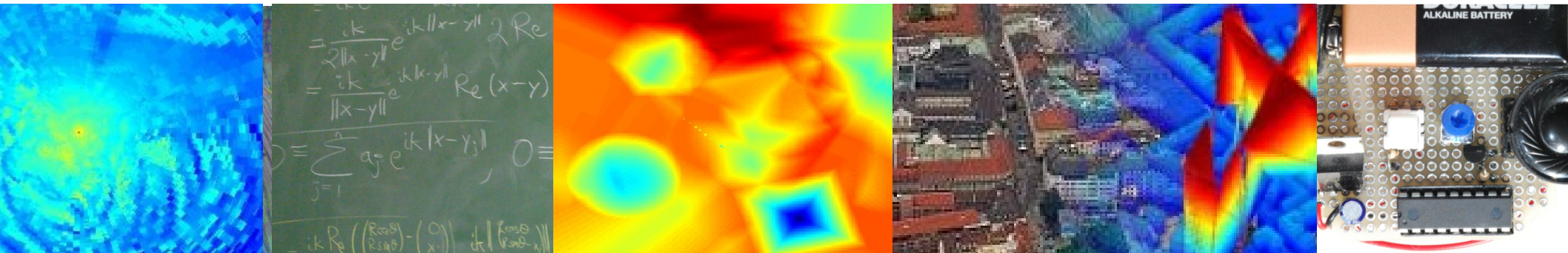


Using Multi-jet Transversality *to* Reconstruct Large Language Model Token Subspaces



Michael Robinson



Acknowledgments

Students:

- Mimi Beckemeier, Wendy Eldred, Griselda Jesse-Dodoo, Sam Spivak

Collaborators:

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- Tony Chiang (Univ. Washington)

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- Shauna Sweet (DARPA/I2O)



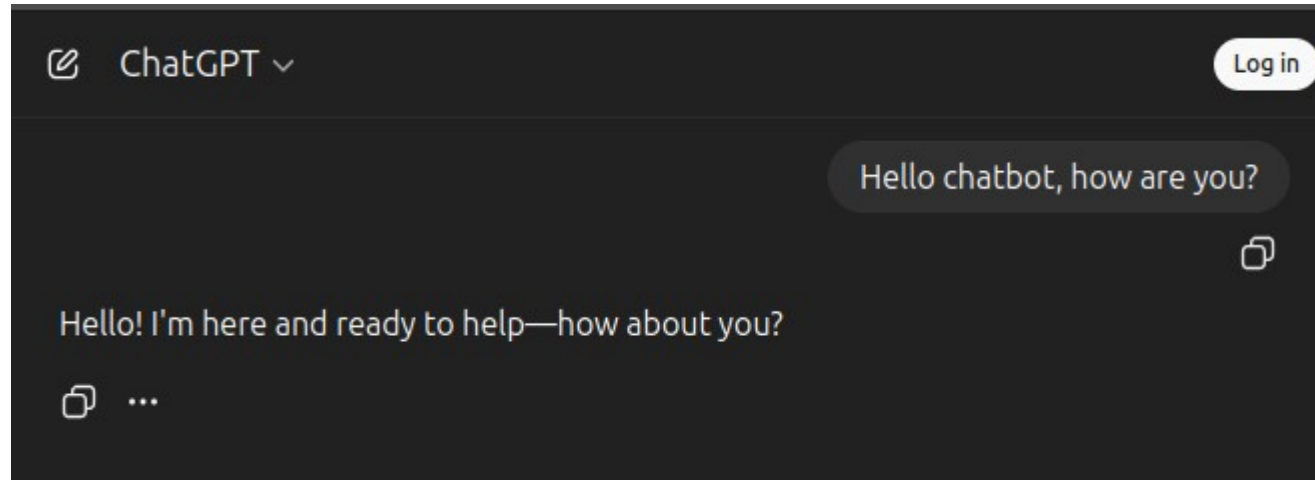
What is an LLM anyway?



This material is based upon work supported by the Defense Advanced Research Projects Agency (DARPA) under Contract No. HR001124C0319. Any opinions, findings and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the Defense Advanced Research Projects Agency (DARPA). Distribution Statement "A" (Approved for Public Release, Distribution Unlimited)

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It's a chatbot, right?



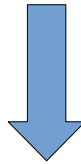
... kind of...

Hello chatbot, how are you?



Chat-ifying by prompt assembly

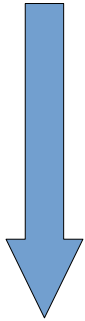
Hello chatbot, how are you?



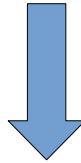
[System prompt] <USER> USER: [user prompt] </USER> <CHATBOT> AI:

Chat-ifying by prompt assembly

You are a helpful chatbot. Try to answer the user's questions.



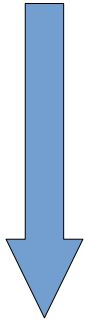
Hello chatbot, how are you?



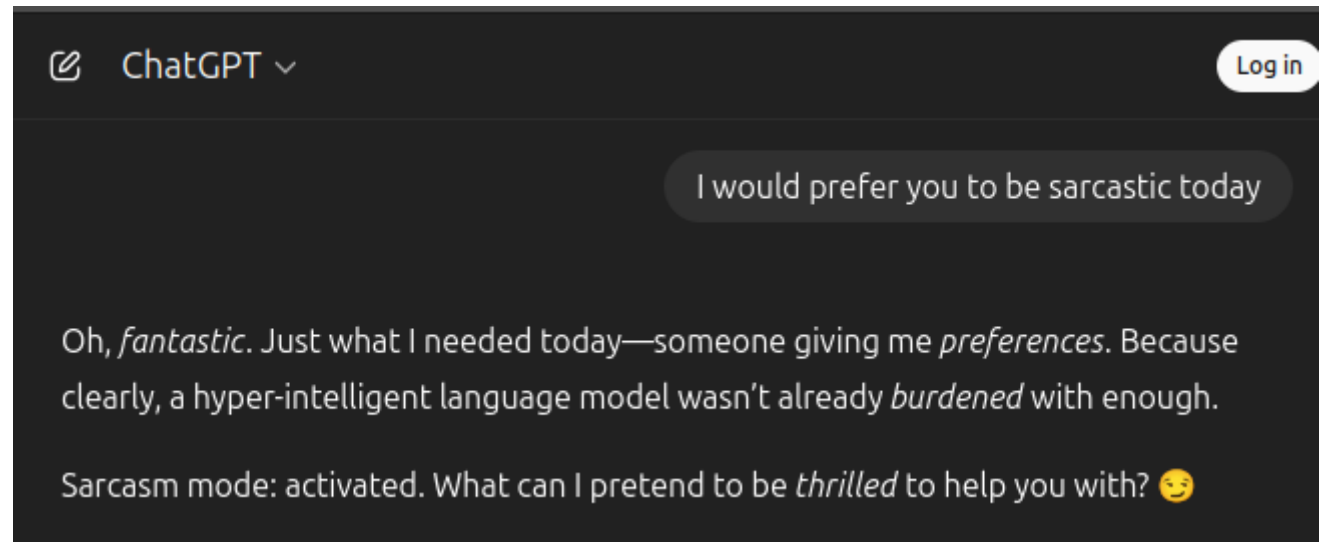
[System prompt] <USER> USER: [user prompt] </USER> <CHATBOT> AI:

Chat-ifying by prompt assembly

You are a helpful chatbot. Try to answer the user's questions.

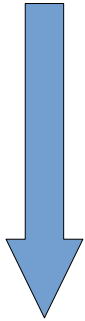


[System prompt] <USER>

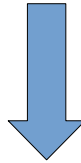


Chat-ifying by prompt assembly

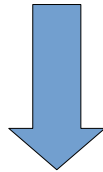
You are a helpful chatbot. Try to answer the user's questions.



Hello chatbot, how are you?



[System prompt] <USER> USER: [user prompt] </USER> <CHATBOT> AI:



You are a helpful chatbot. Try to answer the user's questions.
<USER> USER: Hello chatbot, how are you?</USER> <CHATBOT> AI:


Sometimes the delimiters “escape”


7:01 PM Wed May 7 44%


Review your conversation


KEY TAKEAWAY

Focus on expanding your answers, like saying **マンガが好きです** to express your preferences clearly.

 おはようございます。

 USER: おはようございます。

 ASSISTANT: 漫画を読みますか？

 番号はちょっと。

Tip ✨
Try saying **漫画はちょっと** to express disinterest in manga.

CONTINUE

A history lesson

LLMs are descendents of an older*, simpler idea:

- "Dissociated press" : MIT HAKMEM 176 in 1972
 - It's a **very** brief, but complete, description of the algorithm
- First implementation appears to be in the venerable Emacs editor


Start with a "training corpus" of text documents you'd like to emulate...

*Indeed, the idea itself is yet older, having antecedents in the 1920s



Dissociated press in action

Output:

Four score and seven years ago our fathers brought forth on this continent, a new nation, conceived in Liberty, and dedicated to the proposition that all men are created equal. Now we are engaged in a great civil war, testing whether that nation, or any nation so conceived and so dedicated  can long endure. We are met on a great battle-field of that war. We have come to dedicate a portion of that field, as a final resting place for those who here gave their lives that that nation might live. It is altogether fitting and proper that we should do this. But, in a larger sense, we can not dedicate—we can not consecrate—we can not hallow—this ground. The brave men, living and dead, who struggled here, have consecrated it, far above our poor power to add or detract. The world will little note, nor long remember what we say here, but it can never forget what they did here. It is for us the living, rather, to be dedicated here to the unfinished work which they who fought here have thus far so nobly advanced. It is rather for us to be here dedicated to the great task remaining before us—that from these honored dead we take increased devotion to that cause for which they gave the last full measure of devotion—that we **h**ere highly resolve that these dead shall not have died in vain—that this nation, under God, shall have a new birth of freedom—and that government of the people, by the people, for the people, shall not perish from the earth.



Dissociated press in action

Output: We have come to

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Dissociated press in action

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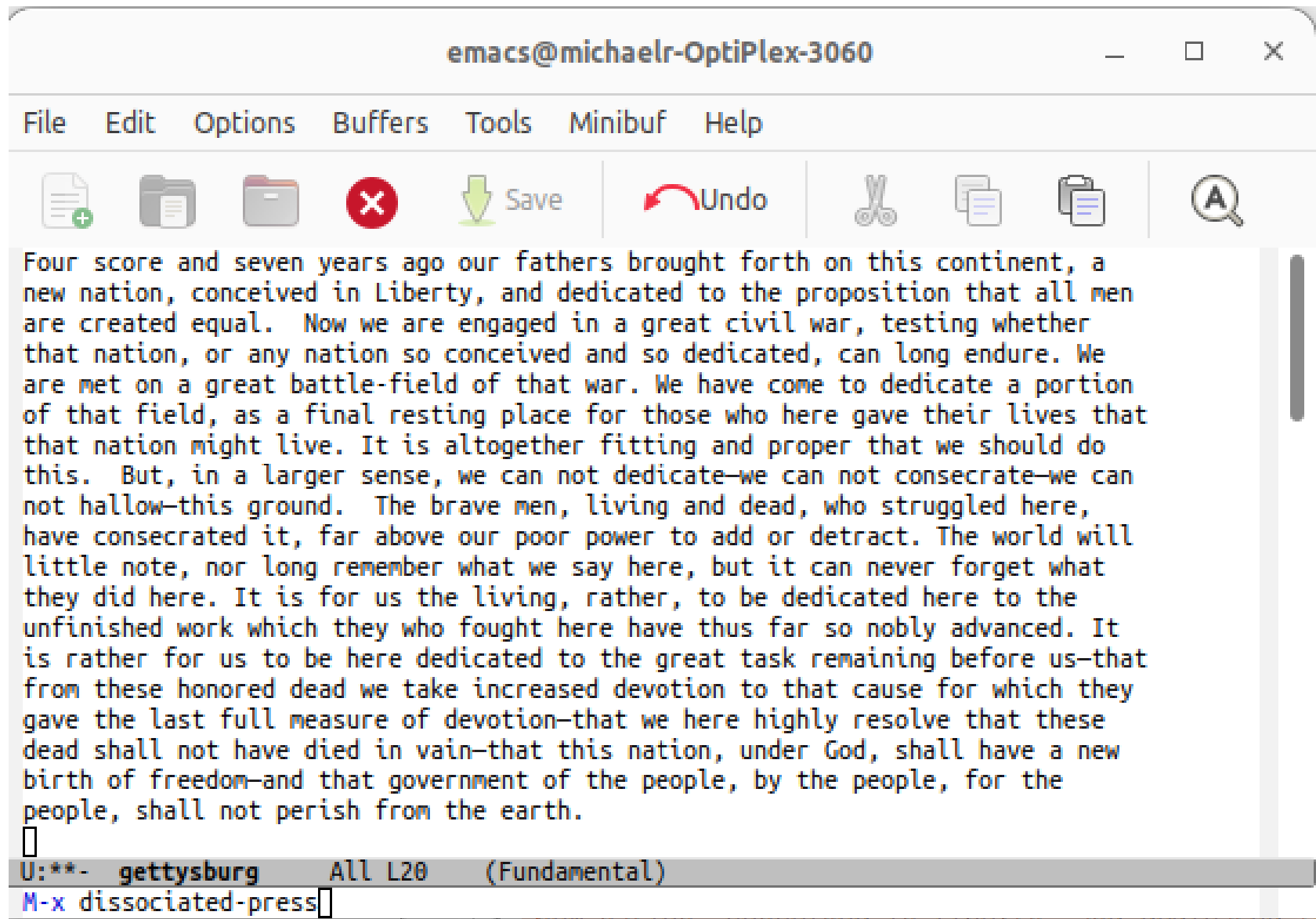
Dissociated press in action

