

Coarse-to-Fine Syntactic MT using Language Projections



Slav Petrov

Aria Haghighi

Dan Klein



Why is MT so slow?



Why is MT so slow?

quiero

salir

Why is MT so slow?

quiero

- ➔ i want
- ➔ i like
- ➔ i love
- ➔ i want to
- ➔ i like to
- ➔ i love to

salir

- ➔ leave
- ➔ to leave
- ➔ go
- ➔ to go
- ➔ exit
- ➔ to exit

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Translations:

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- leave
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- go
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Translations:

- ▶ i want leave **x**

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Translations:

- ▶ i want leave ~~x~~
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Translations:

- ▶ i want leave ✗
- ▶ i like leave ✗
- ▶ i love leave ✗
- ▶ i want to leave ✓

Why is MT so slow?

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- ▶ i want to leave ✓
- ▶ i like to leave ✗
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Why is MT so slow?

There are too
many words!

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- ▶ i want to leave ✗
- ▶ i like to leave ✗
- ▶ ...



Coarse-to-Fine Decoding

Spanish: reanudación del período de sesiones



Coarse-to-Fine Decoding

Spanish: reanudación del período de sesiones

English:

Coarse-to-Fine Decoding

Spanish: reanudación del período de sesiones

English:

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Coarse-to-Fine Decoding

Spanish: reanudación del período de sesiones

English:

resumption	of	the	session
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Coarse-to-Fine Decoding

Spanish: reanudación del período de sesiones

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Coarse-to-Fine Decoding

Spanish: reanudación del período de sesiones

English: resumption of the session

Coarse-to-Fine Decoding

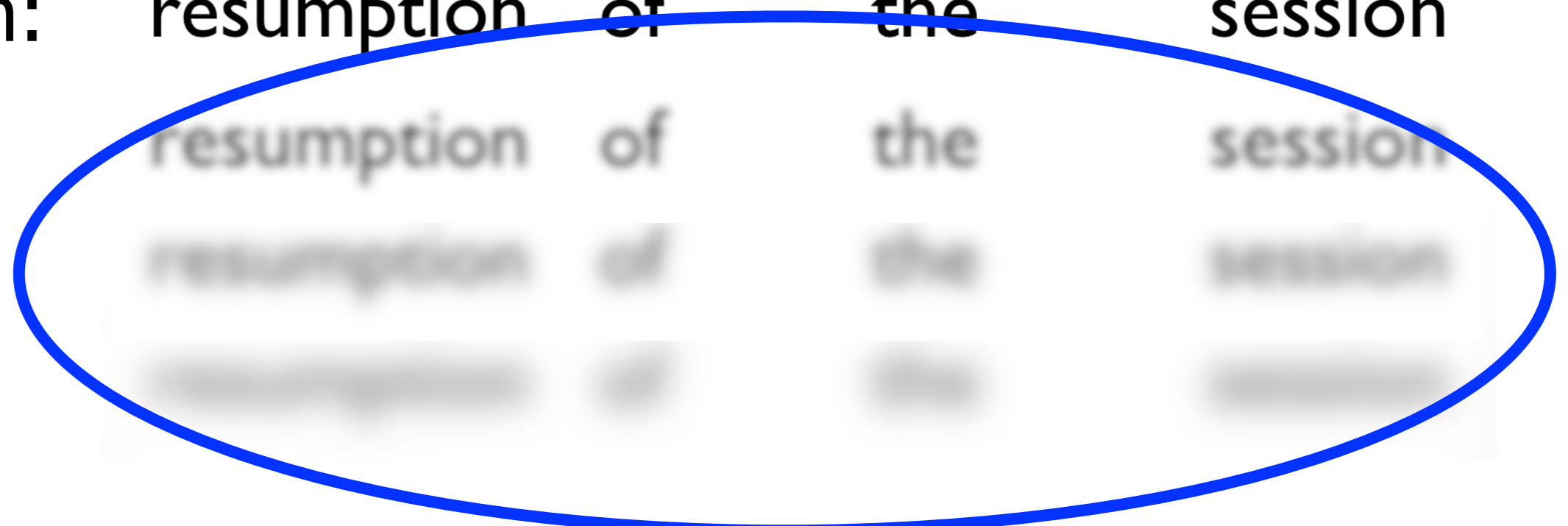
Spanish: reanudación del período de sesiones

