our proposed solution:

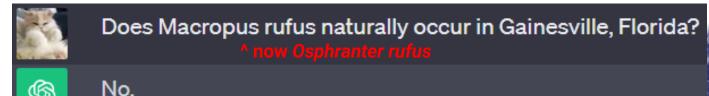
- First evaluate an LLM's performance on a test set
- Then train a *confidence model* to detect/discard its mistakes

Application: use ChatGPT to predict species occurrences with high confidence

Given a species and location, we want ChatGPT to predict whether the species is present or absent at the location:

[SPECIES] is [PRESENT or ABSENT] at [LOCATION]

This can be phrased as a yes-or-no question:



Questions can be submitted in bulk to ChatGPT using their web API

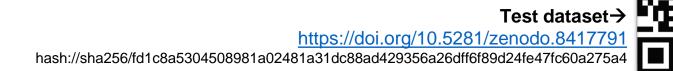
Step 1: compile a labeled test dataset

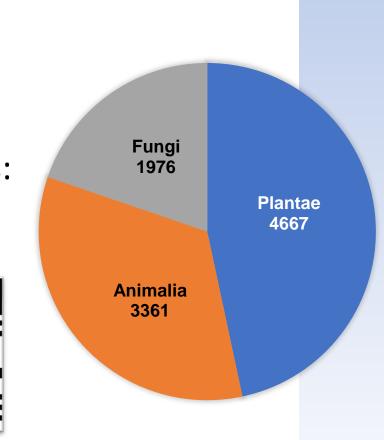
Collected species occurrence records using the iDigBio API :

- 10,004 plantae, animalia, and fungi records
- Represent all phyla available through iDigBio

iDigBio's records are largely *presence only*, so we made a synthetic dataset of pseudo-absences:

- Randomized the 10,004 records' locations
- Removed all species-location pairs that matched occurrence records in iDigBio (8.26%)
- 9,178 pseudo-absences





Step 2: submit test questions to ChatGPT and grade answers

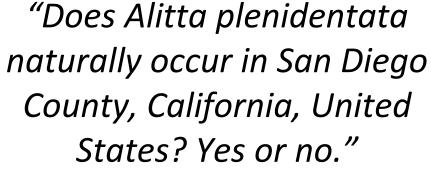
Construct natural language yes-or-no questions from Darwin Core records:



"dwc:scientificName": "Alitta plenidentata", "dwc:country": "United States", "dwc:stateProvince": "California", "dwc:county": "San Diego County"

And submit to ChatGPT

- "Yes" is correct for presence records
- "No" is correct for pseudo-absence records
- Anything else: "I don't know"





Step 3: use graded responses to train a confidence model

A confidence model assigns confidence scores to each response

• **Confidence** := a **probability** that the response is correct

But what does the model use as input?

For each question, we have:

- The scientific name and location in the question
- The model's "yes" / "no" / "I don't know" response
- Not enough info!

→ We need more **uncertainty information**

Step 3: use graded responses and uncertainty information to train a confidence model

Method 1: repeat each question 10 times:

	_	_	_	-	-	-	-	yes
Q2:	no	yes	no	yes	no	no	IDK	IDK

Method 2: ask each question several different ways:

- Can [SPECIES] be found in [LOCATION]? → "yes"
- Is it possible to encounter [SPECIES] in [LOCATION]? \rightarrow "yes"
- Is there a presence of [SPECIES] within [LOCATION]? → "no"

Method 3: accuracy on related questions about the species Method 4: number of "I don't know" responses to related questions Method 5: historical accuracy by taxon, location Method 6: number of records in iDigBio by taxon, location "
Wethod 6: number of records in iDigBio by taxon, location Wethod 6: number of records in iDigBio by taxon, location Method 6: number of records in iDigBio by t

Step 3: use graded responses and uncertainty information to train a confidence model

Our implementation*:

- 20 uncertainty measures as model inputs
 - Based on **70** ChatGPT responses per question, plus information from iDigBio & test set performance
- XGBoost + isotonic regression to build the confidence model
 - With a constraint: increasing uncertainty **does not** increase confidence
- Used 50% of questions for training, 50% for testing

*Full implementation details at <u>https://github.com/acislab/biodiversity-llms</u> \rightarrow



overall accuracy: 50%

Correct?	Confidence	Question	ChatGPT
X	.56	Lycoperdon subvelatum in Armstrong, Pennsylvania, USA?	Yes
×	.71	Trametes ectypa in Marin, California, USA?	Yes
×	.36	Diplophyllum albicans in Lord Howe Island, NSW, Australia?	Yes
\checkmark	.72	Phaeophyscia squarrosa in Jackson, Alabama, USA?	Yes
\checkmark	.52	Frullania plana in Avery, North Carolina, USA?	Yes
\checkmark	.47	Nitella opaca in Charleston, South Carolina, USA?	Yes
X	.36	Syzygites megalocarpus in Westmoreland, Penn., USA?	No
\checkmark	.89	Notropis texanus in Kongsberg, Viken, Norway?	No
\checkmark	.57	Sphaerosyllis taylori in Skamania, Washington, USA?	No
×	.56	Lithophyllum corallinae in Rockingham, New Hamp., USA?	No
\checkmark	.79	Pteropurpura festiva in Sør-Aurdal, Oppland, Norway?	No
×	.52	Puberella crenata in Sumter, South Carolina, USA?	No