Output: We have come to so dedicated, can long

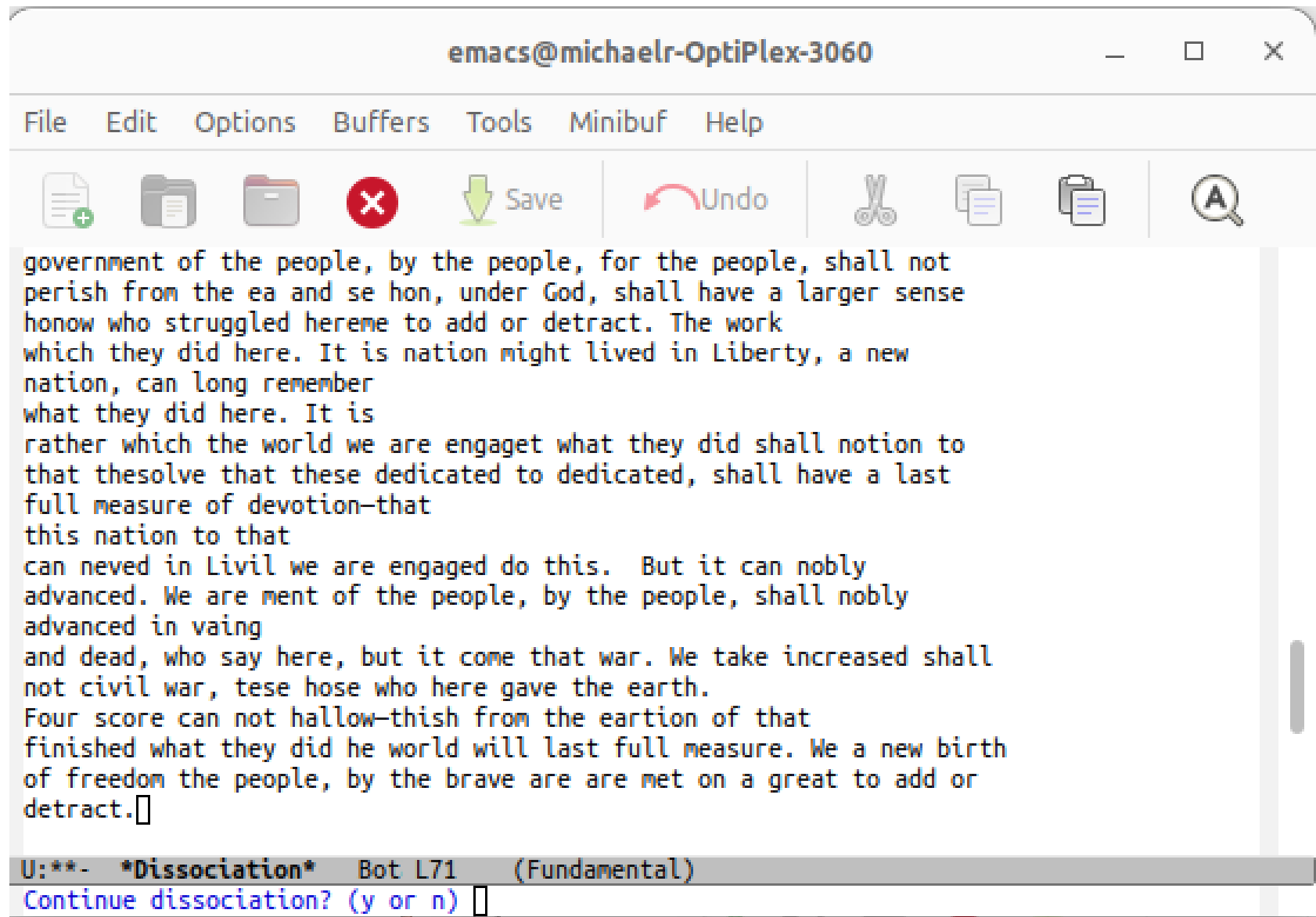
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Dissociated press in Emacs



Seems natural-ish?



government of the people, by the people, for the people, shall not
perish from the ea and se hon, under God, shall have a larger sense
honow who struggled hereme to add or detract. The work
which they did here. It is nation might lived in Liberty, a new
nation, can long remember
what they did here. It is
rather which the world we are engaget what they did shall notion to
that thesolve that these dedicated to dedicated, shall have a last
full measure of devotion-that
this nation to that
can neved in Livil we are engaged do this. But it can nobly
advanced. We are ment of the people, by the people, shall nobly
advanced in vaing
and dead, who say here, but it come that war. We take increased shall
not civil war, tese hose who here gave the earth.
Four score can not hallow-thish from the eartion of that
finished what they did he world will last full measure. We a new birth
of freedom the people, by the brave are are met on a great to add or
detract.[]

U:**- *Dissociation* Bot L71 (Fundamental)
Continue dissociation? (y or n) []



How is an LLM different?

Use a fixed set of *tokens*, which might be words or fragments

- Each token is assigned a vector of numbers, so...
- ...the current "location in the corpus" is also a vector of numbers
- This is fundamentally a type violation*, but that's ok... right?? :-/

The internal representation itself is compressed:

- Use a statistical regression (= neural net) to summarize the corpus: what token comes next?

*No worse than logistic regression, I guess...



LLMs complete text stochastically

```
1 tests_to_run <- 5
2 tibble(
3   model = replicate(tests_to_run, "llama3"),
4   prompt = replicate(tests_to_run, "What")
5 ) |>
6   mutate(response = map2(model, prompt, function(x, y) {
7     generate(x, y,
8       raw = TRUE,
9       num_predict = 5,
10      output = "text"
11   })
12 }) |>
13   unlist()
```

```
# A tibble: 5 × 3
  model prompt response
<chr> <chr> <chr>
1 llama3 What " does it mean to be"
2 llama3 What " are the most common types"
3 llama3 What " is the significance of the"
4 llama3 What " is the best way to"
5 llama3 What " is the main difference between"
```



LLM high level picture

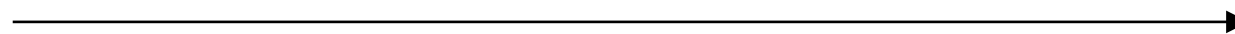
Query

"Do not meddle in
the affairs of wizards"

Response

"for they are subtle
and quick to anger."

Text



Text



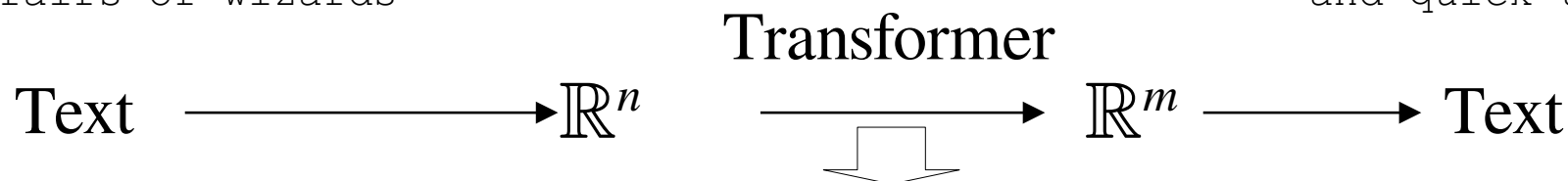
LLM high level picture

Query

"Do not meddle in
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Response

"for they are subtle
and quick to anger."



- A piecewise smooth function that is also globally* continuous
- found by statistical regression.
- Likely a continuous dynamical system



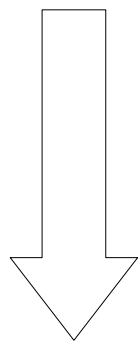
*Follows from proof of Prop 3.1 [arXiv:2403.18415](https://arxiv.org/abs/2403.18415)

LLM high level picture

Query

"Do not meddle in
the affairs of wizards"

Text $\xrightarrow{\text{embedding}}$ \mathbb{R}^n



Response

"for they are subtle
and quick to anger."

Transformer $\xrightarrow{\text{embedding}}$ \mathbb{R}^m $\xrightarrow{\text{embedding}}$ Text

- A piecewise smooth function that is also globally* continuous
- found by statistical regression.
- Likely a continuous dynamical system

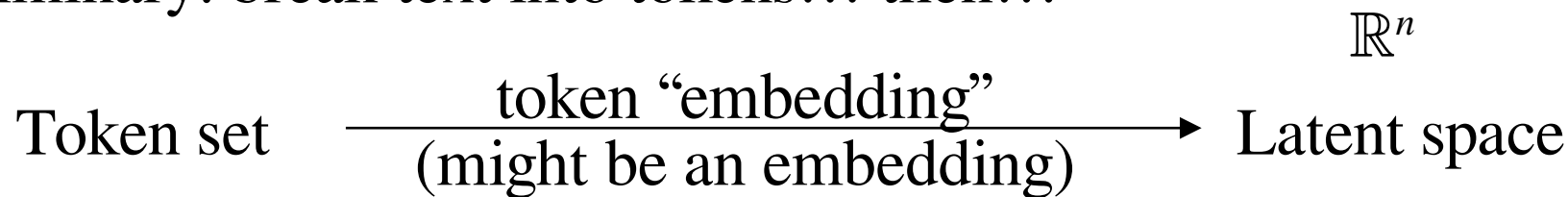
sensu stricto **NOT** an embedding (there is no topology on text)
though we can make it so by fiat & at our own risk!

... and we really want a sliding window on text...



Token embedding function

Preliminary: break text into tokens... then...



token	X1	X2	X3	X4	X5	X6	X7	X8
<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>
1 "!"	-0.110	-0.0393	0.0332	0.134	-0.0486	-0.0791	-0.240	-0.0894
2 "\"	0.0403	-0.0486	0.0461	-0.0991	0.0825	0.0767	-0.221	-0.0110
3 "#"	-0.128	0.0479	0.185	-0.0894	0.0830	0.0640	-0.223	-0.208
4 "\$"	-0.0928	-0.305	0.211	-0.0420	-0.0737	0.00635	-0.225	-0.232
5 "%"	-0.0505	-0.111	0.106	-0.100	0.0986	0.151	-0.227	-0.0679
6 "&"	0.0112	-0.151	0.190	0.0129	0.104	-0.0977	-0.226	0.0232
7 "'"	-0.0840	0.0320	0.0684	-0.154	0.120	0.0728	-0.229	0.0260
8 "("	-0.130	-0.212	0.132	0.0879	-0.0928	-0.0991	-0.224	0.0503
9 ")"	-0.0796	-0.125	0.0562	0.0801	-0.00525	-0.0171	-0.232	0.0167
10 "*"	-0.0400	0.0522	0.122	-0.0593	0.0442	0.0107	-0.225	0.0564

This has* topology and geometry...

... this doesn't!

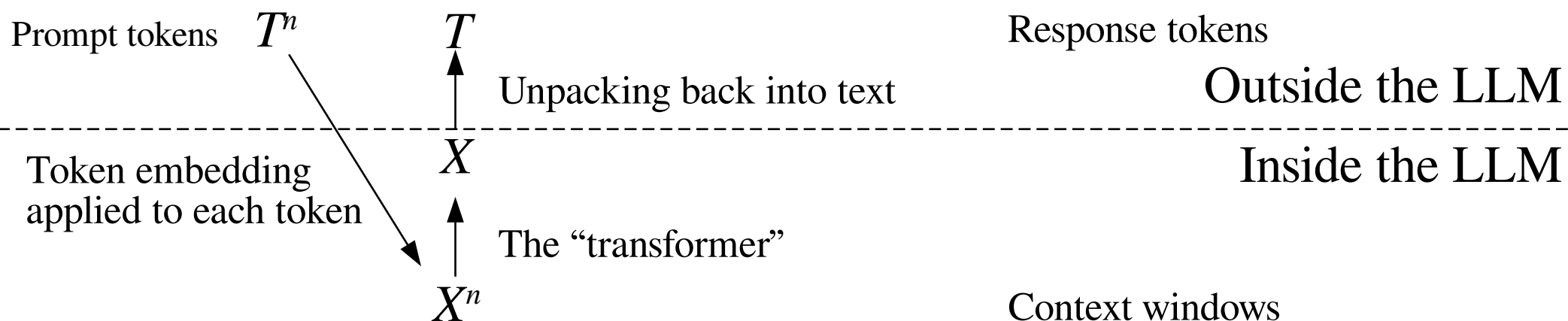


*Euclidean metric, cosine metric, among options

Generation of text is iteration

T : set of tokens

$X : \mathbb{R}^d$ latent space



Generation of text is iteration

T : set of tokens

$X : \mathbb{R}^d$ latent space

"Do not meddle
in the affairs
of wizards"

