458 A Appendix

459 A.1 Supplemental Results

Fig. 6 illustrates model predictions across every Number Game concept in [33].



Figure 6: Model predictions across every Number Game concept in [33]

460

Recall that we deduplicated the proposals instead of performing actual importance sampling. Fig. 7
contrasts model fit for importance sampling and deduplication. We originally did deduplication
simply because importance sampling is not possible with GPT-4, and GPT-4 proved necessary for
the logical concepts. On number concepts we used code-davinci-002, from which we can construct
an importance sampler because it exposes the log probability of its samples. On number concepts
deduplication provides a fit that is on-par (actually slightly better) compared to importance sampling (Fig. 7).



Figure 7: Monte Carlo inference using deduplication instead of importance sampling does not harm model fit to human data. The above figures show Number Game models using a learned prior and 100 samples, and show predictions only on holdout data.

467

468 A.2 Human Study

⁴⁶⁹ 16 participants were recruited primarily through a Slack message sent to a channel populated by ⁴⁷⁰ members of our academic department. Participants had an average age 28.1 (stddev 13.6, all over ⁴⁷¹ 18), and were 7 male/5 female/1 nonbinary/3 declined to answer. Participants were randomly split ⁴⁷² between the concepts of *most common color / least common color*. Each participant went through 15 ⁴⁷³ trials, and took an average of 294s to complete those 15 trials. In exchange for participating in the ⁴⁷⁴ study, participants received \$10 in Amazon gift cards. Fig. 8 illustrates the web interface shown to ⁴⁷⁵ our human participants, including the cover story.

476 A.3 Modeling

477 A.3.1 Temperature and Platt Transform

Adding a temperature parameter T to a model corresponds to computing the posterior via

$$p_{\text{Temp}}(X_{\text{test}} \in C | X_{1:K}) \approx \sum_{C \in \{C^{(1)}, \dots, C^{(S)}\}} w^{(C)} \mathbb{1} [X_{\text{test}} \in C], \text{ where}$$

$$w^{(C)} = \frac{\left(\tilde{w}^{(C)}\right)^{1/T}}{\sum_{C'} \left(\tilde{w}^{(C')}\right)^{1/T}} \text{ and } \tilde{w}^{(C)} = p(C)p(X_{1:K}|C)\mathbb{1} [C \in \{C^{(1)}, \dots, C^{(S)}\}]$$
(10)

Adjusting the predictions of a model using a Platt transform corresponds to introducing parameters aand b which transform the predictions according to

$$p_{\text{Platt}}(X_{\text{test}} \in C | X_{1:K}) = \text{Logistic}(b + a \times \text{Logistic}^{-1}(p(X_{\text{test}} \in C | X_{1:K})))$$
(11)

For the number game, every model has its outputs transformed by a learned Platt transform. This is because we are modeling human ratings instead of human responses. We expect that the ratings correspond to some monotonic transformation of the human's subjective probability estimates, and so this transformation gives some extra flexibility by inferring the correspondence between probabilities and ratings. Logical concept models do not use Platt transforms.

Trial 1: Please read these instructions carefully

You are going to attempt to learn the meaning of a new word in an alien language, which the aliens call "Wudsy." On each trial, you are going to see a collection of shapes at the bottom of the webpage, and your job is to select which ones you think are "Wudsy." Afterward, the aliens tell you which shapes are "Wudsy."

The meaning of the word Wudsy is the same during the whole experiment. However, it is possible that whether something is Wudsy depends on what other shapes it is in the context of. Wudsy may or may not correspond to an English word.

To start with, no one has given you any examples of what counts as "Wudsy." So just do your best below and pick which ones you think might belong to the concept called "Wudsy." Right after you do so, the aliens are going to label the Wudsy objects by drawing a black box around them, and then you are going to get another round of guessing which objects are "Wudsy."

You will go through 15 trials of guessing what counts as Wudsy. Remember that the meaning of Wudsy does not change during the experiment, but it might depend on the other shapes in the collection.