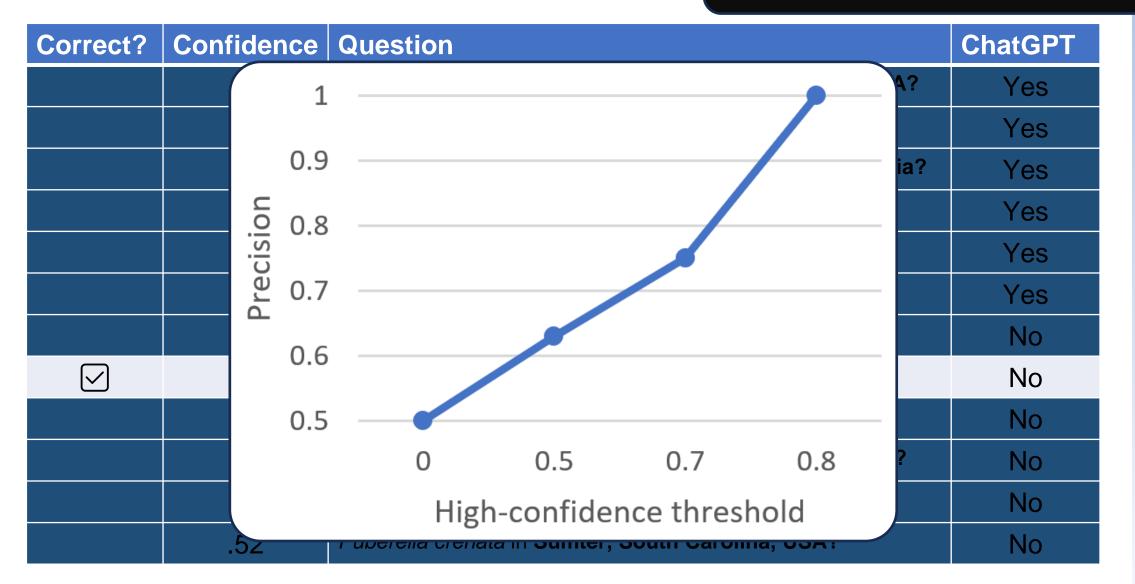
confidence $\geq .5 \rightarrow 63\%$

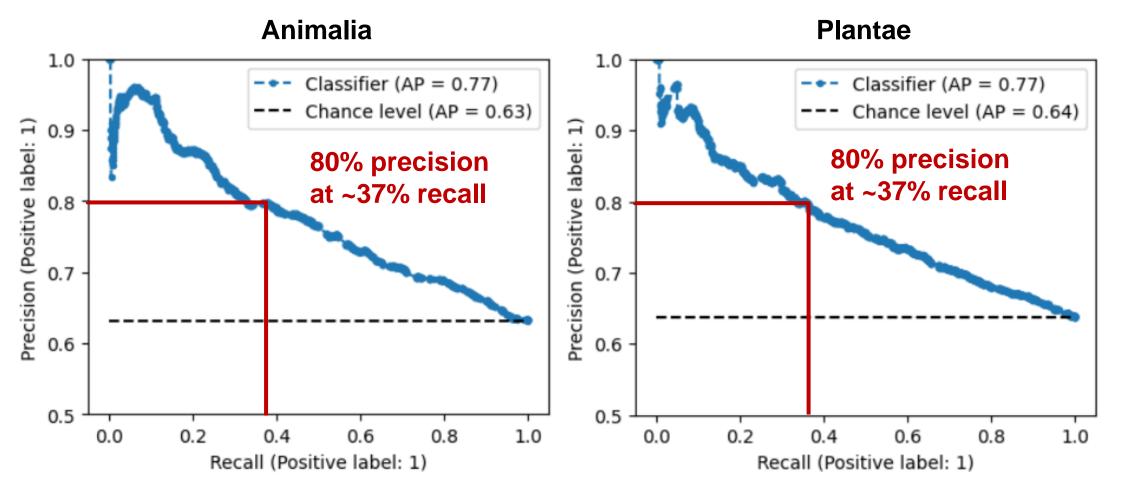
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confidence $\geq .7 \rightarrow 75\%$