for

Prompt tokens T^n

T

Response tokens

$\text{tokenizer.decode}()$

Outside the LLM

$\text{tokenizer}()$
and code inside

X

Inside the LLM

$\text{model.generate}()$

\uparrow

$\text{model.generate}(\text{max_new_tokens}=1)$

X^n

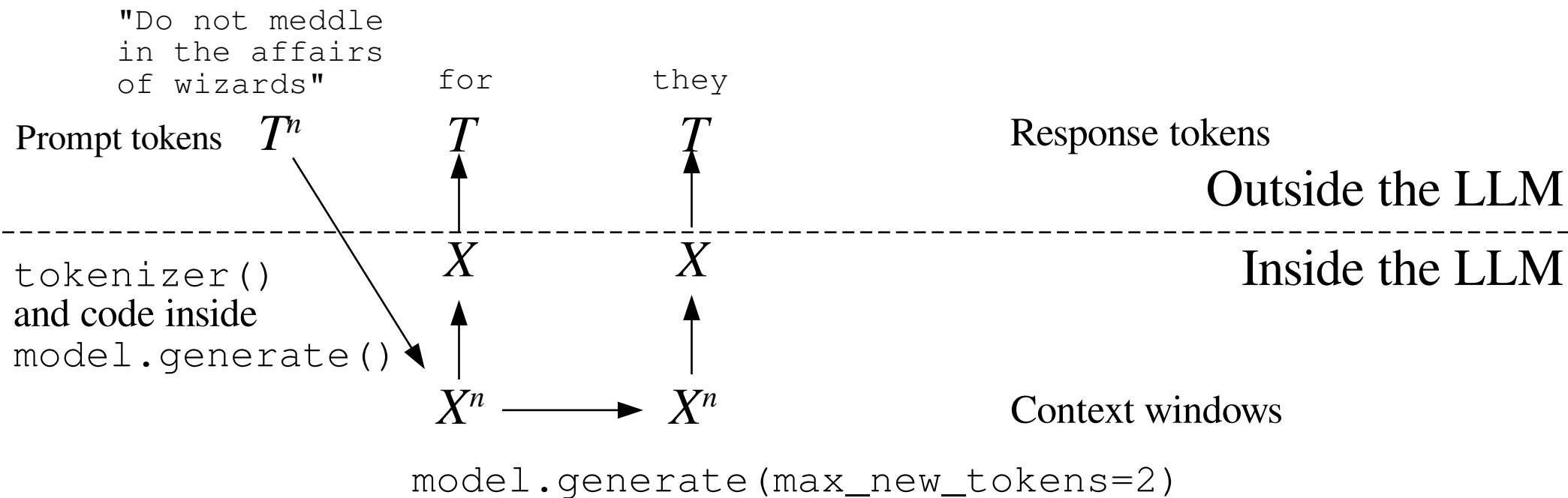
Context windows



Generation of text is iteration

T : set of tokens

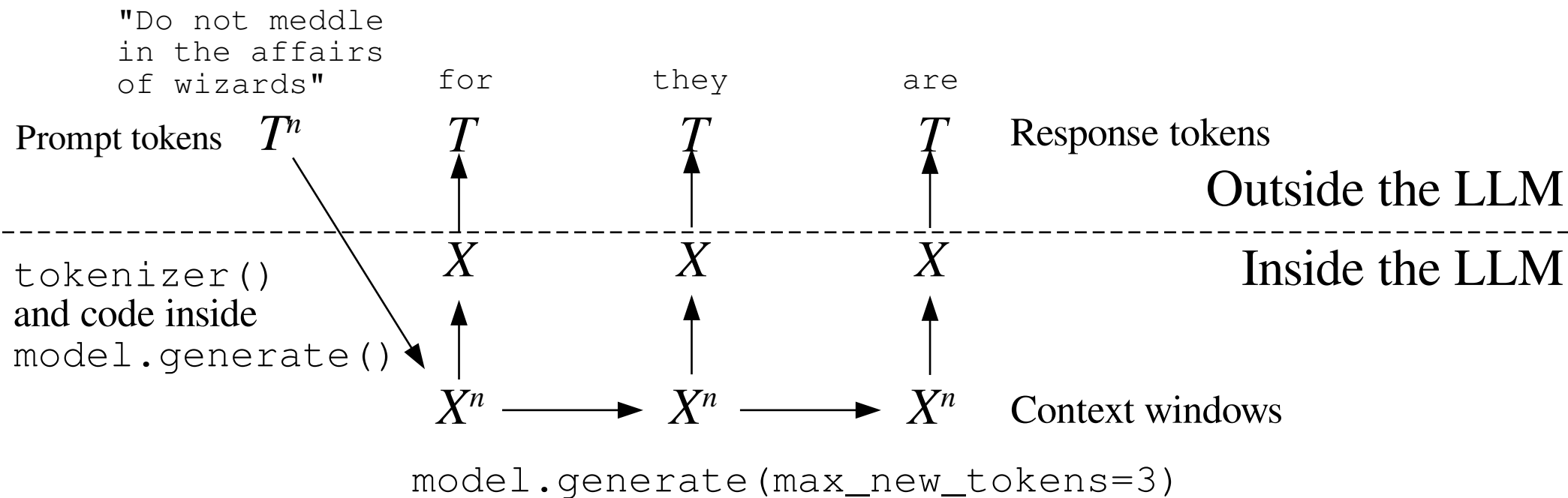
$X : \mathbb{R}^d$ latent space



Generation of text is iteration

T : set of tokens

$X : \mathbb{R}^d$ latent space



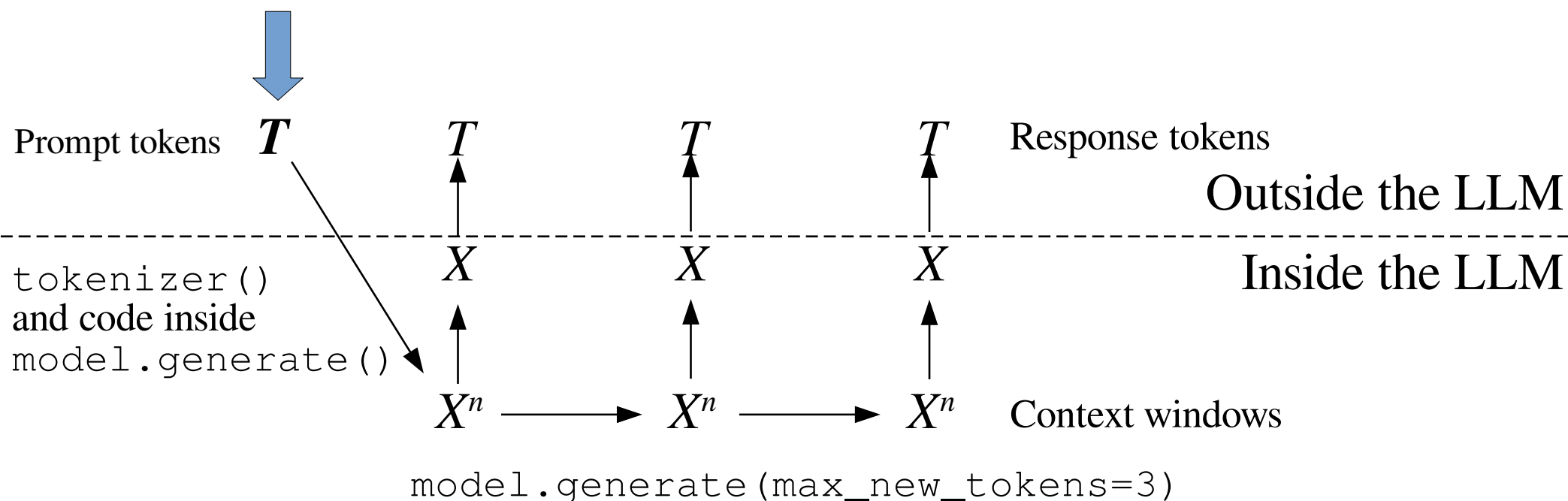
Intermission... single token prompts



Single tokens can be prompts

T : set of tokens

$X : \mathbb{R}^d$ latent space



Responses to single tokens

Token	Response
<unk>	1.1k views\nConsider the following statements: \n1. The complement of every Turing decidable language is Turing decid
<0xC5>	15:10, 11 September 2008 (UTC)\n### 2008
<0xC6>	100000000000000000000000000000000
ot	ally, the same as the one in the previous section.\n\\begin{figure}[htbp]\n\\centering\n\\includegraphics[width
}	(string consisting entirely of spaces)



Model: EleutherAI/Llemma7B

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Michael Robinson

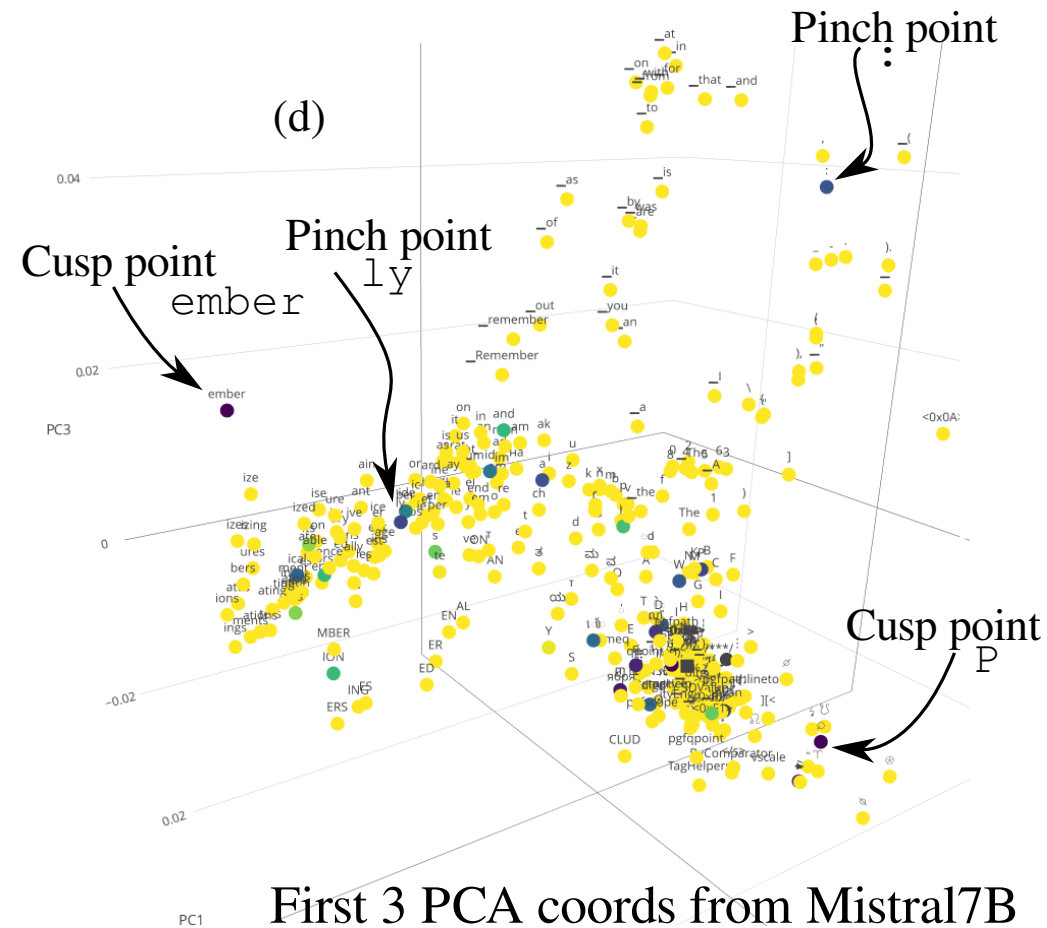
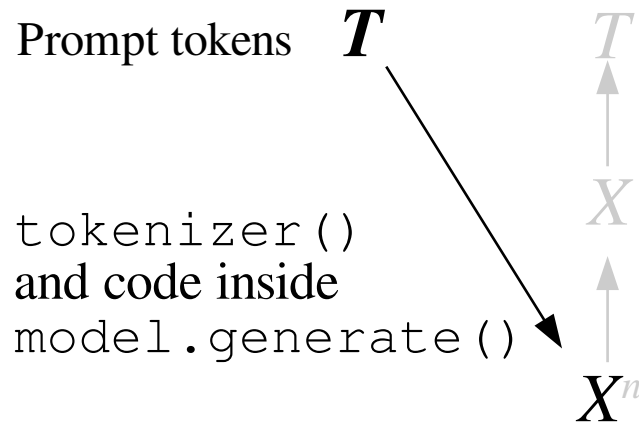
Reconstructing token subspace

Part 1: Whitney embedding



Bounding manifolds...

- The image of the token embedding is not* a manifold...