(trial 1/15) Click yes on the objects that you think are Wudsy, and No on the other objects. Then click Next.



Figure 8: Cover story and web interface for our version of the logical concept learning study, which is based on [45]

486 A.3.2 Parameter fitting

⁴⁸⁷ Training consists of fitting the parameters T, θ (for the prior), ϵ (for the likelihood), α and β (for the ⁴⁸⁸ logical concepts likelihood), and Platt transform parameters a, b (for the Number Game). In practice, ⁴⁸⁹ this amounts to around 400 parameters, almost all of which come from θ .

We fit these parameters using Adam with a learning rate of 0.001. We perform 1000 epochs of training for the Number Game, and 100 epochs for logical concepts. There is a tenfold difference in the number of concepts, so this way they take about the same number of gradient steps.

For the number game we do 10-fold cross validation to calculate holdout predictions. For logical concepts we use the train-test split introduced in [45], which involves running different groups of human subjects on each concept twice, with different random examples. One sequence of random examples is arbitrarily designated as training data, and the other as holdout data.

⁴⁹⁷ All model were trained on a laptop using no GPUs. Training takes between a few minutes and an ⁴⁹⁸ hour, depending on the domain and the model.

Some of the parameters that we fit, namely ϵ , α , β , cannot be negative. To enforce this we actually optimize the inverse logistic of those parameters.

501 A.3.3 MCMC over Logical Expressions

Fleet was used¹ to perform MCMC over logical expressions with the domain-specific primitives in this file, which include:

true, false : boolean blue, yellow, green : object \rightarrow boolean rectangle, circle, triangle : object \rightarrow boolean small, medium, large : object \rightarrow boolean and, or, \iff , \implies : boolean \times boolean \rightarrow boolean $\forall, \exists : (shape \rightarrow boolean) \times 2^{object} \rightarrow boolean$ filter : (object \rightarrow boolean) $\times 2^{object} \rightarrow 2^{object}$ $\in : object \times 2^{object} \rightarrow boolean$ $\iota : 2^{object} \rightarrow object \cup \{\bot\}$, unique set element empty : $2^{object} \rightarrow boolean$ same_shape, same_color, same_size : object \times object \rightarrow boolean size<, size<, size>, size>, : object \times object \rightarrow boolean

The model first constructed a large hypothesis space by performing MCMC for 1 minute per batch, and per learning curve. In one minute, Fleet makes approximately 10⁶ MH proposals. There are a little more than 200 learning curves, and 25 batches per curve, for a total of about 5 billion MCMC

proposals. In the main text, we abbreviate this analysis by referring to 10^9 proposals.

The top 10 samples per batch and per learning curve were retained. These top 10 samples samples were then deduplicated to yield 45 thousand hypotheses. Parameter fitting and posterior estimation was performed solely over those 45 thousand hypotheses.

Quantitatively, these are vastly more proposals than the models introduced in this paper. Quantitatively, these proposals are also derived in a very different way: the hypothesis space for the BPL learner is actually informed by data on other learning curves, and also on the same learning curve, but in the future batches.

It is in this sense that the BPL model is a computational-level theory, and not a process model, because human subjects could not be proposing hypotheses using data that is going to be seen in the future, or on other learning curves. However, the above strategy for proposing hypotheses is a very reasonable heuristic for constructing the support of the space of plausible logical hypotheses that a human learner might ever think of.