English: resumption of the session

resumption of the session

resumption of the session

resumption of the session



Coarse-to-Fine Decoding

Spanish: reanudación del período de sesiones

Syntactic
MT
Decoder

English: resumption of the session

resumption of the session

resumption of the session

resumption of the session



First Pass

Coarse Pass

Coarse-to-Fine Syntactic MT using Language Projections



Slav Petrov

Aria Haghighi

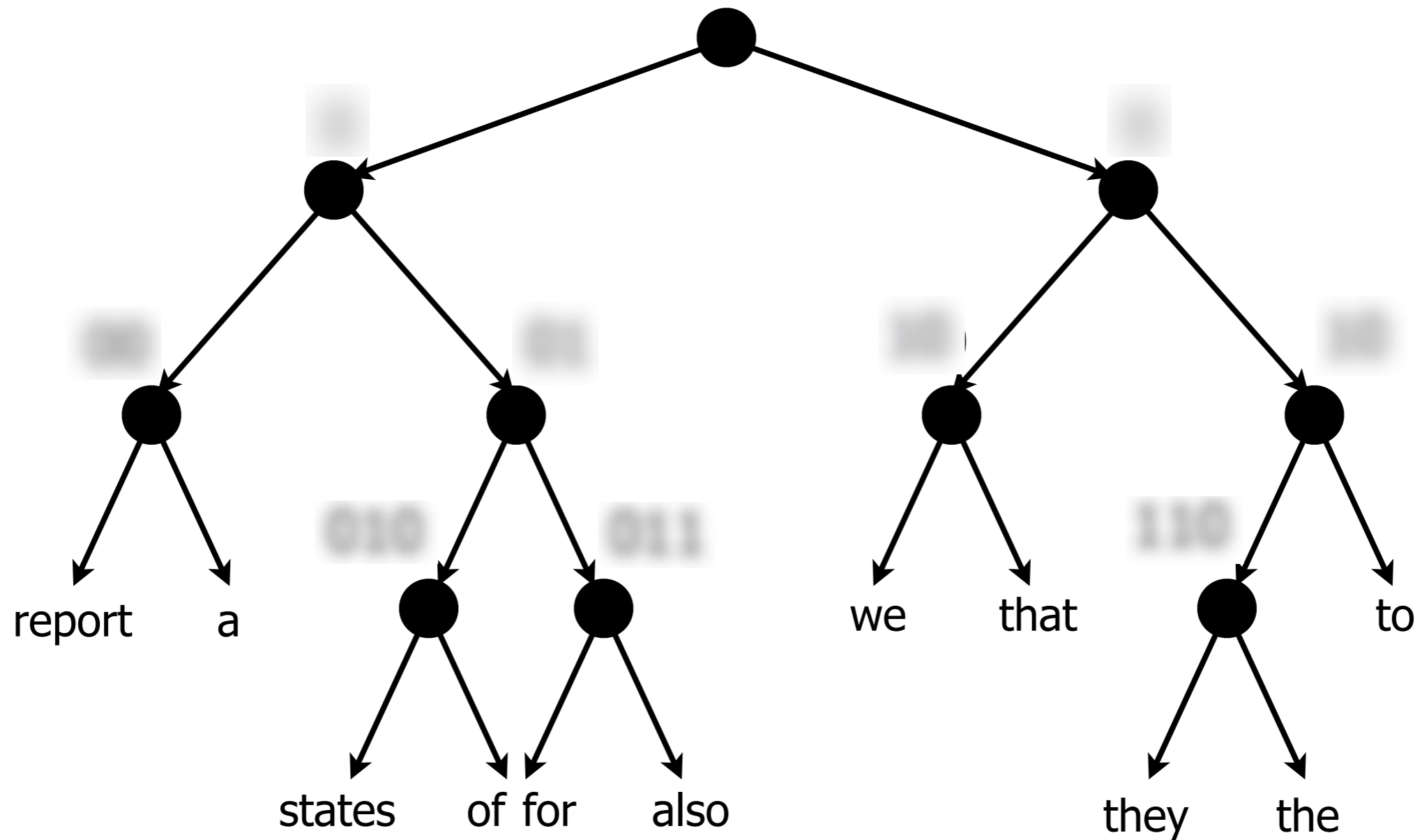
Dan Klein



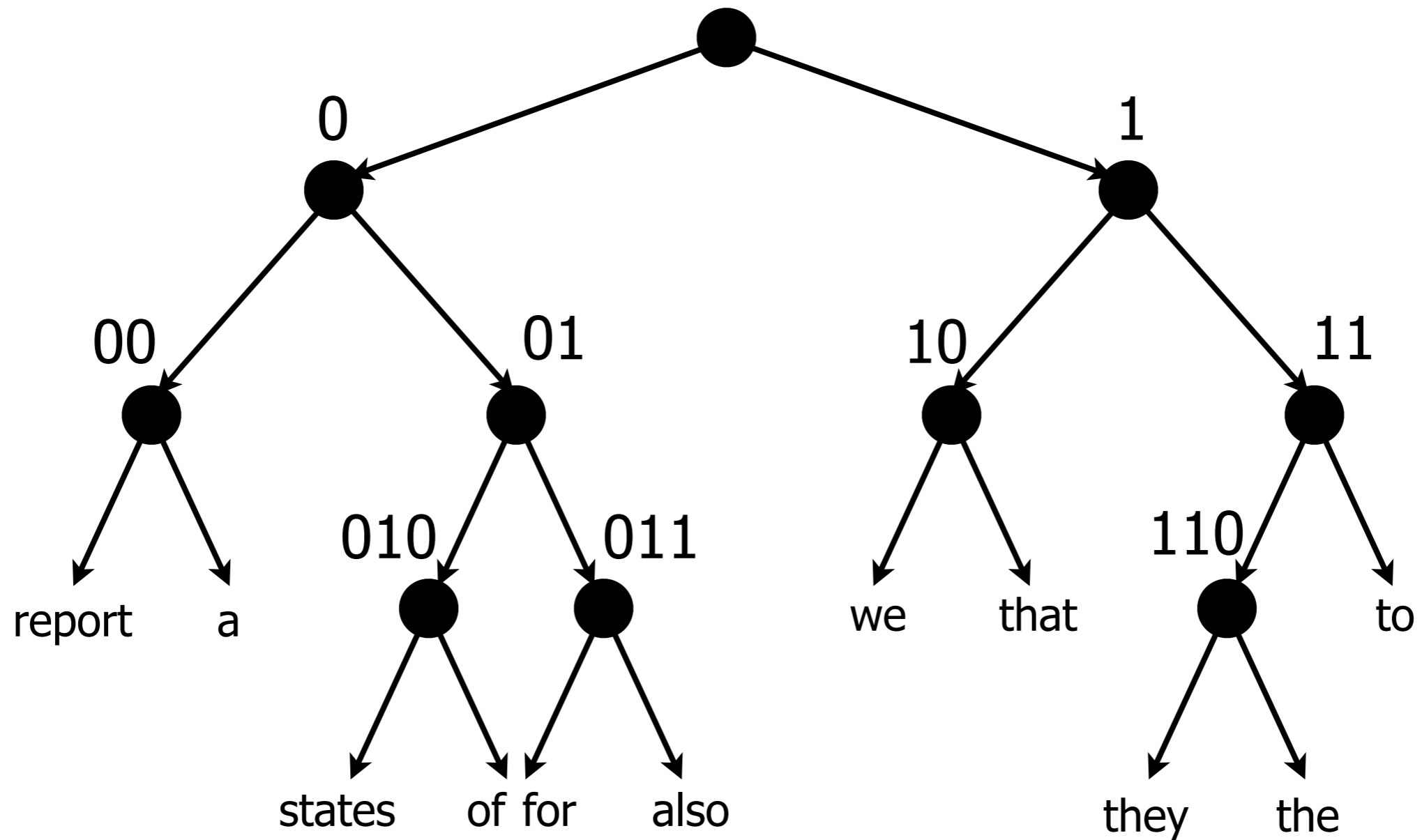
Second Pass

Fine Pass

Hierarchical Clustering



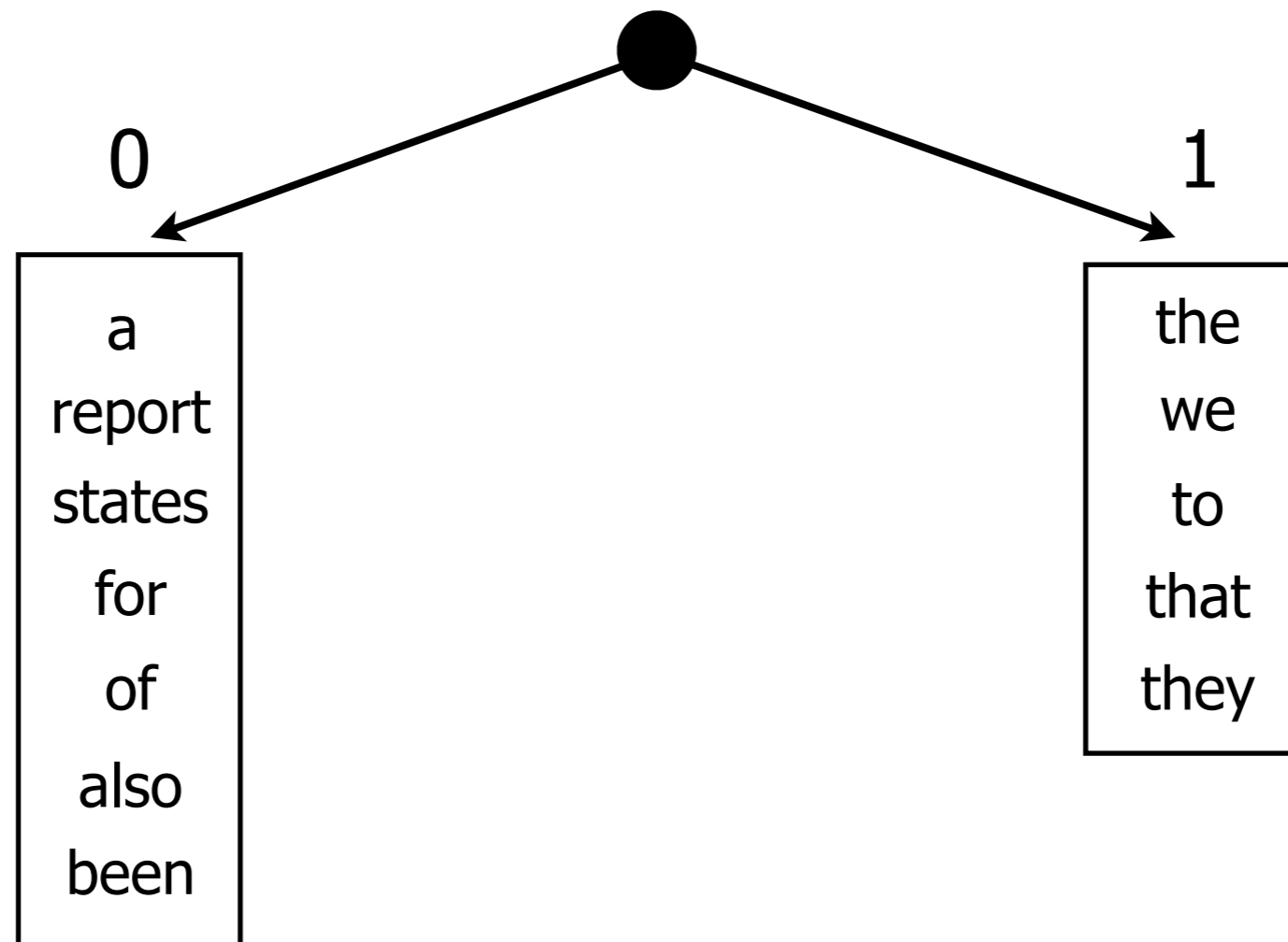
Hierarchical Clustering



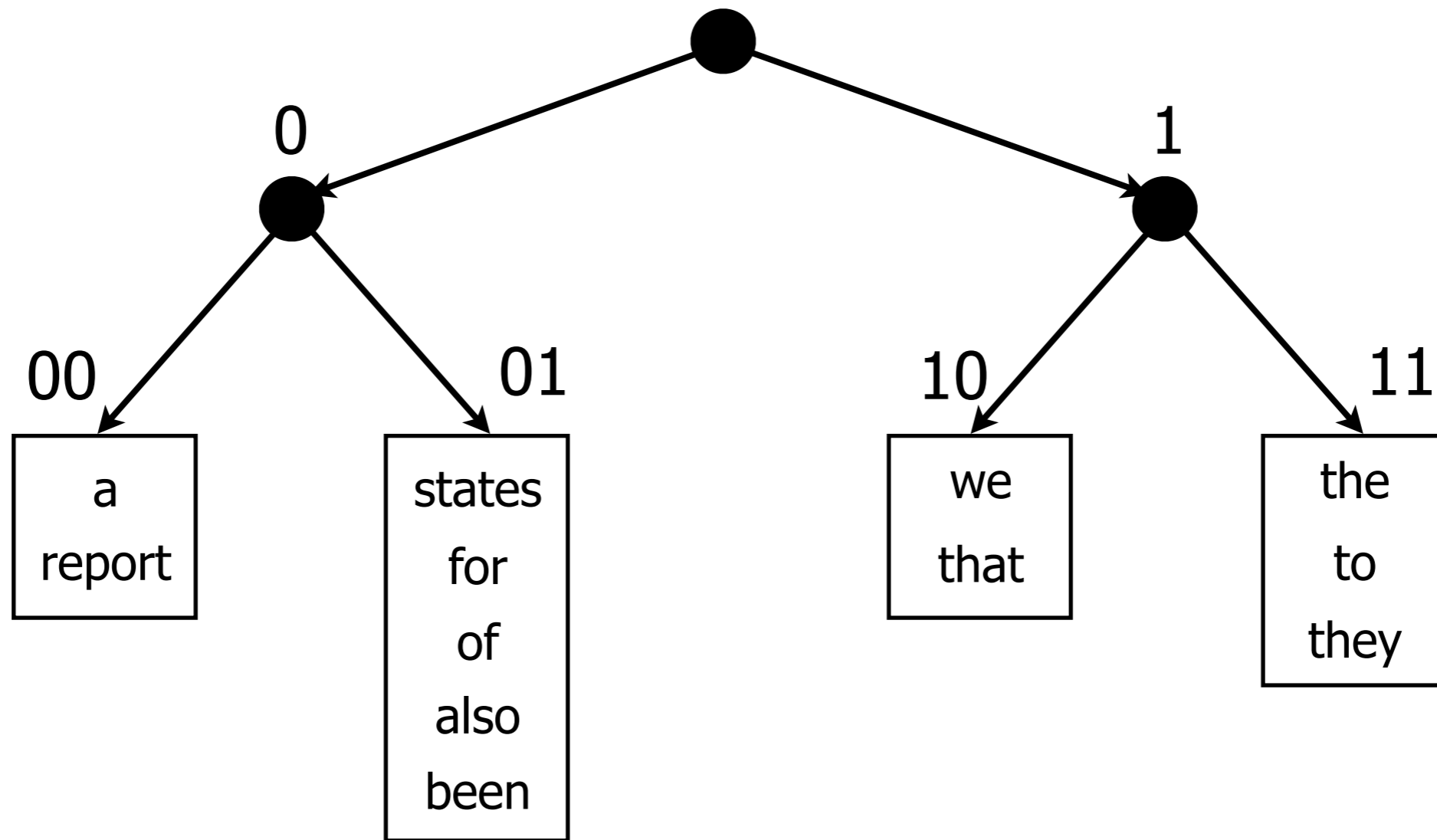
Random Clustering

the
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report
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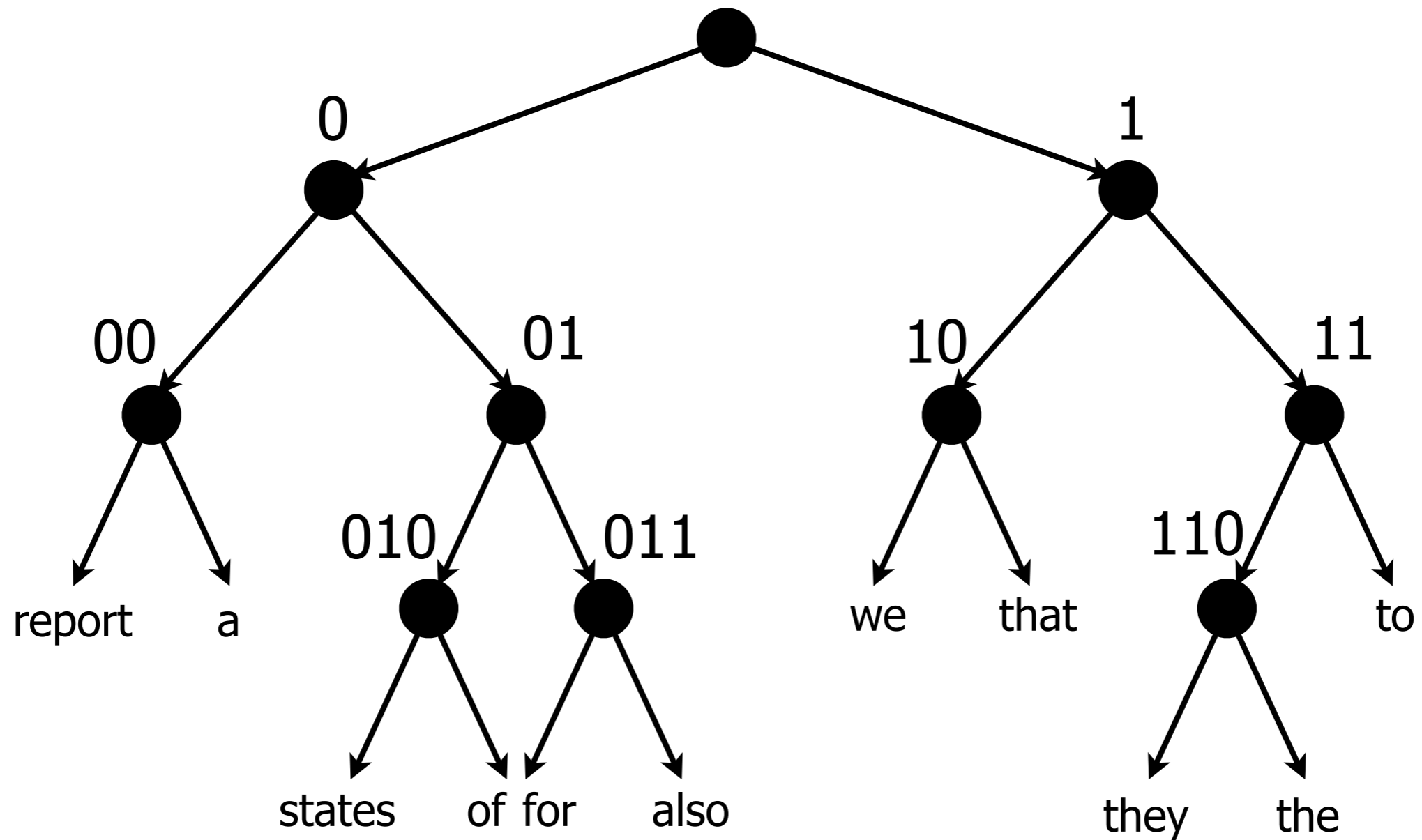
Random Clustering