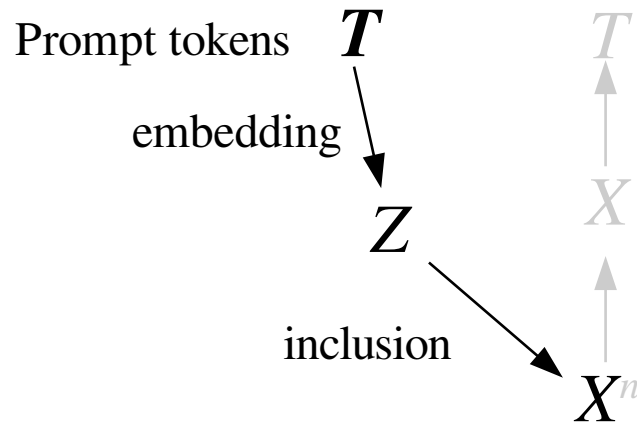


*Robinson, Dey, Chiang, <https://arxiv.org/abs/2504.01002> NeurIPS 2025

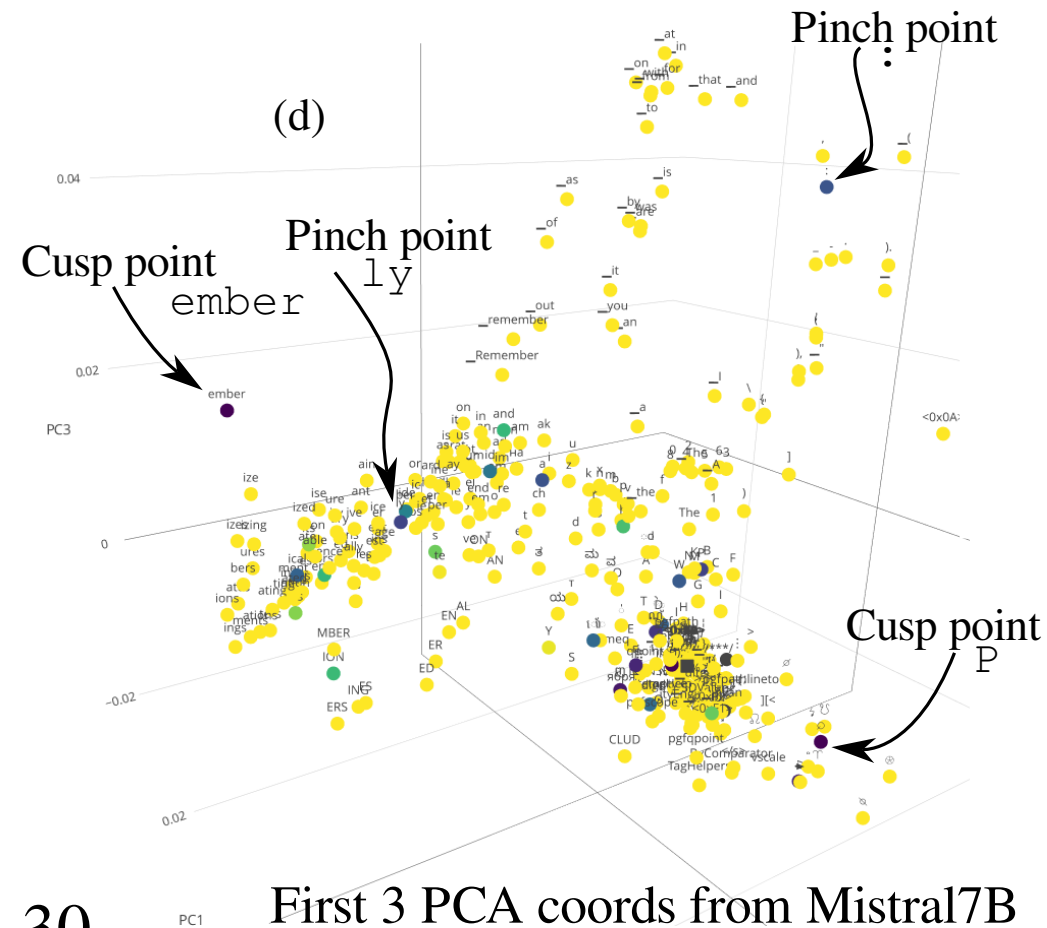
Bounding manifolds...

- The image of the token embedding is not* a manifold...

... but it lies within a bounding manifold Z



$\dim Z$ is typically around 30

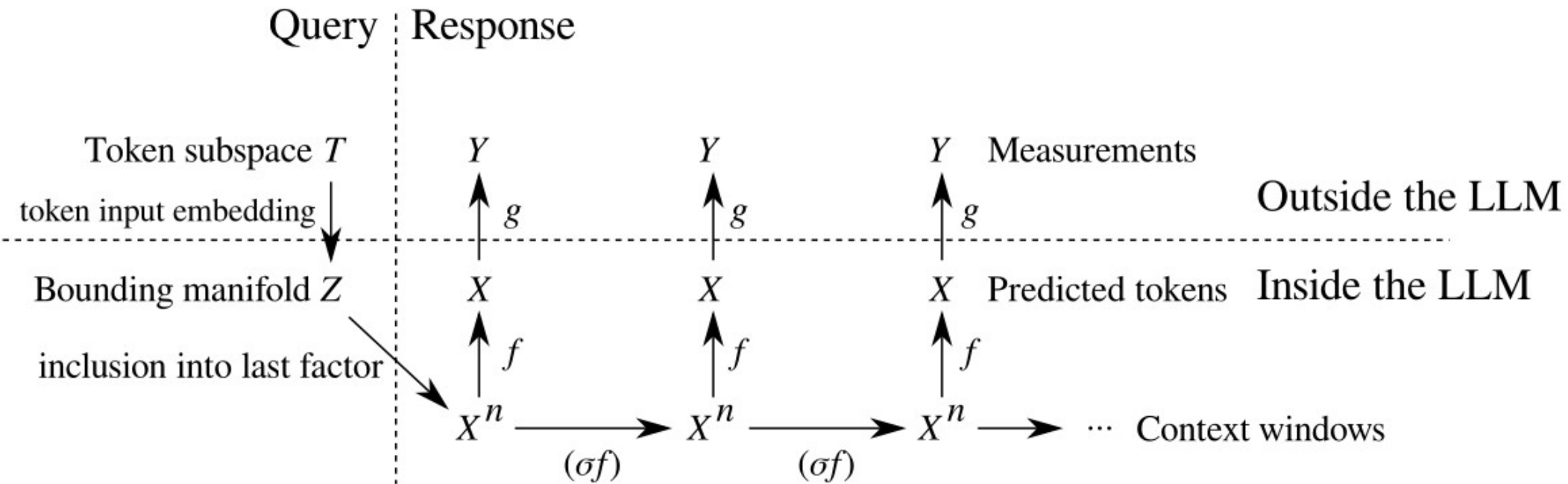


First 3 PCA coords from Mistral7B

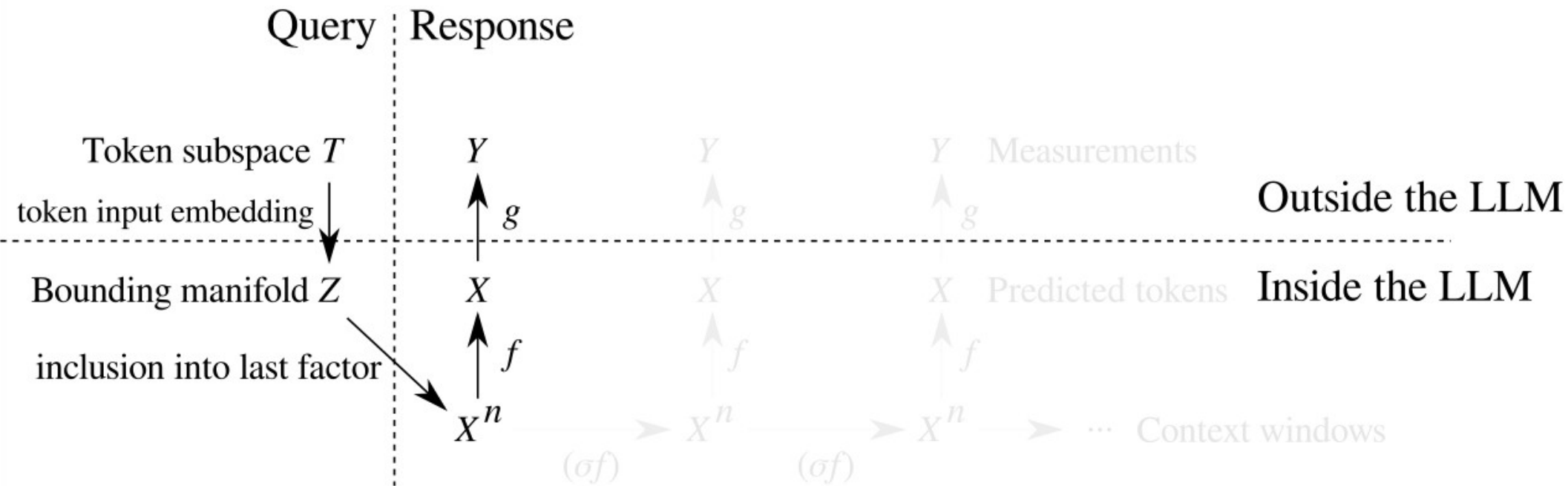
*Robinson, Dey, Chiang, <https://arxiv.org/abs/2504.01002> NeurIPS 2025



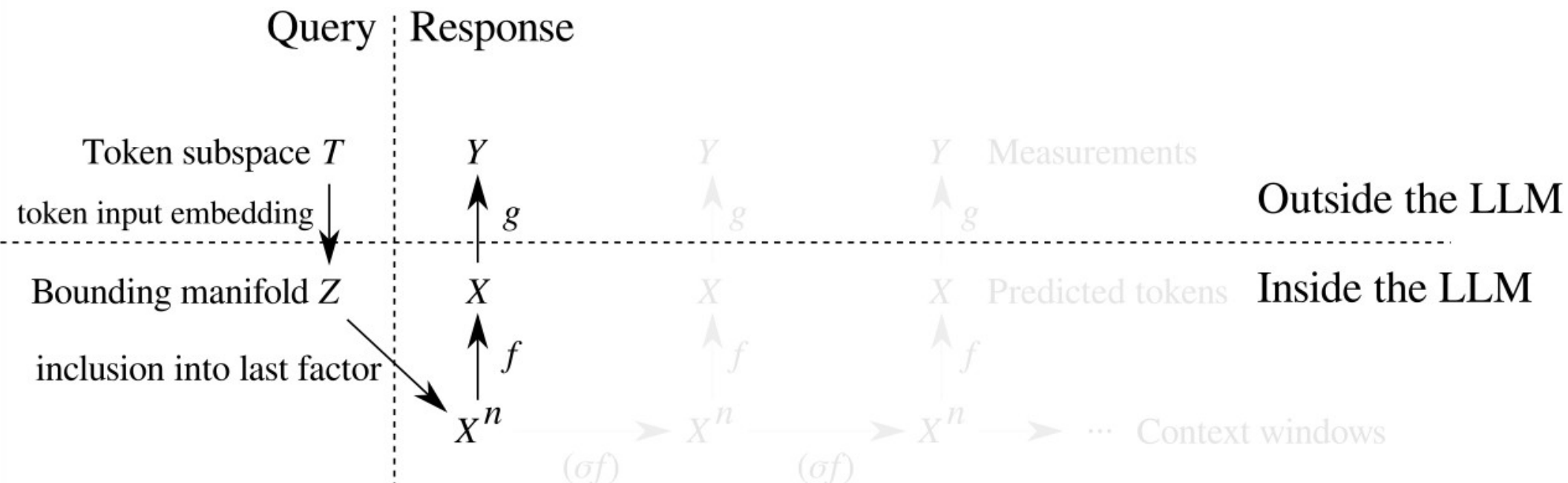
Data pipeline



Data pipeline: just one response token



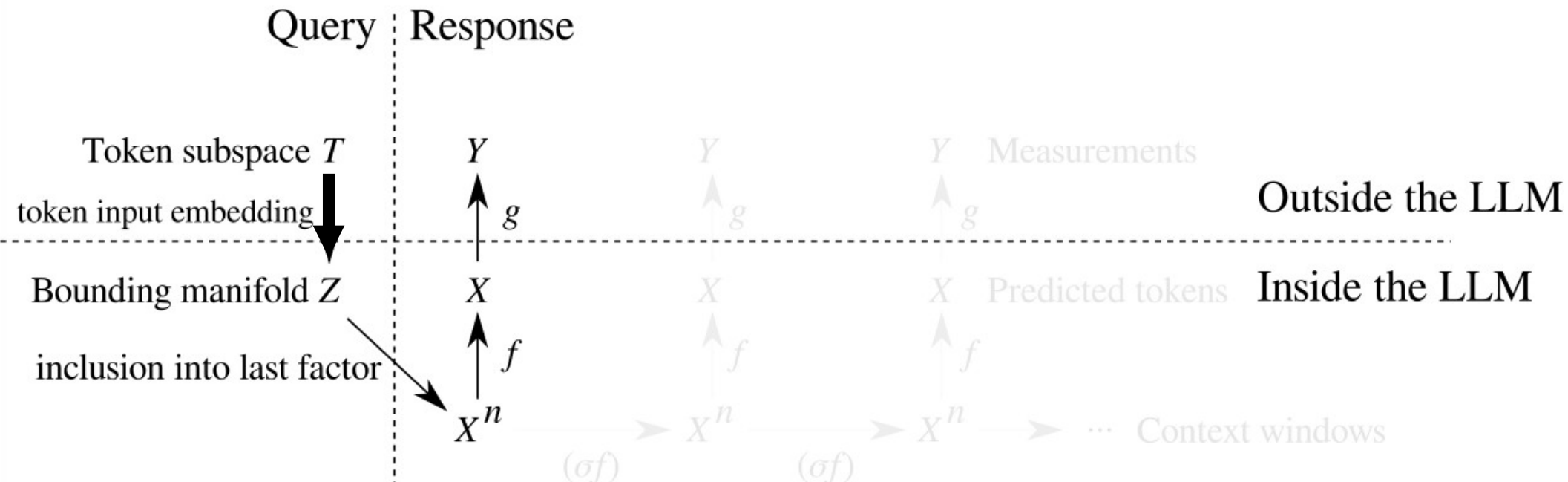
If we have an “open weights model”



```
from transformers import AutoCausalModelForLM, AutoTokenizer
tokenizer = AutoTokenizer.from_pretrained("yourtokenizer")
model = AutoCausalModelForLM.from_pretrained("yourmodelhere")
inputs = tokenizer("a", return_tensors="pt")
outputs = model.generate(inputs, max_new_tokens=1,
                        return_dict_in_generate=True, output_scores=True)
probs = outputs.scores[0].softmax(-1)
```