520 A.4 Prompting

521 A.4.1 Proposing hypotheses

For the number game we use the following prompt for code-davinci-002 to generate candidate concepts in natural language, given examples $X_{1:K}$. The example number concepts given in the prompt come from the cover score given to human participants [33]:

```
# Pvthon 3
525
    #
     Here are a few example number concepts:
526
      -- The number is even
527
    #
     -- The number is between 30 and 45
    #
528
529
    #
     -- The number is a power of 3
    #
      -- The number is less than 10
530
531
    #
      Here are some random examples of numbers belonging to a different \varkappa
532
533
        └→ number concept:
   # X_{1:K}
534
```

¹Running the model was graciously performed by the authors of [45], who provided us with the raw data.

```
# The above are examples of the following number concept:
535
    # -- The number is
536
    where X_{1:K} is formatted by listing the numbers with a comma and a space between them.
537
    For the number game we used the following prompt to generate candidate concepts in python (code
538
    baseline):
539
540
    # Python 3
    # Here are a few example number concepts:
541
    # -- The number is even
542
    \# -- The number is between 30 and 45
543
    # -- The number is a power of 3
544
    # -- The number is less than 10
545
546
    #
    # Here are some random examples of numbers belonging to a different \varkappa
547
        └→ number concept:
548
549
    # X_{1:K}
    # Write a python function that returns true if 'num' belongs to \varkappa
550

    ↓ this number concept.

551
    def check_if_in_concept(num):
552
553
       return
    For logical concepts we used the following few-shot prompt for GPT-4 to generate candidate concepts:
554
    Here three simple concepts, which specify when an object is \varkappa'
555
        \searrow 'positive' relative to an example collection of other \swarrow
556
        \searrow objects. Before giving the rule for each concept, we give \varkappa
557
        arsigma examples of collections of objects, and which objects in the arsigma
558
559
        560
    Concept #1:
561
        An Example of Concept #1:
562
             POSITIVES: (big yellow rectangle)
563
             NEGATIVES: (big green circle), (medium yellow rectangle)
564
565
        Another Example of Concept #1:
             POSITIVES: (medium yellow rectangle)
566
             NEGATIVES: (big red circle), (small green circle)
567
568
    Rule for Concept #1: Something is positive if it is the biggest 2

    yellow object in the example.

569
570
571
    Concept #2:
572
573
        An Example of Concept #2:
             POSITIVES: (small yellow circle), (medium yellow rectangle)
574
             NEGATIVES: (big green circle), (big blue rectangle)
575
        Another Example of Concept #2:
576
             POSITIVES: (big blue circle), (medium blue rectangle)
577
             NEGATIVES: (small green circle), (medium yellow rectangle),
578
    Rule for Concept #2: Something is positive if there is another \varkappa
579
        └ object with the same color in the example.
580
581
    Concept #3:
582
        An Example of Concept #3:
583
             POSITIVES: (small yellow circle), (medium yellow rectangle)
584
585
             NEGATIVES: (big green circle), (big blue rectangle)
        Another Example of Concept #3:
586
             POSITIVES: (small blue circle), (small blue triangle), 🖌
587
                 \, (medium blue rectangle)
588
             NEGATIVES: (medium green triangle), (big yellow rectangle)
589
        Another Example of Concept #3:
590
             POSITIVES: (big red rectangle), (medium red rectangle), 🖌
591
592
                 └ (big red triangle)
             NEGATIVES: (medium green triangle), (big yellow rectangle)
593
594
    Rule for Concept #3: Something is positive if it is the same color \varkappa
        \backsim as the smallest triangle in the example.
595
```

```
596
    Now here are some examples of another concept called Concept #4, \varkappa
597
        \searrow but this time we don't know the rule. Infer ten different \swarrow
598
        \searrow possible rules, and make those ten rules as simple and \swarrow
599
         ५ general as you can. Your simple general rules might talk 🖌
600
        \searrow about shapes, colors, and sizes, and might make comparisons \swarrow
601
602
        \checkmark between these features within a single example, but it \swarrow
        \searrow doesn't have to. Remember that a rule should say when \swarrow
603
        4
          something is positive, and should mention the other objects \varkappa'
604
605
        6
          in the example, and should be consisting with what you see \checkmark
606
        Selow.
607
    Concept #4:
608
         X_{1:K}
609
    Rule for Concept #4: Something is positive if...
610
611
    Now make a numbered list of 10 possible rules for Concept #4. Start \varkappa
612
        \searrow by writing "1. Something is positive if". End each line with \swarrow
613
614
```

Each sample from the above prompt generates 10 possible concepts formatted as a numbered list. We draw 10 times at temperature=1 to construct 100 hypotheses. To obtain fewer than 100 hypotheses we take hypotheses from each sampled list in round-robin fashion. We found that asking it to generate a list of hypotheses generated greater diversity without sacrificing quality, compared to repeatedly sampling a single hypothesis.