Random Clustering



Random Clustering





Frequency Clustering

Frequency Clustering

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states

report

Frequency Clustering

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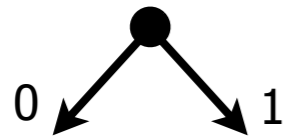
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report

Frequency Clustering



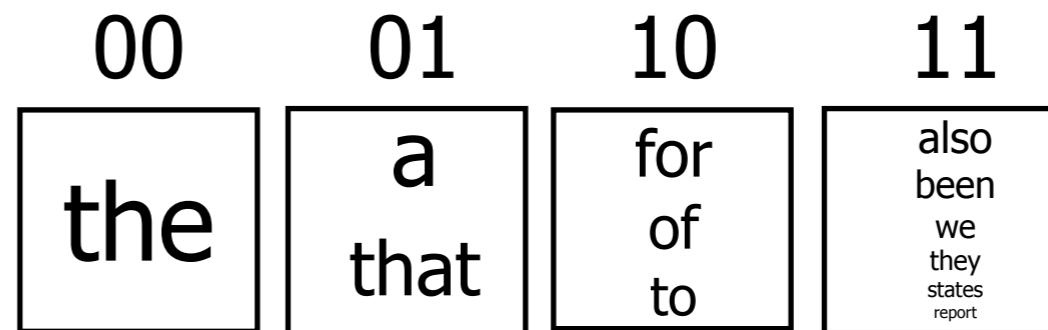
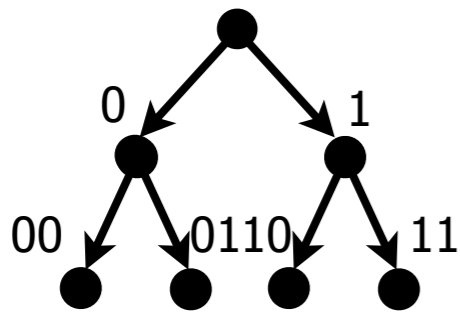
0