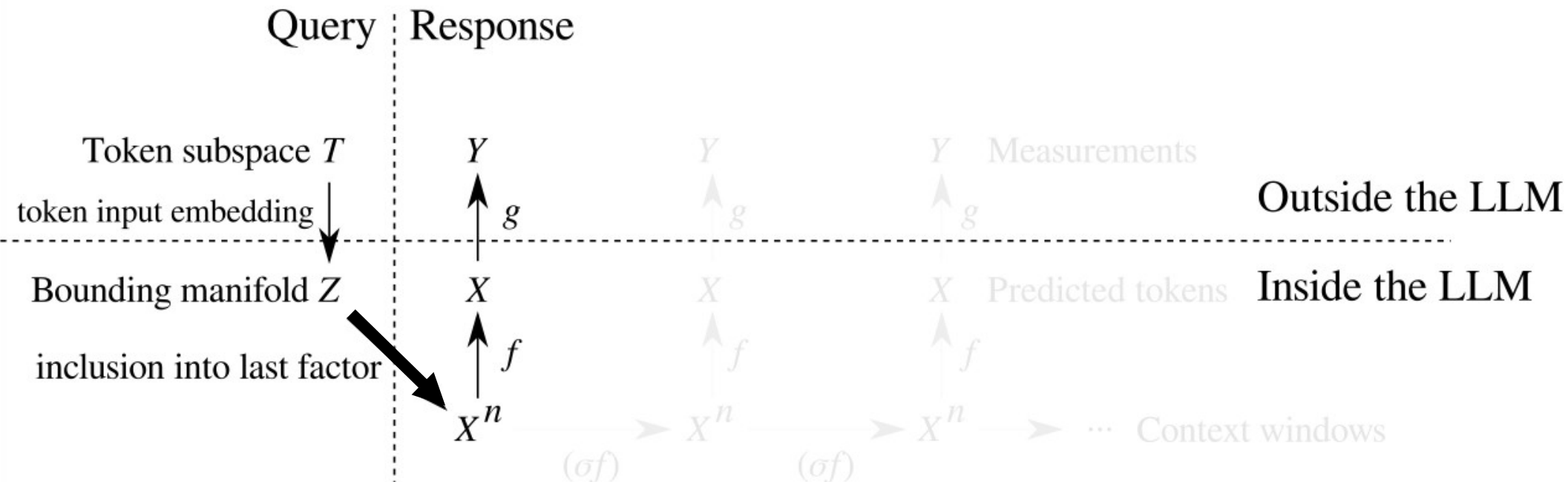
Text to tokens...



```
from transformers import AutoCausalModelForLM, AutoTokenizer
tokenizer = AutoTokenizer.from_pretrained("yourtokenizer")
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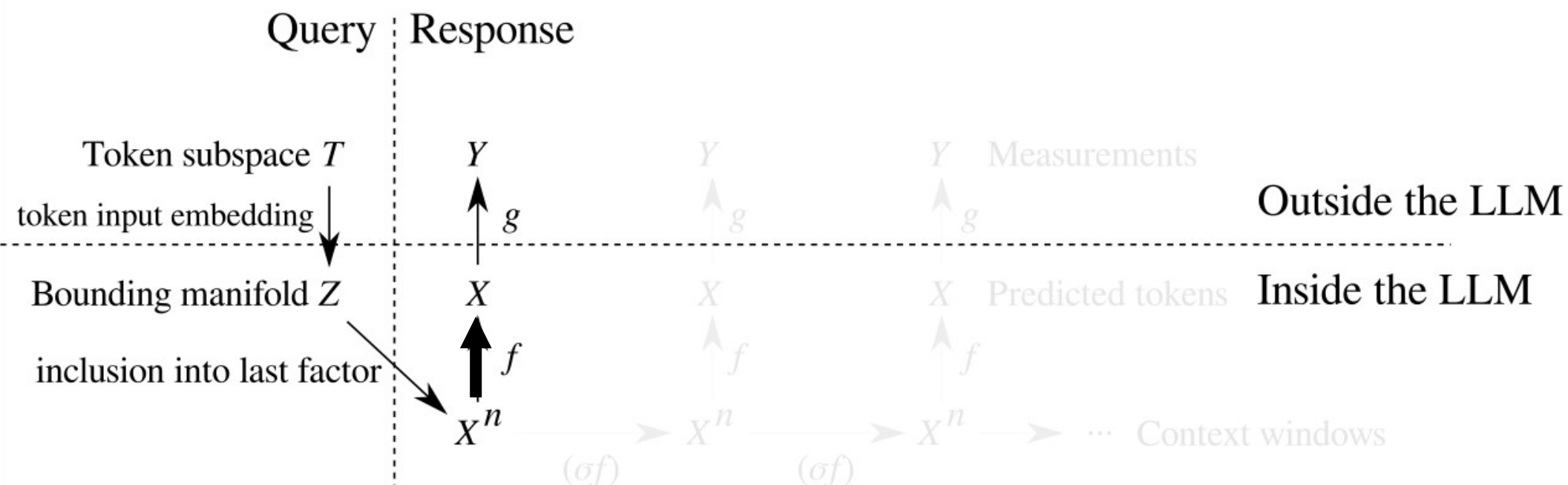
Tokens to latent space



```
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tokenizer = AutoTokenizer.from_pretrained("yourtokenizer")
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probs = outputs.scores[0].softmax(-1)
```



Generate next token distribution

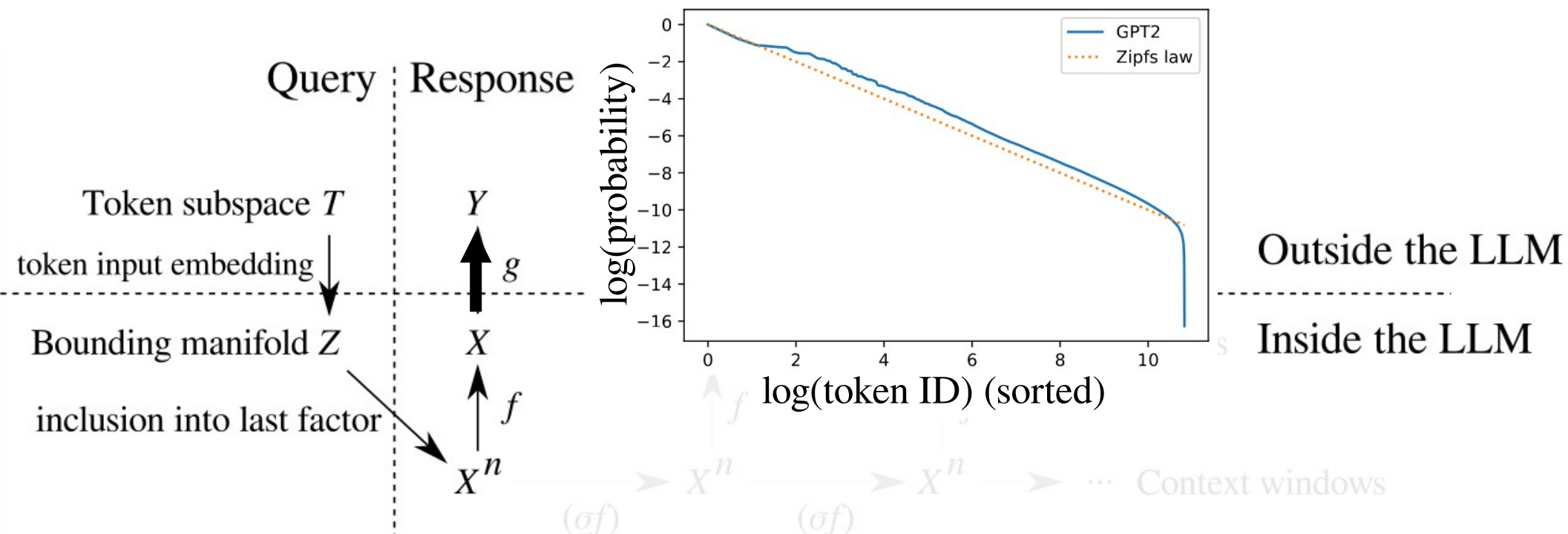


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                        return_dict_in_generate=True, output_scores=True)
probs = outputs.scores[0].softmax(-1)
```



Final outputs are generation probabilities*

*which may not reflect the training data distribution! [arXiv:2401.17377](https://arxiv.org/abs/2401.17377)

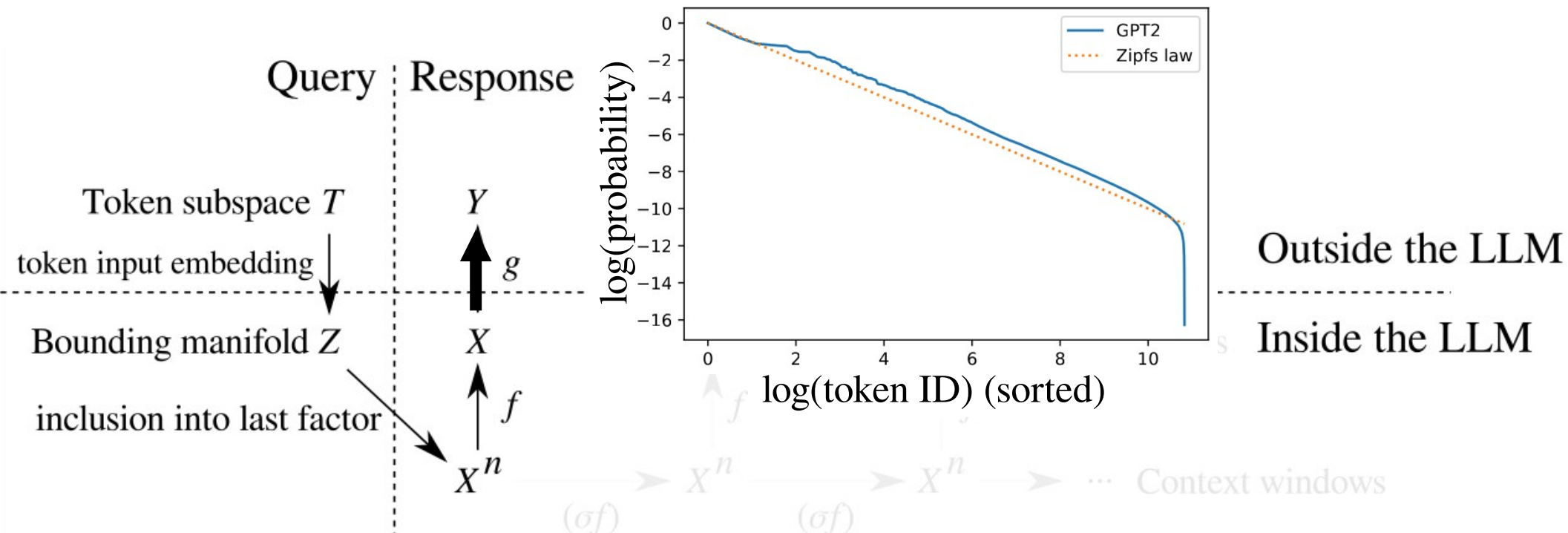


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probs = outputs.scores[0].softmax(-1)
```



Apply Whitney embedding theorem

$2 \dim X = 8192 \leq \dim Y = 32016 \dots g$ is an embedding

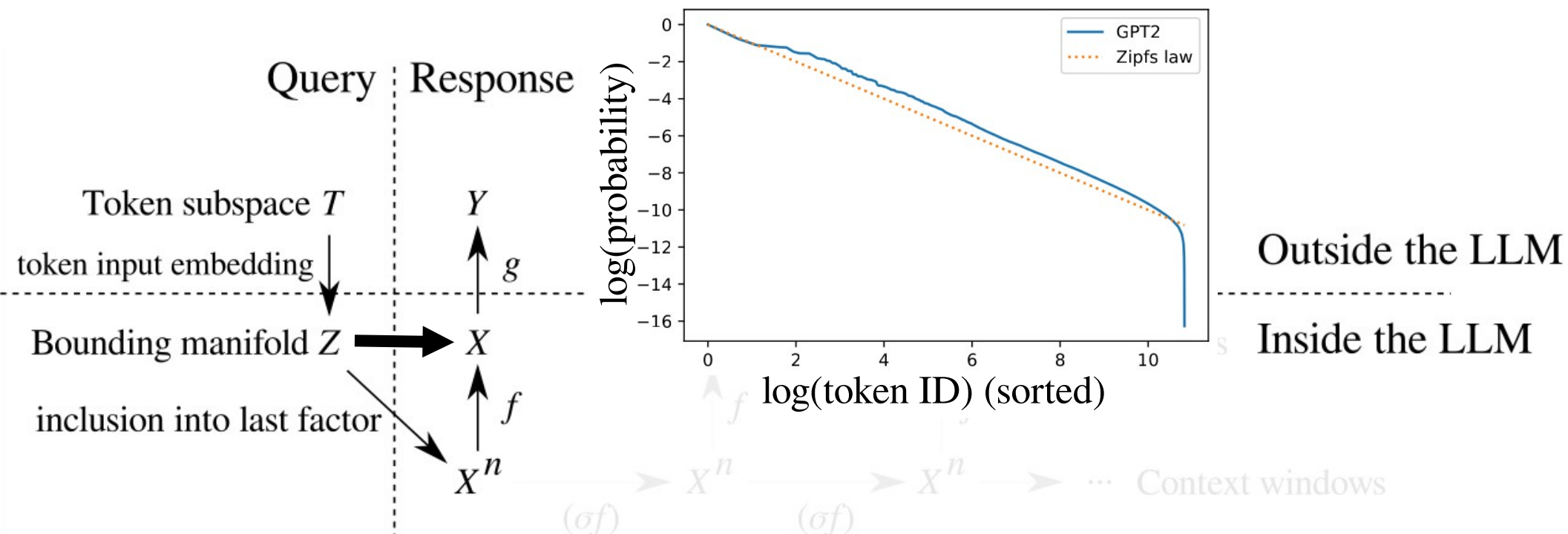


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probs = outputs.scores[0].softmax(-1)
```



Apply Whitney again

$2 \dim Z = 58 \leq 4096 = \dim X \dots$ another embedding!