The above prompt provides in-context examples of first-order rules. We also tried using a different prompt for propositional concepts that illustrated the examples as a truth table, and gave in-context example rules that were propositional:

```
Here are some example concepts defined by a logical rule:
623
624
    Rule: a triangle.
625
    Rule: a green rectangle.
626
    Rule: big or a rectangle (unless that rectangle is blue).
627
628
    Rule: not both big and green.
    Rule: either big or green, but not both.
629
    Rule: either a rectangle or not yellow.
630
    Rule: a circle.
631
632
633
    Now please produce a logical rule for a new concept. Your rule \swarrow
634
635
        ert should be consistent with the following examples. It must be arepsilon
          true of all the positive examples, and not true of all the \swarrow
636
          negative examples. The examples are organized into a table \checkmark
637
        \backsim with one column for each feature (size, color, shape):
638
639
640
    X_{1:K}
641
    Please produce a simple rule that is consistent with the above \vec{k}
642
        \backsim table. Make your rule as SHORT, SIMPLE, and GENERAL as \varkappa
643
644
        \searrow possible. Do NOT make it more complicated than it has to be, \swarrow
          or refer to features that you absolutely do not have to refer
645
        6
          to. Begin by writing "Rule: " and then the rule, followed by \swarrow
646
        6
        647
```

Using the first order prompt for every concept gives a $R^2 = .80$ fit to the human responses. Using both prompts gives the $R^2 = .81$ result in the main paper: the propositional prompt for the propositional problems, and the first order prompt for the higher order problems. We strongly suspect that a single prompt that just showed both propositional and higher-order in-context examples would work equally well, given that a single first-order prompt works about as well also, but we did not try that because of the high cost of using GPT-4. On the first batch, the learner has not observed any examples. Therefore the above prompts do not apply, and we use a different prompt to construct an initial hypothesis space:

```
Here are some example concepts defined by a logical rule:
656
657
   Rule: color is purple.
658
   Rule: shape is not a hexagon.
659
   Rule: color is purple and size is small.
660
   Rule: size is tiny or shape is square.
661
   Rule: size is not enormous.
662
   Rule: color is red.
663
664
   Please propose a some new concepts, defined by a logical rule. \checkmark
665
       └ These new concepts can only refer to the following features:
666
   - shape: triangle, rectangle, circle
667
   - color: green, blue, yellow
668
669
   - size: small, medium, large
670
   Please make your rules short and simple, and please write your \swarrow
671
       \backsim response on a single line that begins with the text "Rule: ".
672
       673
```

We generate from the above prompt at temperature=0, and split on line breaks to obtain candidate rules.

676 A.4.2 Translating from natural language to Python

We translate Number Game concepts from English to Python via the following prompt for codedavinci-002, and generate at temperature=0 until linebreak:

```
679 # Write a python function to check if a number is C.
680 def check_number(num):
681 return
```

We translate the logic cool concepts from English to Python using a series of in-context examples, again generating with temperature=0 until the text #DONE is produced.²

```
def check_object(this_object, other_objects):
684
        .....
685
        this_object: a tuple of (shape, color, size)
686
        other_objects: a list of tuples of (shape, color, size)
687
688
        returns: True if 'this_object' is positive according to the \checkmark
689
            └→ following rule:
690
691
            Something is positive if it is not a small object, and not 2
692
                .....
693
        # shape: a string, either "circle", "rectangle", or "triangle"
694
        # color: a string, either "yellow", "green", or "blue"
# size: an int, either 1 (small), 2 (medium), or 3 (large)
695
696
        this_shape, this_color, this_size = this_object
697
698
        # 'this_object' is not a part of 'other_objects'
699
700
         to get all of the examples, you can use \swarrow
            \, 'all_example_objects', defined as 'other_objects + 2
701
            \ [this_object]'
702
        703
704
        all_example_objects = other_objects + [this_object]
705
706
        # Something is positive if it is not a small object, and not a \varkappa
707
            708
```