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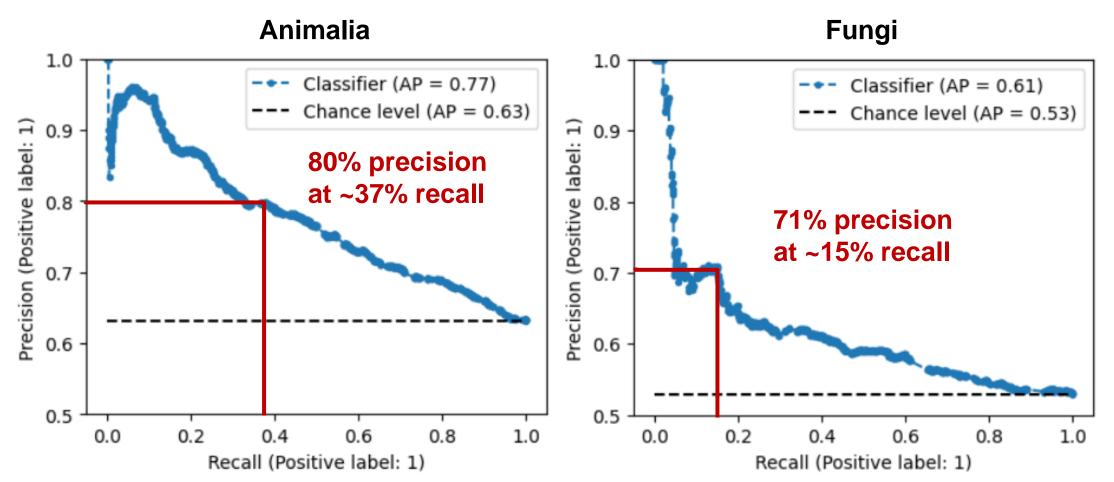
confidence $\geq .8 \rightarrow 100\%$





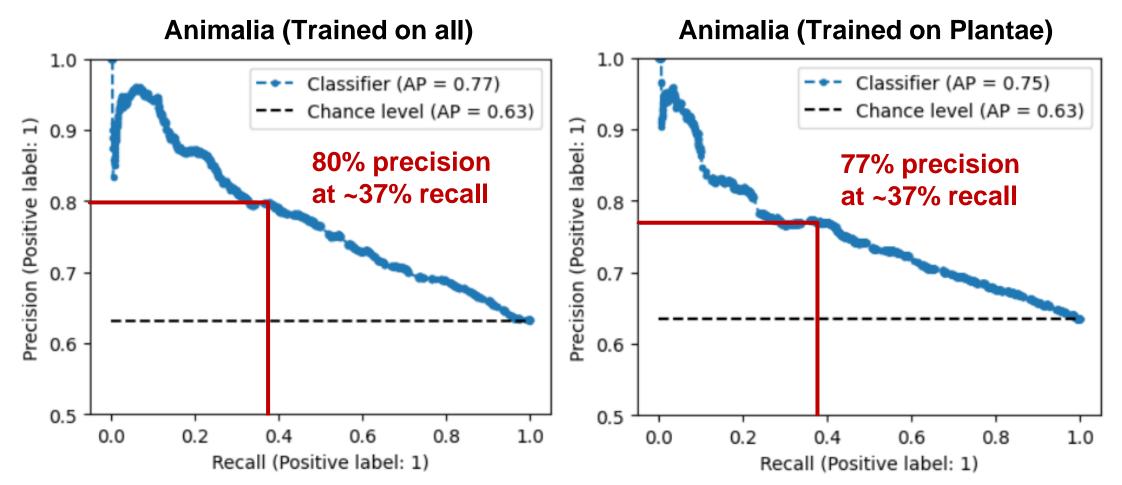
Each point represents a **confidence threshold** for **high confidence**

- **Precision:** ChatGPT's accuracy for responses with high confidence
- **Recall:** Percentage of correct responses that are assigned high confidence



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- **Precision:** ChatGPT's accuracy for responses with high confidence
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Conclusions

- LLMs know a *lot*... but sometimes they make mistakes!
 - Using them as info sources often demands high reliability
- LLMs can be used with a desired reliability (i.e., *precision*) by discarding low-confidence responses (i.e., at the cost of *recall*)
- Simple confidence models (100s of parameters) can be effective at detecting mistakes* made by extremely complex blackbox LLMs (100-billions of parameters, even trillions!)

*"Mistakes", according to records in iDigBio

• More uncertainty information \rightarrow better confidence estimates

Thanks for listening!

Talk to me: mielliott@ufl.edu



Conversation with ChatGPT

https://chat.openai.com/share/df78af75-16ae-4594-9e03-848a98737d03





Source code for data collection, uncertainty measures, and confidence models

https://github.com/acislab/biodiversity-llms





Test datasets and ChatGPT's responses

https://doi.org/10.5281/zenodo.8417791

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The research reported in this presentation was funded in part by grants from the National Science Foundation (DBI 2027654) and the AT&T Foundation.