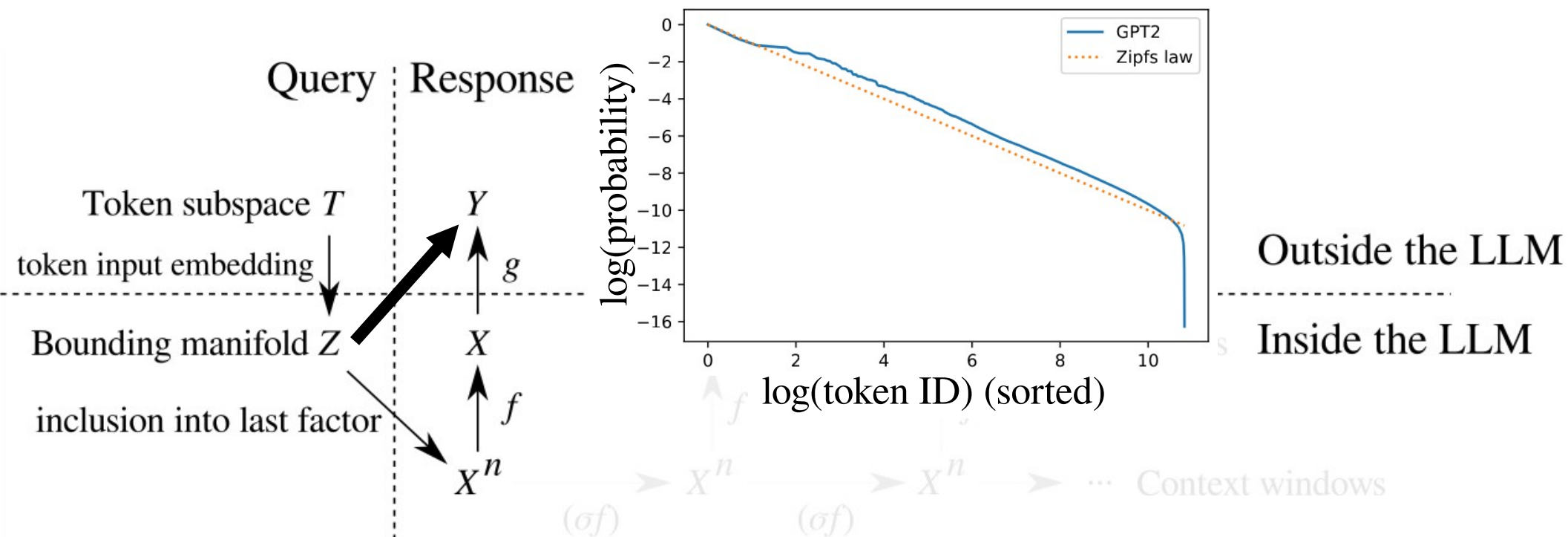


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model = AutoCausalModelForLM.from_pretrained("yourmodelhere")
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outputs = model.generate(inputs, max_new_tokens=1,
                        return_dict_in_generate=True, output_scores=True)
probs = outputs.scores[0].softmax(-1)
```



Apply Whitney again

$2 \dim Z = 58 \leq 4096 = \dim X \dots$ another embedding!



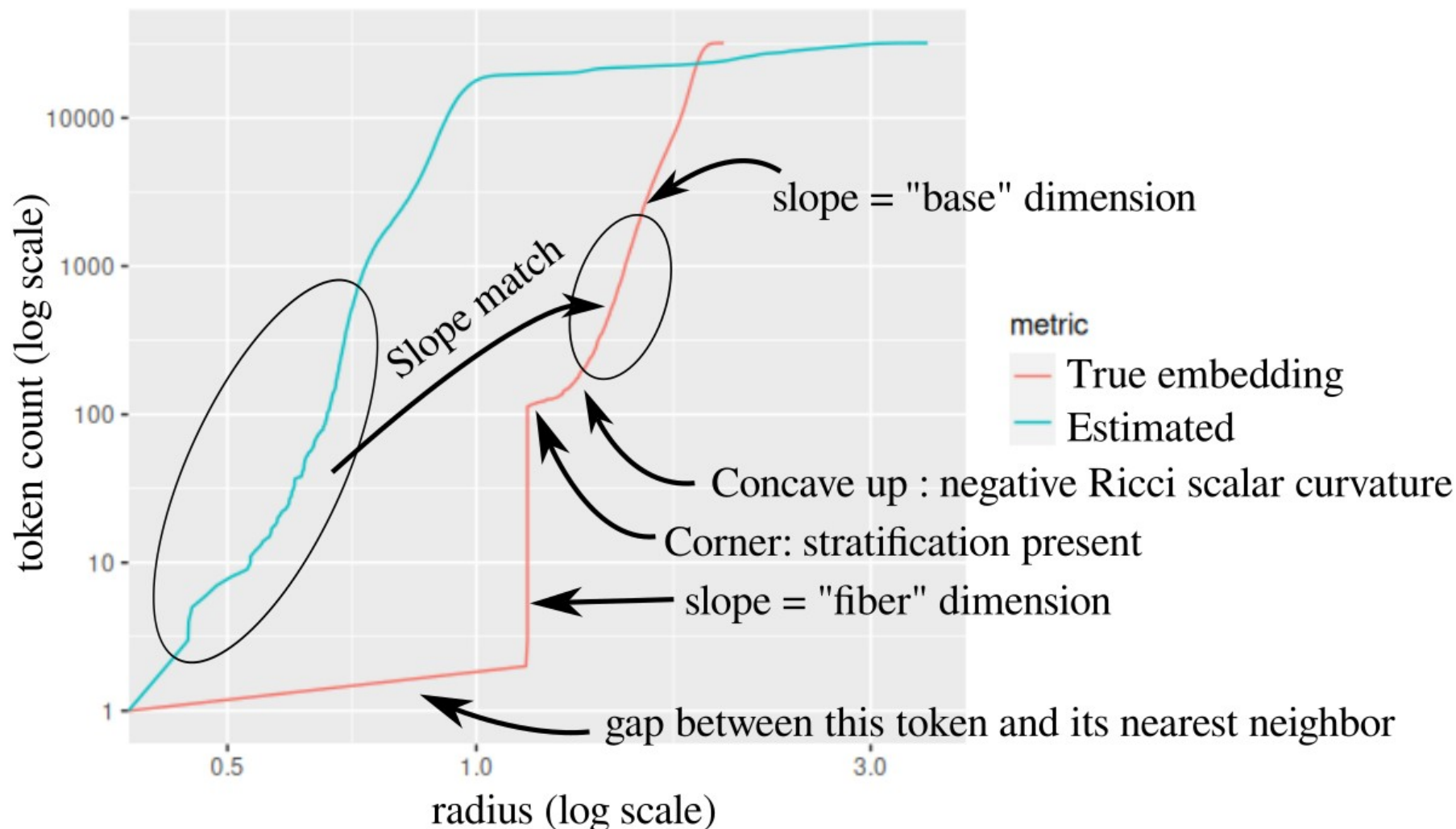
Procedure:

For each token, just use the next token distribution **as** its coordinates.
This recovers original coordinates up to diffeomorphism



Test results: dimension* recovered

- Dimension as a proxy for homeomorphism...



*Some tokens don't have well-defined dimension

Michael Robinson

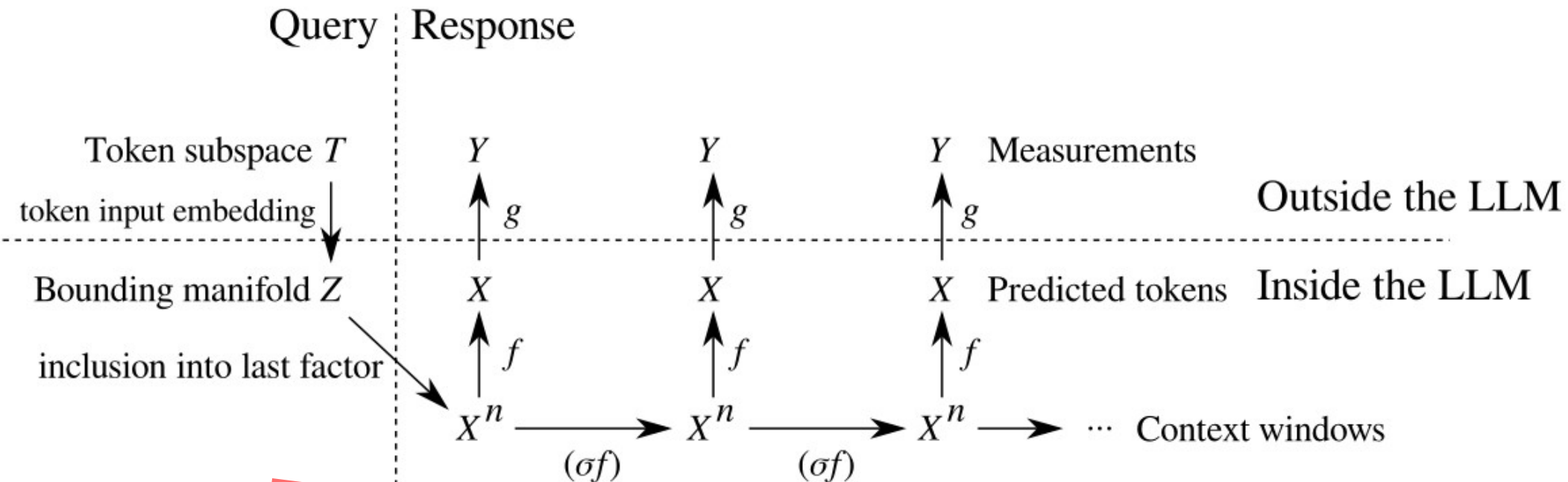
Reconstructing token subspace

Part 2: Partial views of sliding windows



Data pipeline

- Motivation: no direct access to the LLM “insides”

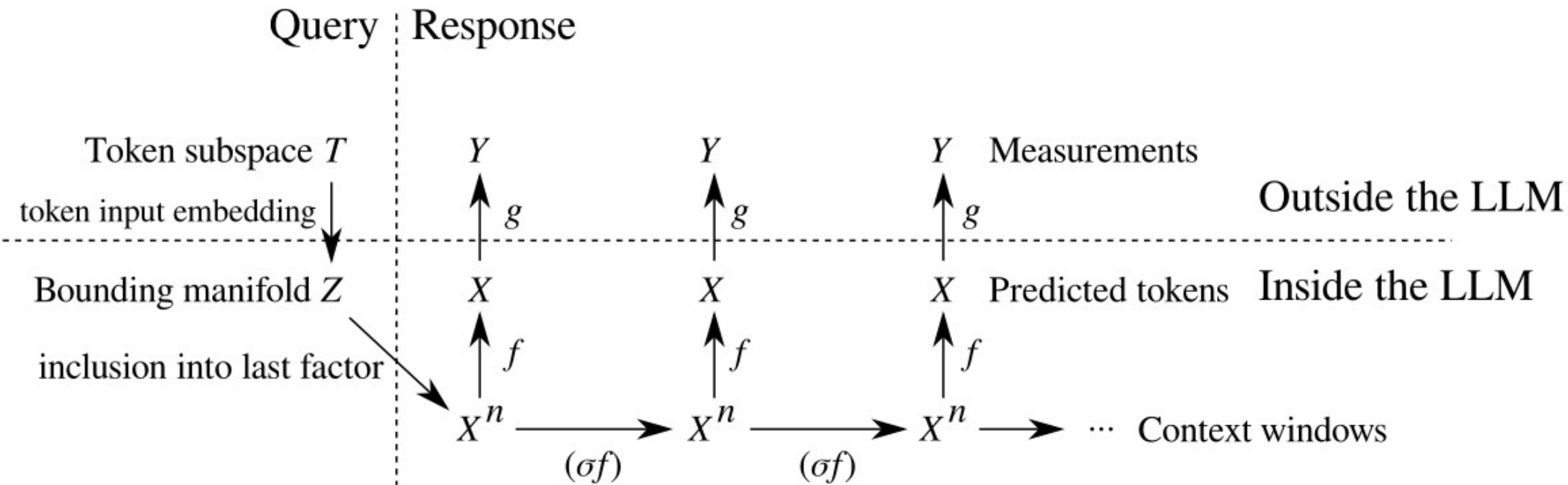