²This prompt is pretty long, and probably could be much shorter. Preliminary experiments suggested that a few in-context examples were very helpful, and so to increase the odds of the model working without much time spent prompt-engineering, we provided a large number of in-context examples.

```
#START
709
        return (not this_size == 1) and (not this_color == "green")
710
711
    #DONE
712
   def check_object(this_object, other_objects):
713
714
715
        this_object: a tuple of (shape, color, size)
        other_objects: a list of tuples of (shape, color, size)
716
717
        returns: True if 'this_object' is positive according to the 🖌
718
719
            └→ following rule:
            Something is positive if it is bigger than every other object
720
        .....
721
        # shape: a string, either "circle", "rectangle", or "triangle"
722
        # color: a string, either "yellow", "green", or "blue"
723
        # size: an int, either 1 (small), 2 (medium), or 3 (large)
724
        this_shape, this_color, this_size = this_object
725
726
727
        # 'this_object' is not a part of 'other_objects'
728
        # to get all of the examples, you can use \checkmark
            \lor 'all_example_objects', defined as 'other_objects + \swarrow
729
            \ [this_object]'
730
        # be careful as to whether you should be using \swarrow
731
            \, 'all_example_objects' or 'other_objects' in your code
732
733
        all_example_objects = other_objects + [this_object]
734
        # Something is positive if it is bigger than every other object
735
736
        #START
        return all( this_size > other_object[2] for other_object in 2
737
738
            \u03c6 other_objects )
   #DONE
739
740
   def check_object(this_object, other_objects):
741
742
743
        this_object: a tuple of (shape, color, size)
        other_objects: a list of tuples of (shape, color, size)
744
745
        returns: True if 'this_object' is positive according to the \checkmark
746
            └→ following rule:
747
            Something is positive if it is one of the largest
748
        .....
749
        # shape: a string, either "circle", "rectangle", or "triangle"
# color: a string, either "yellow", "green", or "blue"
750
751
        # size: an int, either 1 (small), 2 (medium), or 3 (large)
752
        this_shape, this_color, this_size = this_object
753
754
        # 'this_object' is not a part of 'other_objects'
755
        # to get all of the examples, you can use \checkmark
756
            \lor 'all_example_objects', defined as 'other_objects + \swarrow
757
            \ [this_object]'
758
        759
760
        all_example_objects = other_objects + [this_object]
761
762
763
        # Something is positive if it is one of the largest
764
        #START
        return all( this_size >= other_object[2] for all_example_object 🖉
765
            \u03c5 in all_example_objects )
766
   #DONE
767
768
769
   def check_object(this_object, other_objects):
770
771
        this_object: a tuple of (shape, color, size)
772
        other_objects: a list of tuples of (shape, color, size)
773
```

```
returns: True if 'this_object' is positive according to the 🖌
775
776
            └→ following rule:
             Something is positive if it is smaller than every yellow \checkmark
777
778
                 ♦ object
        .....
779
        # shape: a string, either "circle", "rectangle", or "triangle"
780
        # color: a string, either "yellow", "green", or "blue"
781
        # size: an int, either 1 (small), 2 (medium), or 3 (large)
782
783
        this_shape, this_color, this_size = this_object
784
        # 'this_object' is not a part of 'other_objects'
785
        # to get all of the examples, you can use \checkmark
786
             \, 'all_example_objects', defined as 'other_objects + 2
787
            \ [this_object]'
788
        # be careful as to whether you should be using \swarrow
789
            \u00ed 'all_example_objects' or 'other_objects' in your code
790
        all_example_objects = other_objects + [this_object]
791
792
793
        # Something is positive if it is smaller than every yellow object
794
        #START
        return all( this_size < other_object[2] for other_object in 2
795
            \u2265 other_objects if other_object[1] == "yellow" )
796
    #DONE
797
798
    def check_object(this_object, other_objects):
799
800
        this_object: a tuple of (shape, color, size)
801
        other_objects: a list of tuples of (shape, color, size)
802
803
        returns: True if 'this_object' is positive according to the 🖌
804
805
            └> following rule:
             Something is positive if there is another object with the 2
806
807
                 .....
808
        # shape: a string, either "circle", "rectangle", or "triangle"
# color: a string, either "yellow", "green", or "blue"
# size: an int, either 1 (small), 2 (medium), or 3 (large)
809
810
811
        this_shape, this_color, this_size = this_object
812
813
        # 'this_object' is not a part of 'other_objects'
814
        # to get all of the examples, you can use \swarrow
815
            \, 'all_example_objects', defined as 'other_objects + 
816
            \ [this_object]'
817
        818
819
        all_example_objects = other_objects + [this_object]
820
821
        # Something is positive if there is another object with the \checkmark
822
823
            Same color
        #START
824
        return any( this_color == other_object[1] for other_object in \varkappa
825
826
            \u03c6 other_objects )
    #DONE
827
828
    def check_object(this_object, other_objects):
829
        .....
830
831
        this_object: a tuple of (shape, color, size)
        other_objects: a list of tuples of (shape, color, size)
832
833
        returns: True if 'this_object' is positive according to the \swarrow
834
            └→ following rule:
835
836
             Something is positive if it has a unique combination of \checkmark
837
                 .....
838
```