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Frequency Clustering





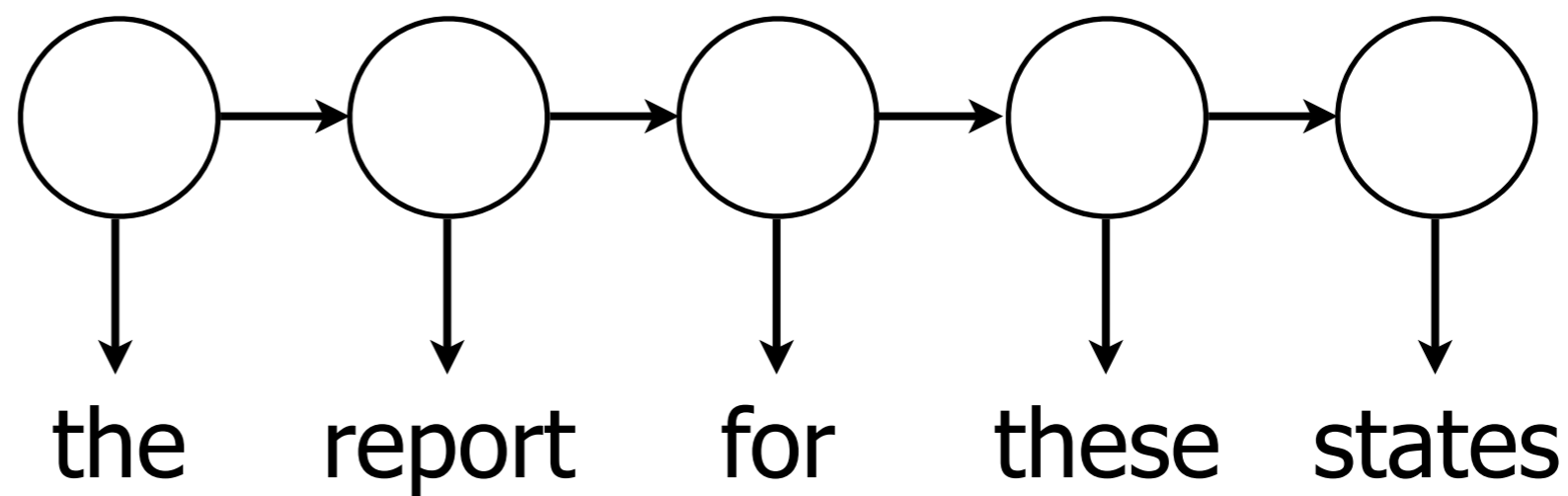
HMM Clustering



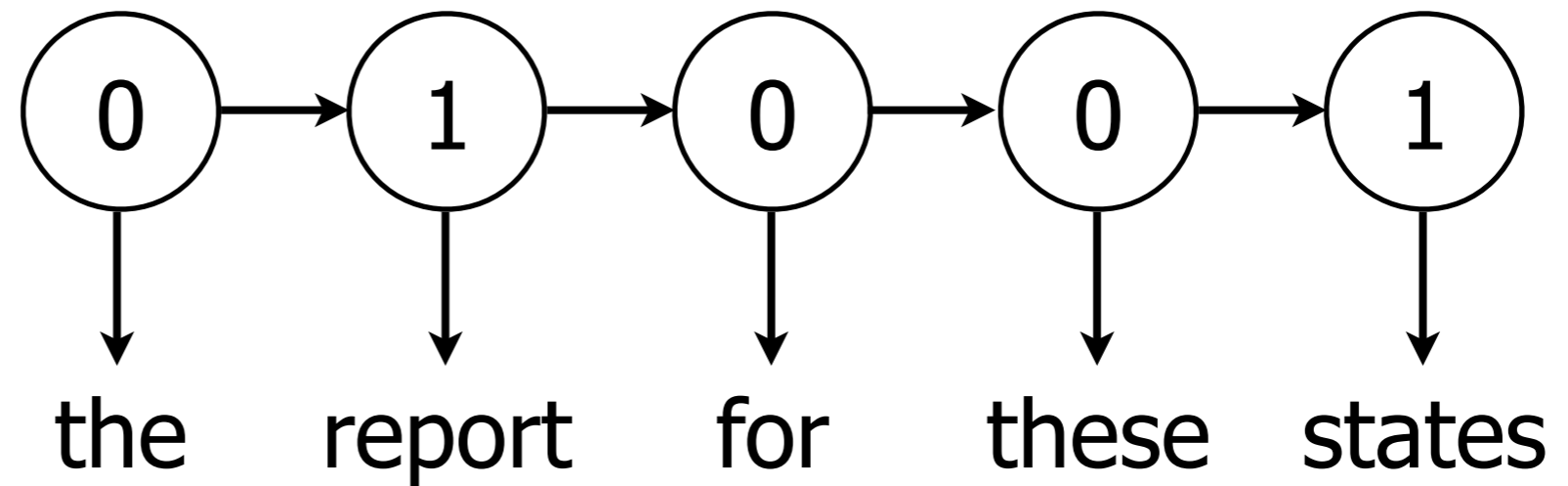
HMM Clustering

the report for these states

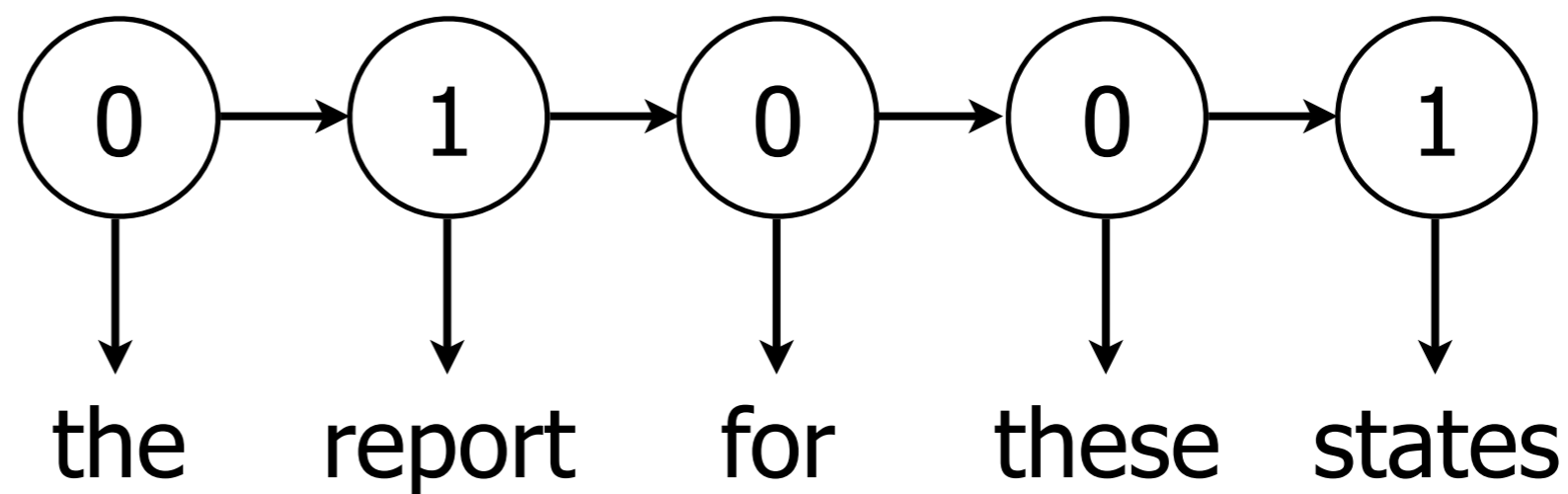
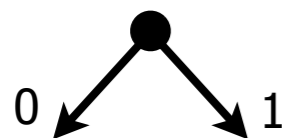
HMM Clustering



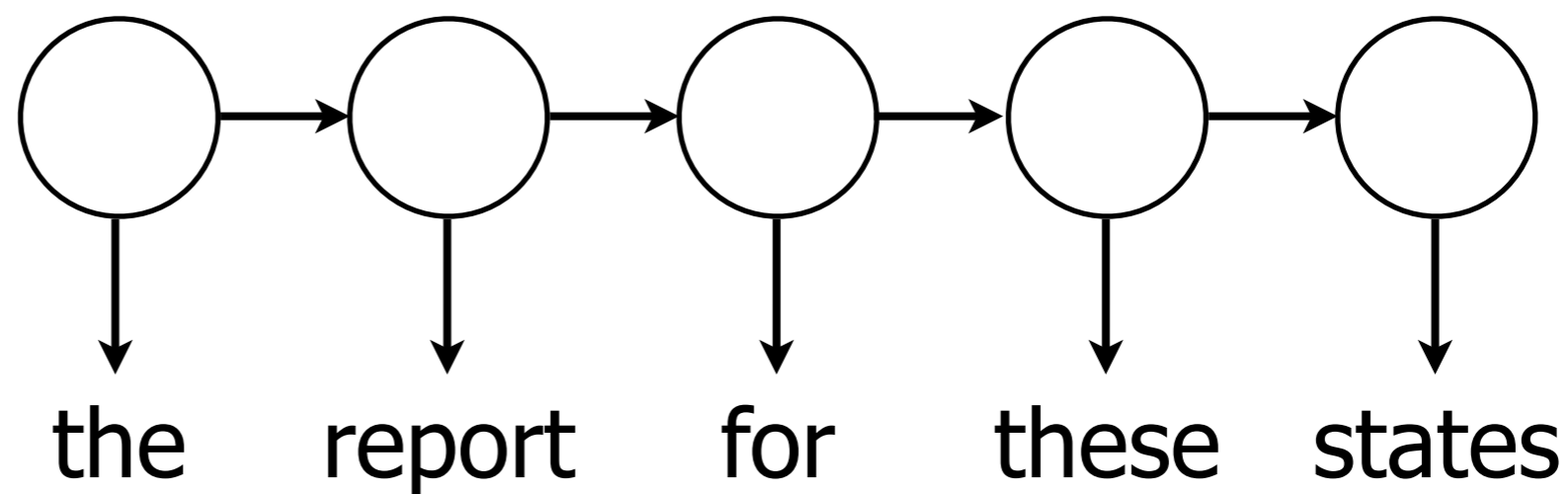
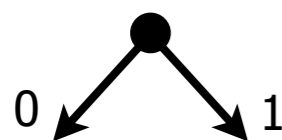
HMM Clustering