```
from transformers import AutoCausalModelForLM, AutoTokenizer
tokenizer = AutoTokenizer.from_pretrained("yourtokenizer")
model = AutoCausalModelForLM.from_pretrained("yourmodelhere")
```



Data pipeline

- Motivation: no direct access to the LLM “insides”



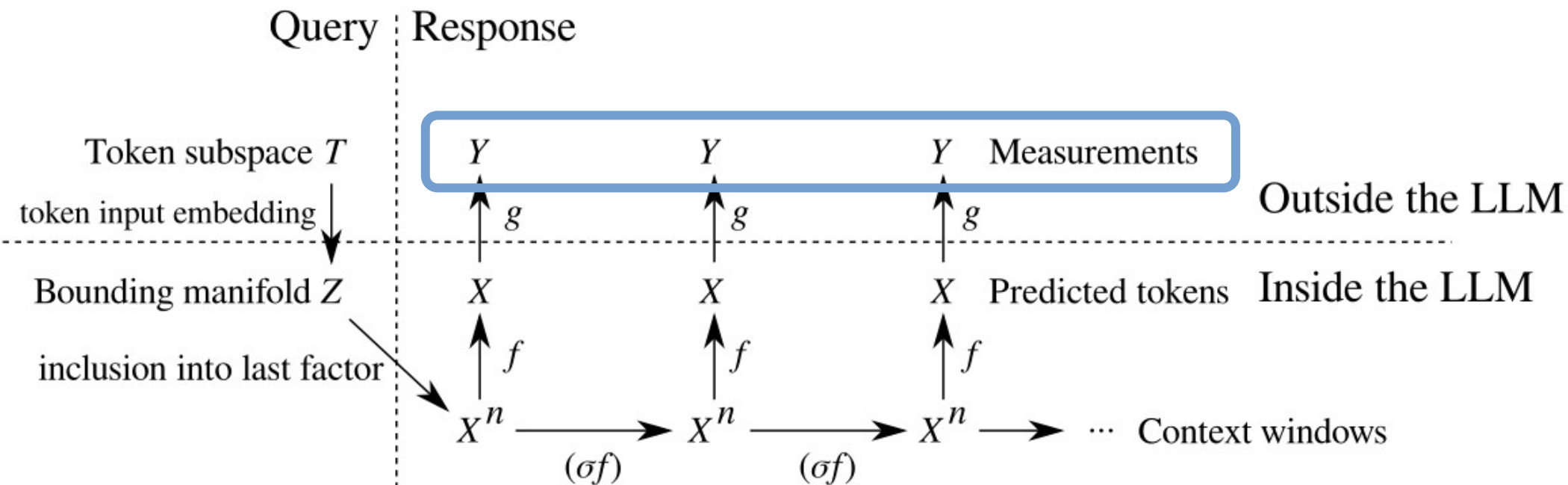
```
import ollama
response=ollama.generate(model="yourmodel",
                        prompt="a",
                        options={'num_predict' : m})
```

NB: Yes, ollama is for open source models, but proprietary APIs look similar, as does `transformers.pipeline()`.



Data pipeline

- Instead: Limited measurement taken from response



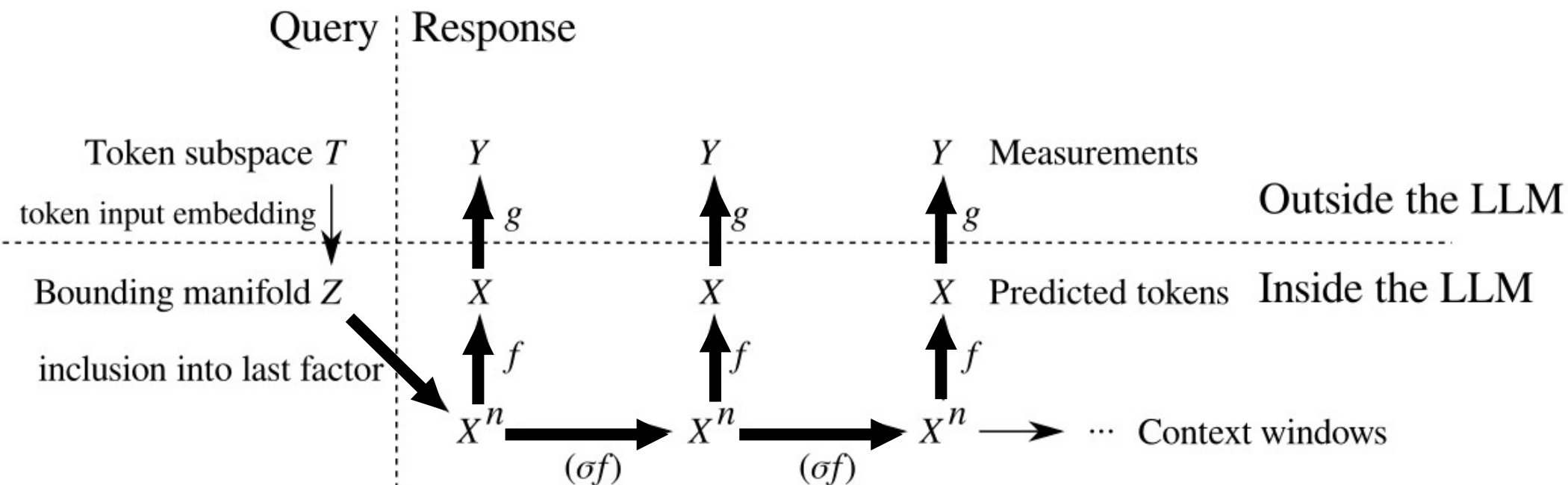
```
import ollama
response=ollama.generate(model="yourmodel",
                        prompt="a",
                        options={'num_predict' : m})
```

NB: Yes, ollama is for open source models, but proprietary APIs look similar, as does `transformers.pipeline()`.



Main theorem

- Theorem: $Z \rightarrow Y^m$ is a generically an embedding if m is large enough



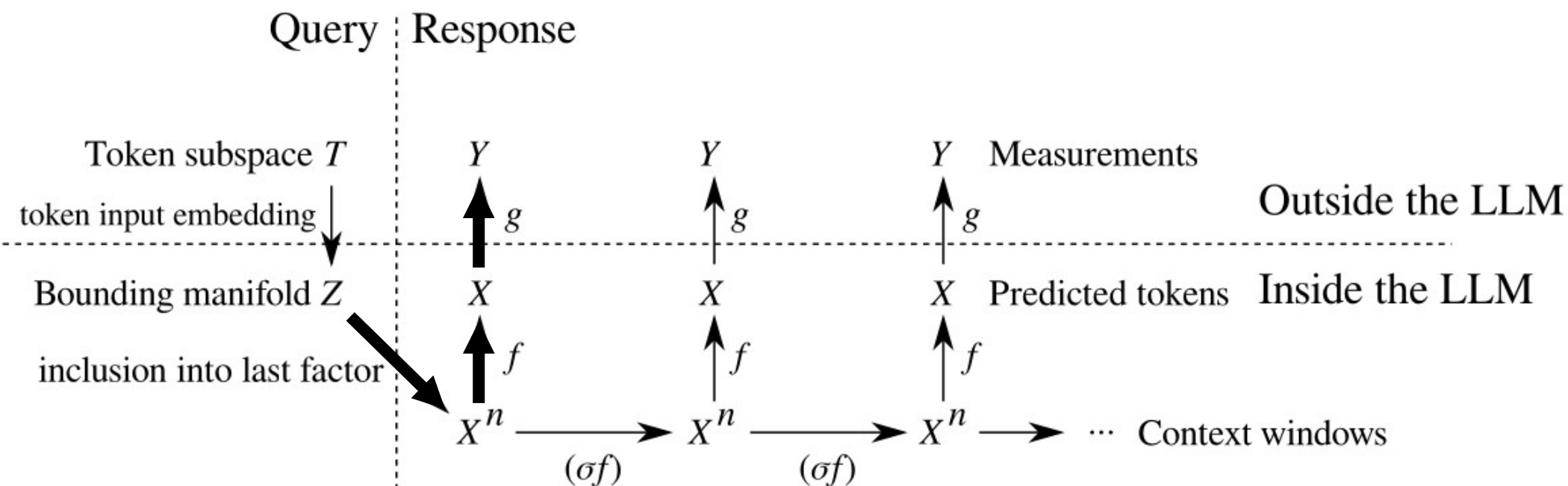
```
import ollama
response=ollama.generate(model="yourmodel",
                        prompt="a",
                        options={'num_predict' : m})
```

NB: Yes, `ollama` is for open source models, but proprietary APIs look similar, as does `transformers.pipeline()`.



Intersection submanifold

- Multiple preimages for a measurement y_1 in Y

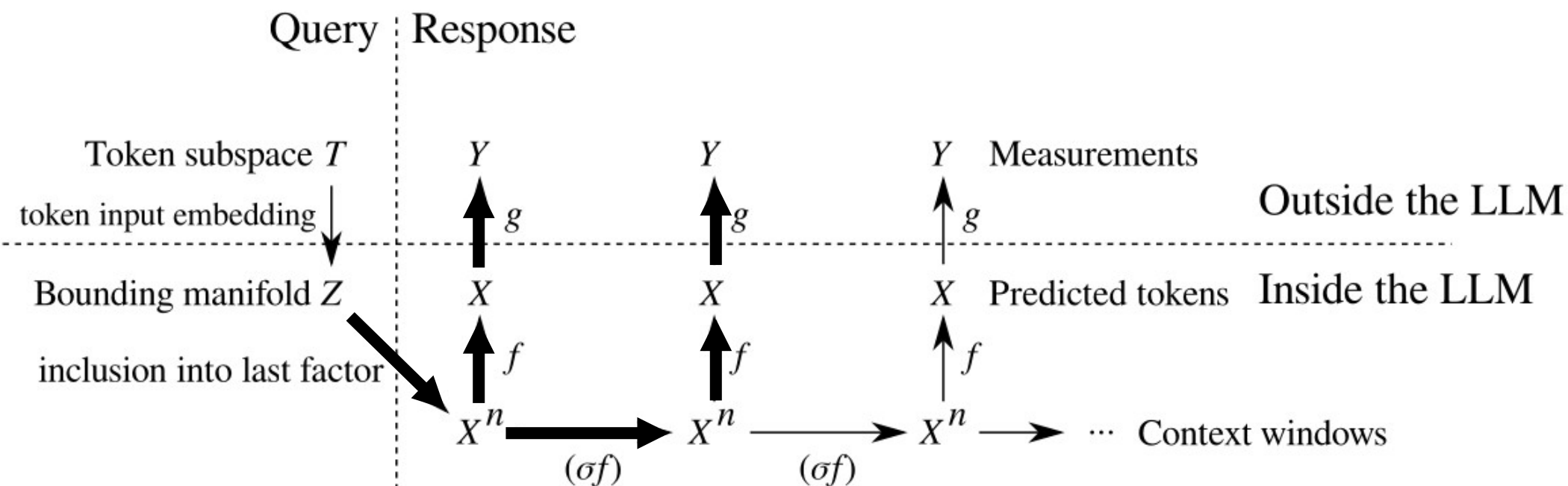


Generically $(f^{-1} \circ g^{-1})(y_1) \subseteq Z$ is a submanifold of positive codimension



Intersection submanifold

- Fewer preimages for a sequence of measurements

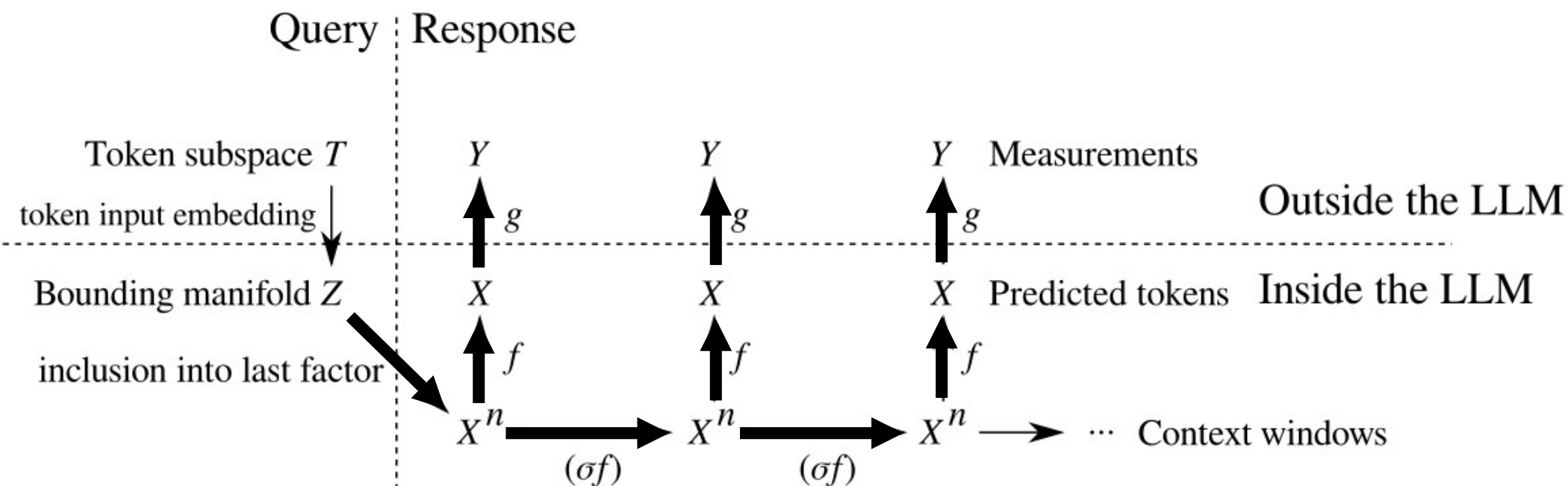


Generically $(f^{-1} \circ g^{-1})(y_1) \subseteq Z$ is a submanifold of positive codimension and $((\sigma f)^{-1} \circ f^{-1} \circ g^{-1})(y_2) \subseteq Z$ is a submanifold of positive codimension



Intersection submanifold

- Multi-jet transversality says, “intersect enough of these and you’ll end up with an empty set!”



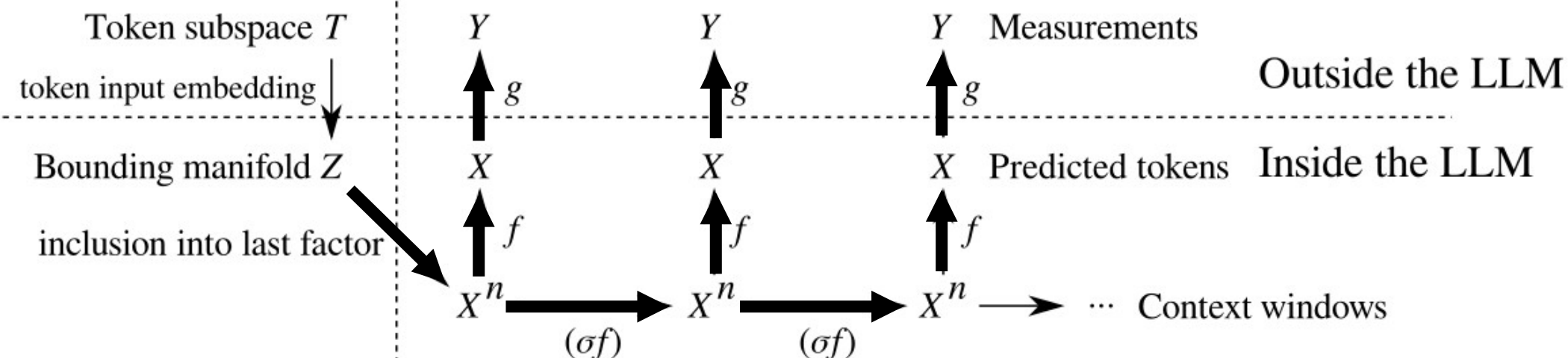
Generically $(f^{-1} \circ g^{-1})(y_1) \subseteq Z$ is a submanifold of positive codimension
 and $((\sigma f)^{-1} \circ f^{-1} \circ g^{-1})(y_2) \subseteq Z$ is a submanifold of positive codimension
 and $((\sigma f)^{-1} \circ (\sigma f)^{-1} \circ f^{-1} \circ g^{-1})(y_3) \subseteq Z$ is a submanifold of positive codimension



Intersection submanifold