774

```
# shape: a string, either "circle", "rectangle", or "triangle"
# color: a string, either "yellow", "green", or "blue"
# size: an int, either 1 (small), 2 (medium), or 3 (large)
839
840
841
        this_shape, this_color, this_size = this_object
842
843
        # 'this_object' is not a part of 'other_objects'
844
845
        # to get all of the examples, you can use \swarrow
             \, 'all_example_objects', defined as 'other_objects + 
846
             \ [this_object]'
847
        848
849
         all_example_objects = other_objects + [this_object]
850
851
852
        # Something is positive if it has a unique combination of color \varkappa
853
             \backsim and shape
        #START
854
        return all( this_shape != other_object[0] or this_color != 🖌
855
             \u03c3 other_object[1] for other_object in other_objects )
856
857
    #DONE
858
    def check_object(this_object, other_objects):
859
         .....
860
861
        this_object: a tuple of (shape, color, size)
        other_objects: a list of tuples of (shape, color, size)
862
863
        returns: True if 'this_object' is positive according to the 🗸
864
             └→ following rule:
865
             Something is positive if it has the same color as the \checkmark
866
867
                 .....
868
        # shape: a string, either "circle", "rectangle", or "triangle"
869
        # color: a string, either "yellow", "green", or "blue"
870
        # size: an int, either 1 (small), 2 (medium), or 3 (large)
871
872
        this_shape, this_color, this_size = this_object
873
874
        # 'this_object' is not a part of 'other_objects'
875
        # to get all of the examples, you can use \swarrow
             \backsim 'all_example_objects', defined as 'other_objects + \swarrow
876
             \ [this_object]'
877
         # be careful as to whether you should be using \swarrow
878
             \ 'all_example_objects' or 'other_objects' in your code
879
         all_example_objects = other_objects + [this_object]
880
881
        # Something is positive if it has the same color as the \checkmark
882
             └→ majority of objects
883
884
         #START
        majority_color = max(["yellow", "green", "blue"], key=lambda 2
885
             └> color: sum(1 for obj in all_example_objects if obj[1] == 
886
             └ color))
887
        return this_color == majority_color
888
    #DONE
889
890
    def check_object(this_object, other_objects):
891
892
893
         this_object: a tuple of (shape, color, size)
894
        other_objects: a list of tuples of (shape, color, size)
895
896
        returns: True if 'this_object' is positive according to the 🖌
             └→ following rule:
897
             Something is positive if there are at least two other \varkappa
898
899
                 .....
900
        # shape: a string, either "circle", "rectangle", or "triangle"
# color: a string, either "yellow", "green", or "blue"
# size: an int, either 1 (small), 2 (medium), or 3 (large)
901
902
903
```

```
this_shape, this_color, this_size = this_object
904
905
        # 'this_object' is not a part of 'other_objects'
906
        # to get all of the examples, you can use \swarrow
907
               'all_example_objects', defined as 'other_objects + 🖉
908
            \ [this_object]'
909
        # be careful as to whether you should be using \swarrow
910
            \ 'all_example_objects' or 'other_objects' in your code
911
        all_example_objects = other_objects + [this_object]
912
913
914
        # Something is positive if there are at least two other objects \checkmark
915

    ↓ with the same shape

        #START
916
917
        return sum(1 for other_object in other_objects if 2
            \u03c3 other_object[0] == this_shape) >= 2
918
   #DONE
919
920
   def check_object(this_object, other_objects):
921
        .....
922
923
        this_object: a tuple of (shape, color, size)
        other_objects: a list of tuples of (shape, color, size)
924
925
        returns: True if 'this_object' is positive according to the 🖌
926
927
            └→ following rule:
            C
928
        .....
929
        # shape: a string, either "circle", "rectangle", or "triangle"
930
        # color: a string, either "yellow", "green", or "blue"
# size: an int, either 1 (small), 2 (medium), or 3 (large)
931
932
        this_shape, this_color, this_size = this_object
933
934
        # 'this_object' is not a part of 'other_objects'
935
        # to get all of the examples, you can use \swarrow
936
            \, 'all_example_objects', defined as 'other_objects + 
937
            \ [this_object]'
938
        939
940
941
        all_example_objects = other_objects + [this_object]
942
        # C
943
        #START
944
```