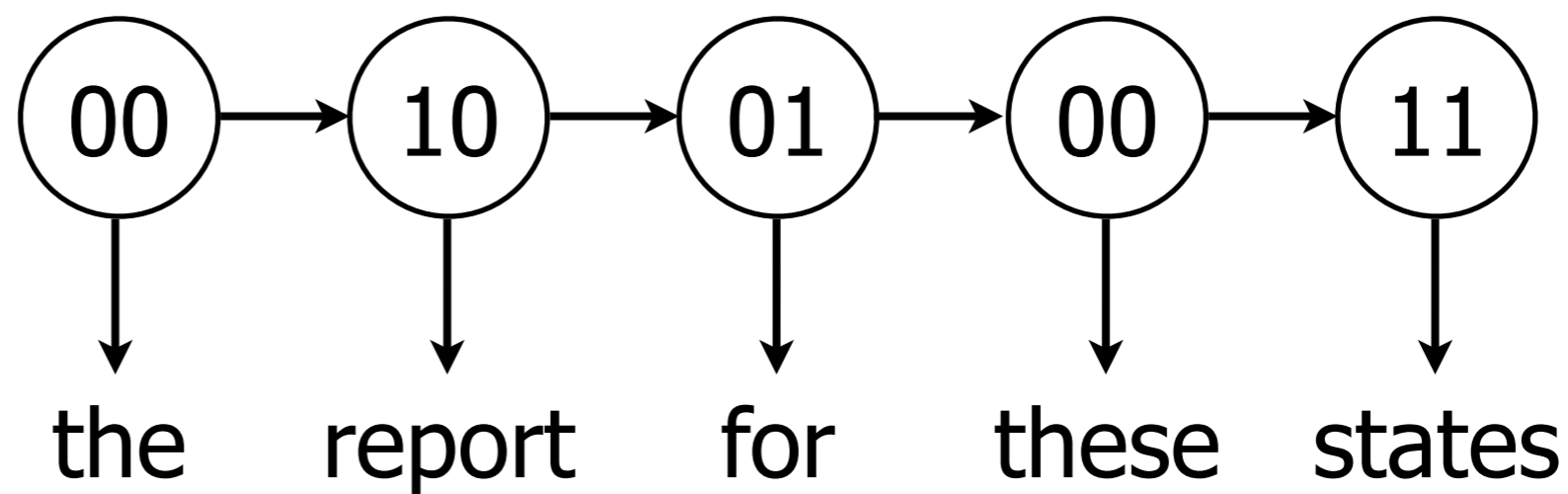
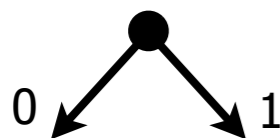
HMM Clustering



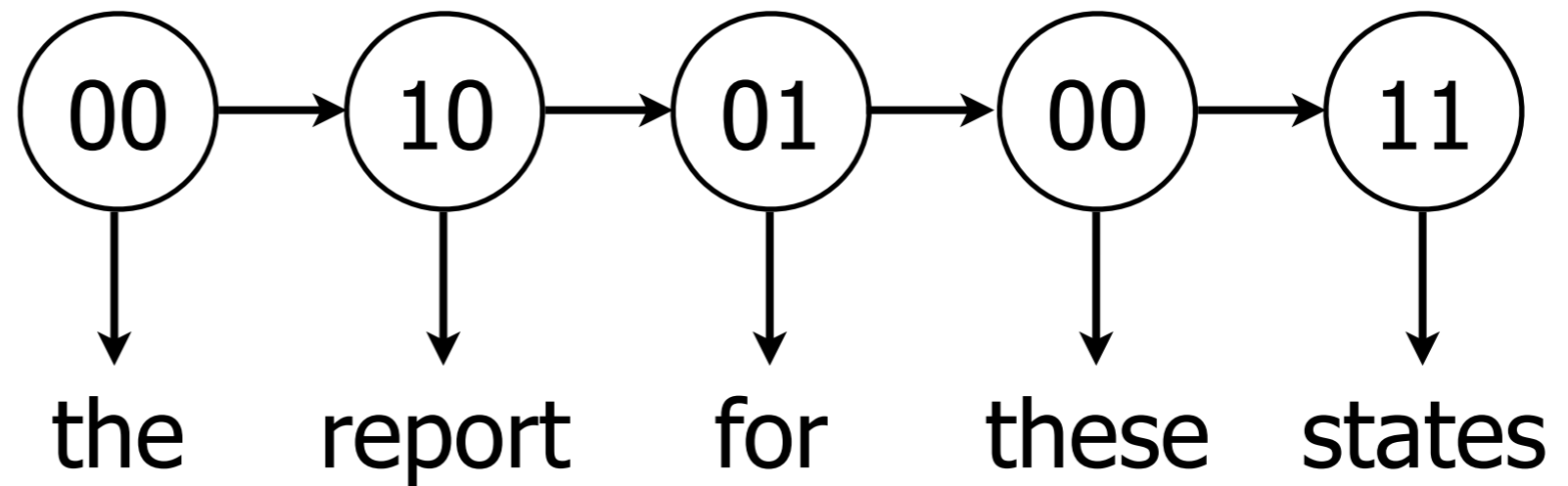
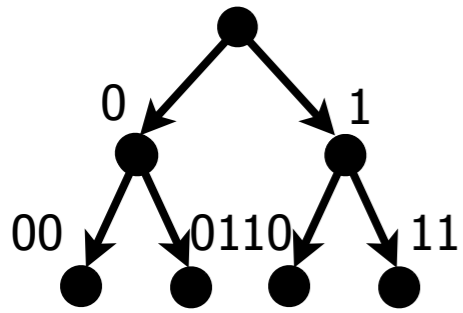
HMM Clustering