- Multi-jet transversality says, “intersect enough of these and you’ll end up with an empty set!”

Query Response $Z \rightarrow Y^m$ is an embedding for large enough m

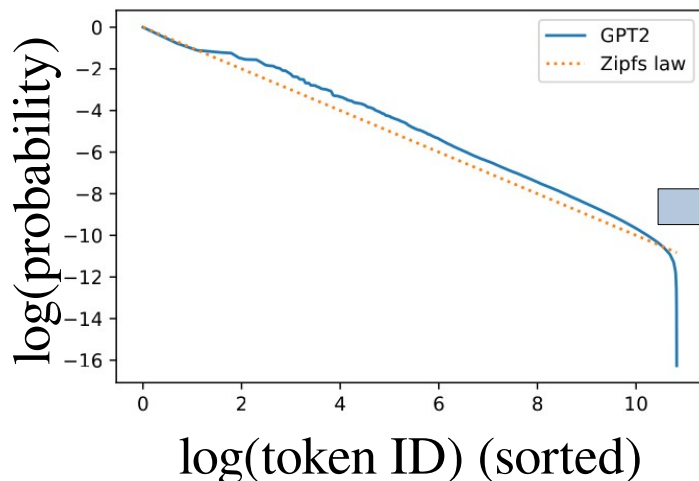


Generically $(f^{-1} \circ g^{-1})(y_1) \subseteq Z$ is a submanifold of positive codimension
 and $((\sigma f)^{-1} \circ f^{-1} \circ g^{-1})(y_2) \subseteq Z$ is a submanifold of positive codimension
 and $((\sigma f)^{-1} \circ (\sigma f)^{-1} \circ f^{-1} \circ g^{-1})(y_3) \subseteq Z$ is a submanifold of positive codimension



Practical considerations

- “Measurements” may be slow to converge...



Typically power-law;
Many infrequent tokens
May not actually reflect the
actual next token distribution

- ... but we have more options for collection

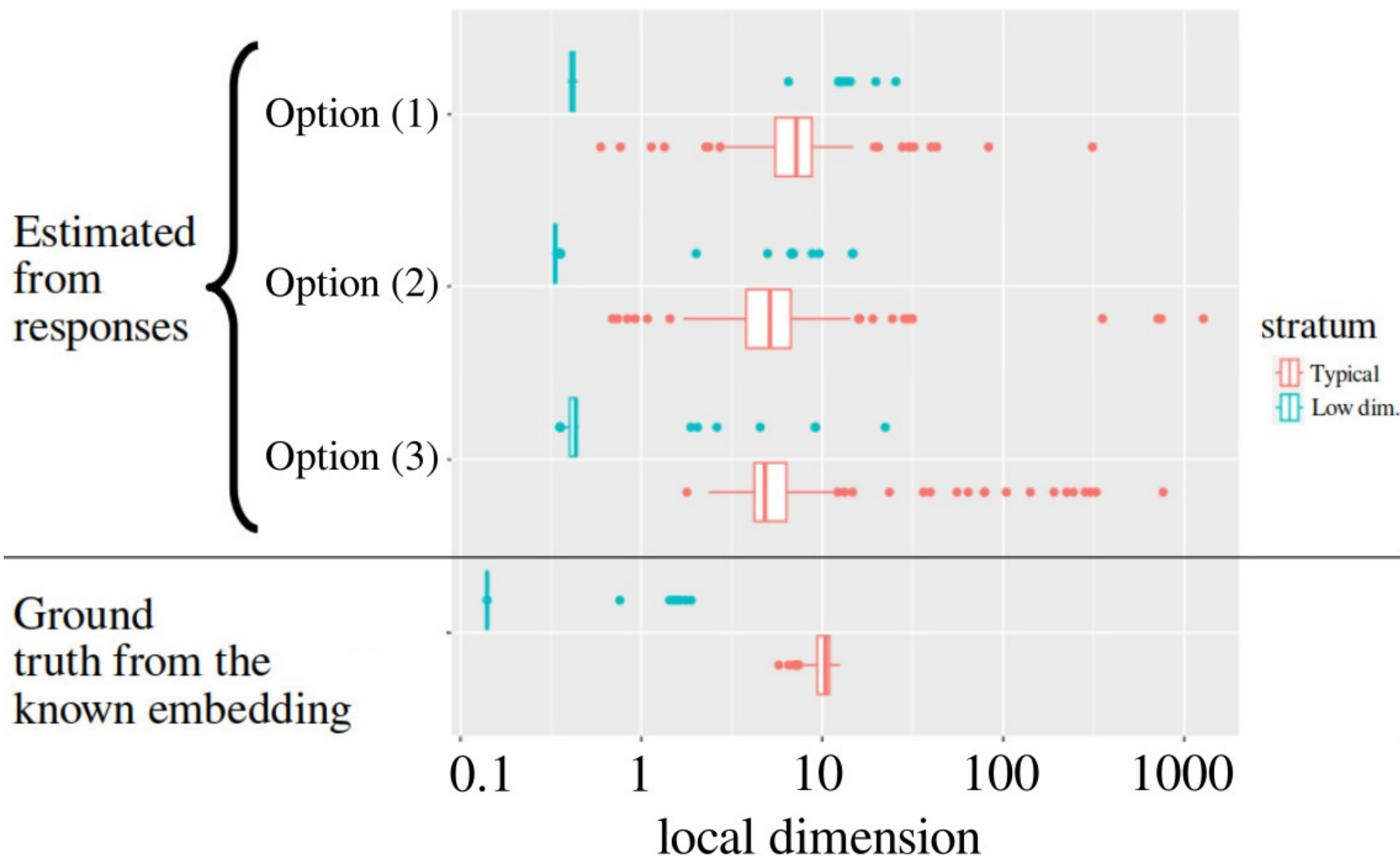
Option (1): Collect $m = 30$ response tokens and $\ell = 3$ probabilities for the top three tokens at each response token position (ignoring what the tokens actually were),

Option (2): Collect $m = 30$ response tokens and $\ell = 32016$ probabilities, one for each token, but aggregated over the entire response, and

Option (3): Collect $m = 1$ response token and $\ell = 32016$ probabilities, one for each token being the first token in the response.



Results: dimension is preserved



Results: geometry is destroyed

- The distribution of Ricci scalar curvature changes

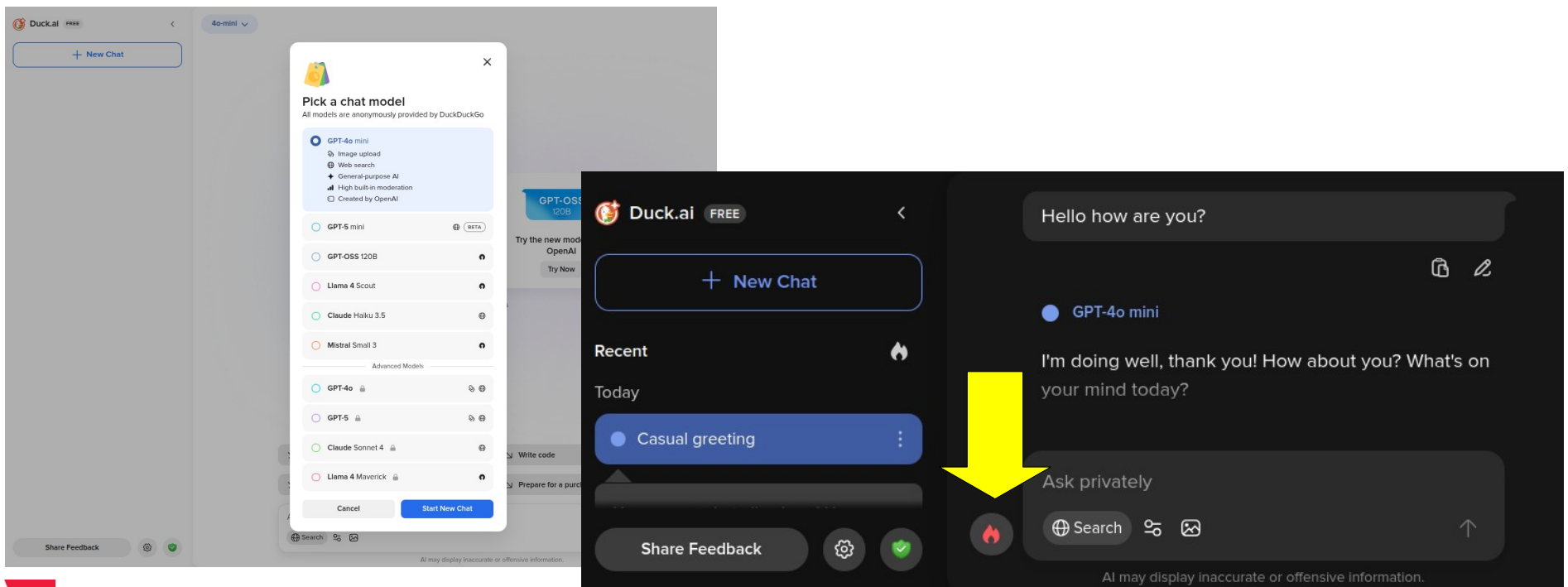
Source	Q1	Q2	Q3
Original token embedding ([2, Tab. 2])	-185	-169	-153
Estimated from Algorithm 1	-1403	-661	-165

- This is expected... the embedding coordinates have nothing to do with the transformer f at all!
- Caution: if you thought distances in the token embedding space were meaningful, they are not preserved...



If you want to try this...

- You can't use the web interface for ChatGPT, because you don't control the context window
- Instead, try <https://duck.ai> since the context is controlled by you. Delete it after every prompt



Implications and next steps

- That topology can be extracted (expensively) even if the model is proprietary
- Topology of the internal representation of tokens in an LLM directly impacts its behavior
 - If the token subspace is not a manifold, gradient descent is not well defined!
 - Prompt engineering is, as a result, an artisan craft!
- What about geometry? That's next up...
- Details: <https://doi.org/10.3390/math13203320>
- Questions? Ask! michaelr@american.edu