945 A.5 GPT-4 Baselines

Our GPT-4 baseline for each domain presented the examples $X_{1:K}$ in string form and then asked GPT-4 to respond Yes/No as to whether a test example X_{test} belonged to the same concept. GPT-4 was then queried at temperature=1 to collect 10 samples. Samples not beginning with 'y'/'n' were discarded, and the ratio of remaining samples that began with 'y' was computed (case insensitive).

⁹⁵⁰ We show below example prompts for the number and logic domains.

```
Here are a few example number concepts:
951
    -- The number is even
952
953
    -- The number is between 30 and 45
    -- The number is a power of 3
954
    -- The number is less than 10
955
956
   Here are some random examples of numbers belonging to a possibly \varkappa
957
        └> different number concept:
958
   98, 81, 86, 93
959
960
   Question: Does the number 42 belong to the same concept as the \swarrow
961
962
       Answer (one word, yes/no):
963
```

964 Logical concept example prompt:

Here are some example concepts defined by a logical rule: 965 966 Rule for Concept #1: Something is positive if it is the biggest \varkappa 967 └, yellow object in the example 968 Rule for Concept #2: Something is positive if there is another \varkappa 969 └ object with the same color in the example 970 Rule for Concept #3: Something is positive if it is the same color \varkappa 971 972 973 974 Now please look at the following examples for a new logical rule. 975 An Example of Concept #4: 976 POSITIVES: none 977 978 NEGATIVES: (large yellow circle), (small green circle), 🖌 ५ (medium green circle), (small yellow triangle) 979 Another Example of Concept #4: 980 981 POSITIVES: (small green circle), (large green circle) NEGATIVES: (large yellow circle), (medium blue circle) 982 Another Example of Concept #4: 983 984 POSITIVES: (small green rectangle) <code>NEGATIVES: (medium yellow circle), (medium blue rectangle), \varkappa </code> 985 986 987 Another Example of Concept #4: POSITIVES: (medium green rectangle) 988 NEGATIVES: (medium yellow circle), (small yellow 🖌 989 \searrow rectangle), (medium yellow rectangle), (medium blue \swarrow 990 └→ rectangle) 991 992 Another Example of Concept #4: 993 POSITIVES: (small green rectangle) <code>NEGATIVES: (large yellow rectangle), (small yellow \checkmark </code> 994 \, triangle), (medium green circle), (small blue rectangle) 995 Another Example of Concept #4: 996 POSITIVES: (medium green triangle) 997 NEGATIVES: (medium blue triangle), (medium blue rectangle), 🖌 998 \backsim (large blue triangle), (small yellow triangle) 999 Another Example of Concept #4: 1000 1001 POSITIVES: none NEGATIVES: (small yellow circle), (large blue circle) 1002 Another Example of Concept #4: 1003 **POSITIVES:** none 1004 NEGATIVES: (large green circle), (small blue rectangle), 🖉 1005 ५ (small green triangle), (medium blue rectangle) 1006 1007 Another Example of Concept #4: POSITIVES: (small green rectangle) 1008 NEGATIVES: (small yellow circle), (large blue rectangle) 1009 1010 Now we get a new collection of examples for Concept #4: 1011 (medium blue triangle) (large yellow triangle) (small blue \swarrow 1012 \, rectangle) (large blue circle) (small yellow circle) 1013 1014 Question: Based on the above example, is a (small yellow circle) in \swarrow \lor the concept? 1015 1016 Answer (one word, just write yes/no):