HMM Clustering



HMM Clustering

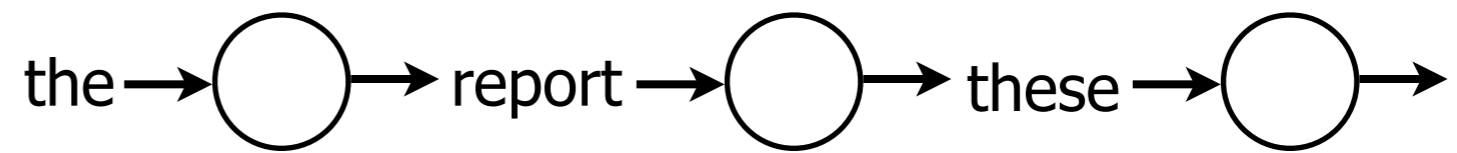




JCluster

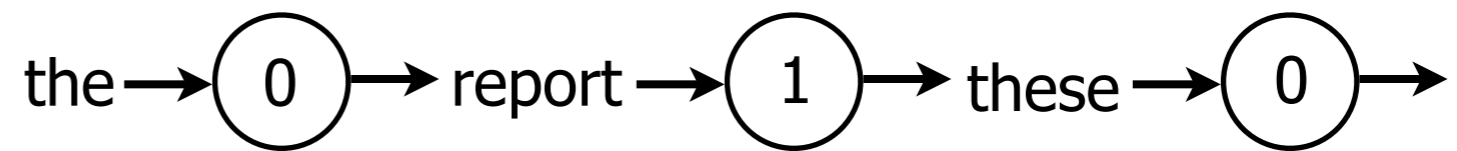
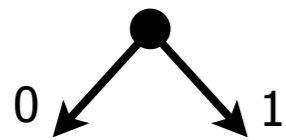
[Goodman '01]

JCluster



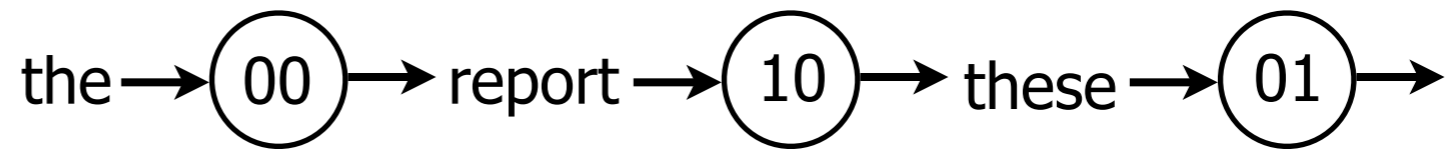
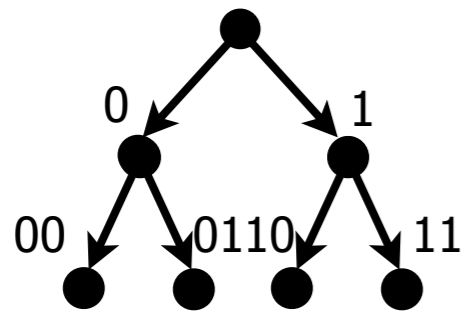
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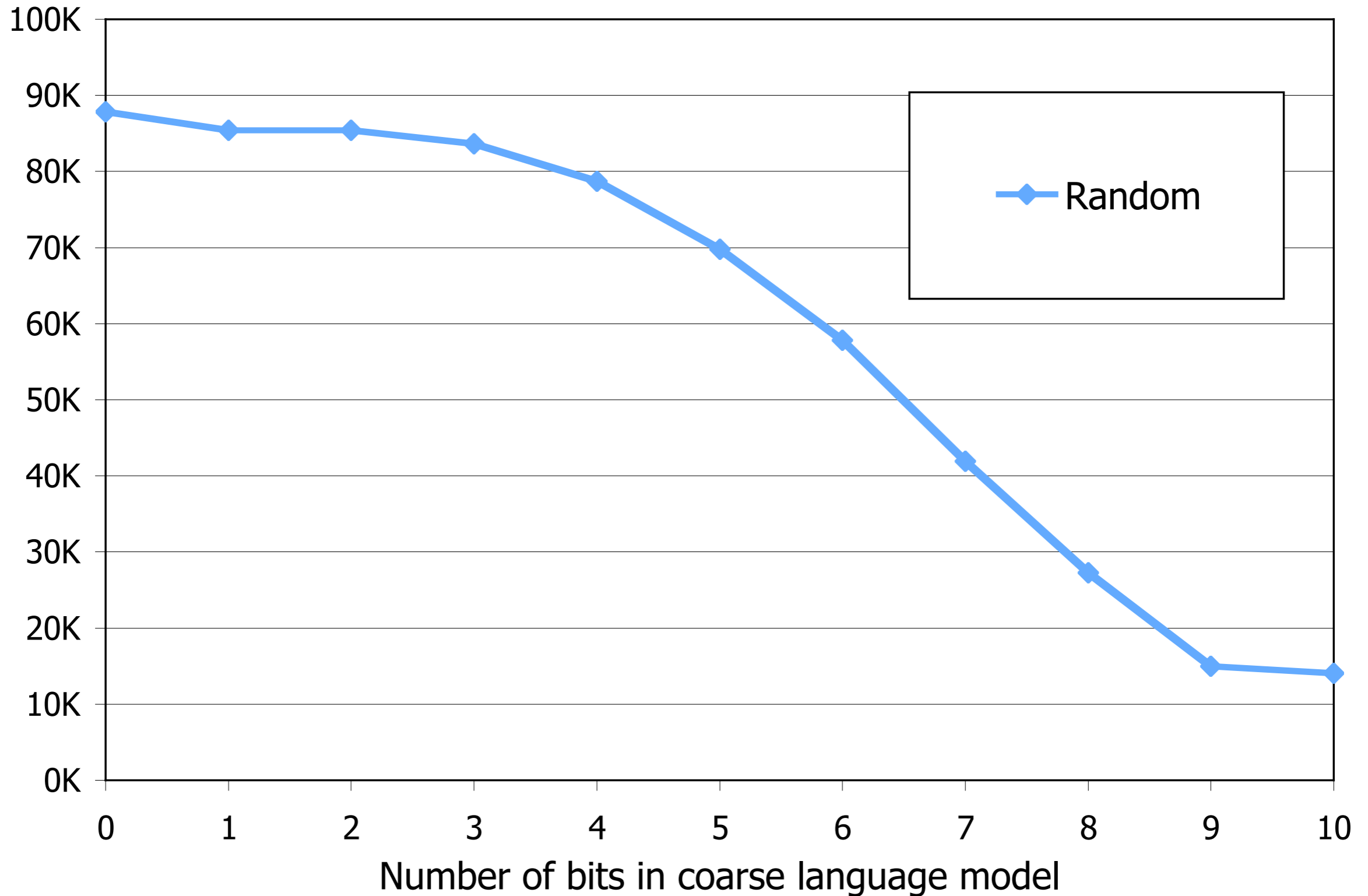
[Goodman '01]