1017 A.6 Latent Language Baseline

For fair comparison, we designed our latent language baseline to be as similar to our system as possible. It performs maximum likelihood estimation of a single concept, rather than estimate a full posterior, but uses the exact same prompts and likelihood functions as our model. The most important difference from the original latent language paper [22] is that instead of training our own neural models for language interpretation and language generation, we use pretrained models (Codex/code-davinci-002 and GPT-4).

1024 A.7 Ablation of the proposal distribution

We ablate the proposal distribution by proposing hypotheses unconditioned on $X_{1:K}$. We accomplish this by drawing concepts from the following alternative prompt, which is designed to resemble the prompt used by the full model except that it does not include $X_{1:K}$:

```
1028 # Python 3
1029 # Here are a few example number concepts:
1030 # -- The number is even
1031 # -- The number is between 30 and 45
1032 # -- The number is a power of 3
1033 # -- The number is less than 10
1034 # -- The number is
```

1035 A.8 Pretrained prior

Our pretrained prior comes from the opensource model CodeGen [34], which was trained on source code. We chose this model because we suspected that pretraining on source code would give better density estimation for text describing precise rules. We formatted the rules as a natural language comment and prefixed it with a small amount of domain-specific text in order to prime the model to put probability mass on rules that correctly talk about numbers or shapes.

¹⁰⁴¹ For the number game, we would query CodeGen for the probability of p(C) via

```
1042   
# Here is an example number concept: 1043   
# The number is {\cal C}
```

For the number game's code baseline, we would query CodeGen for the probability of p(C) via

```
1045 # Python 3
1046 # Let's think of a number concept.
1047 # Write a python function that returns true if 'num' belongs to 2
1048 ↓ this number concept.
1049 def check_if_in_concept(num):
1050 return C
```

¹⁰⁵¹ For logical concepts we would query CodeGen for the probability of p(C) via

```
1052 # Here are some simple example shape concepts:
1053 # 1. neither a triangle nor a green rectangle
1054 # 2. not blue and large.
1055 # 3. if it is large, then it must be yellow.
1056 # 4. small and blue
1057 # 5. either big or green.
1058 # 6. C
```

Because the proposal distribution would generate rules beginning with the prefix "Something is positive if..." we would remove that text before computing p(C) as above.