JCluster

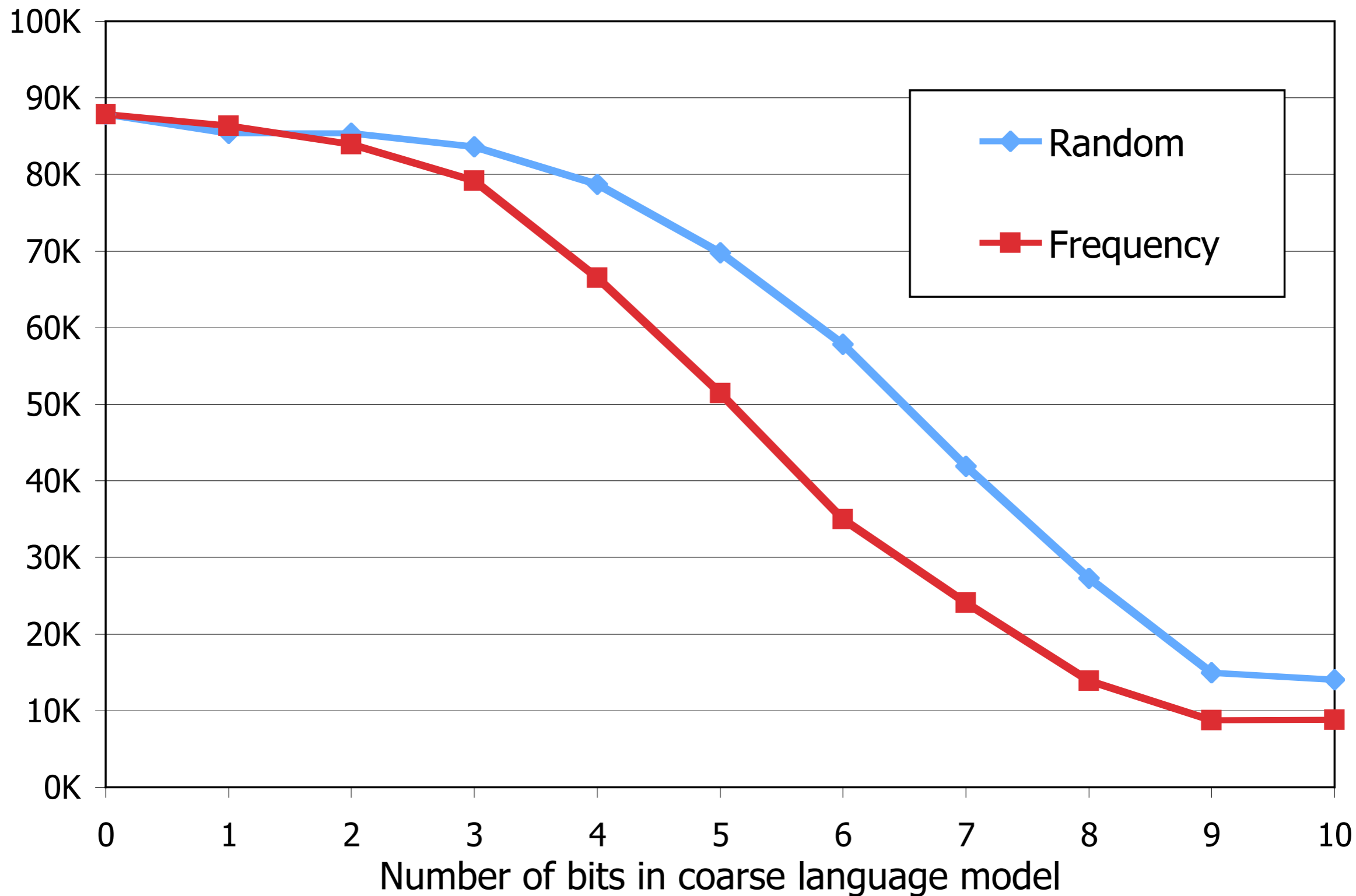


[Goodman '01]

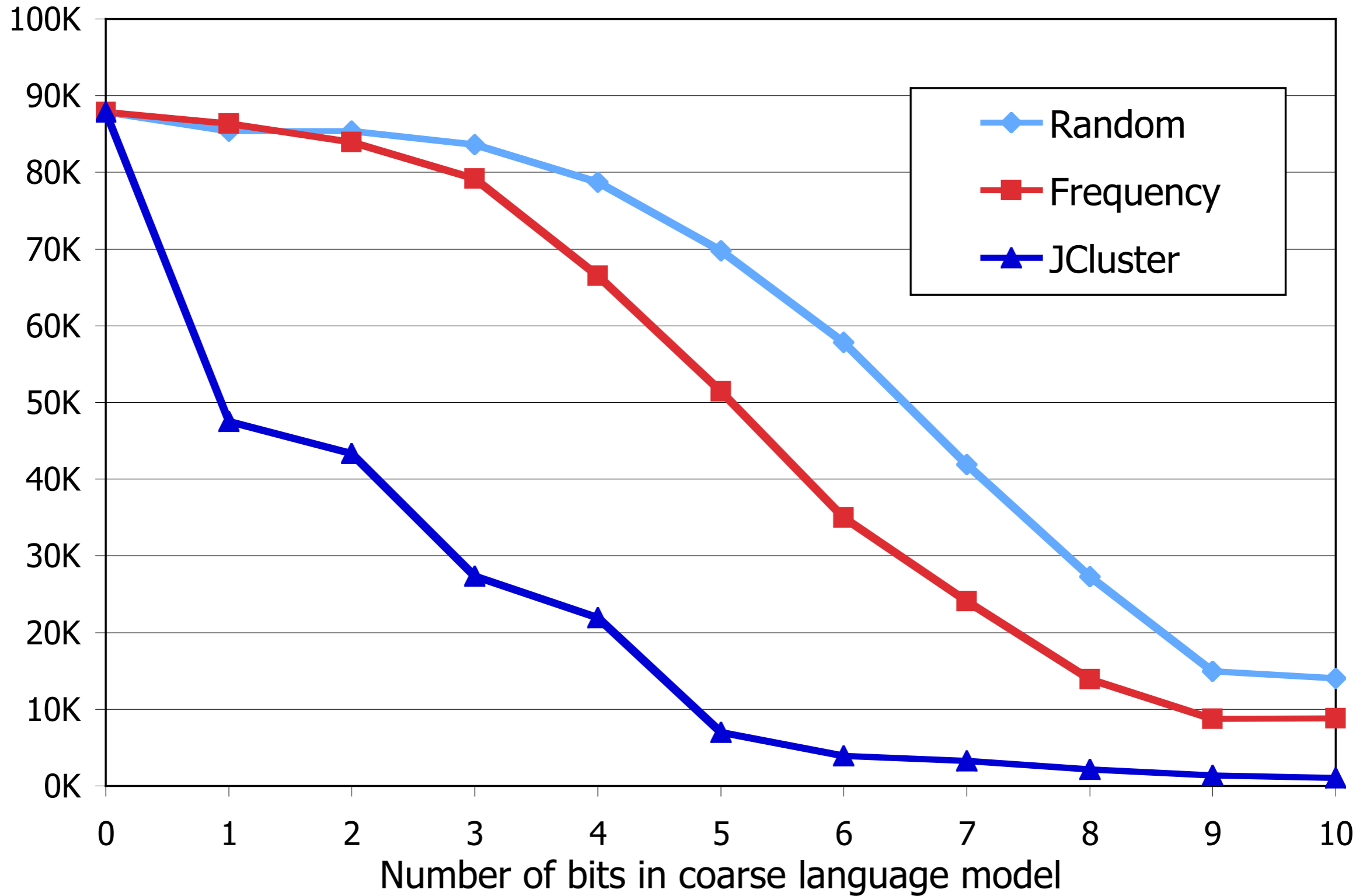
Cluster Quality (Perplexity)



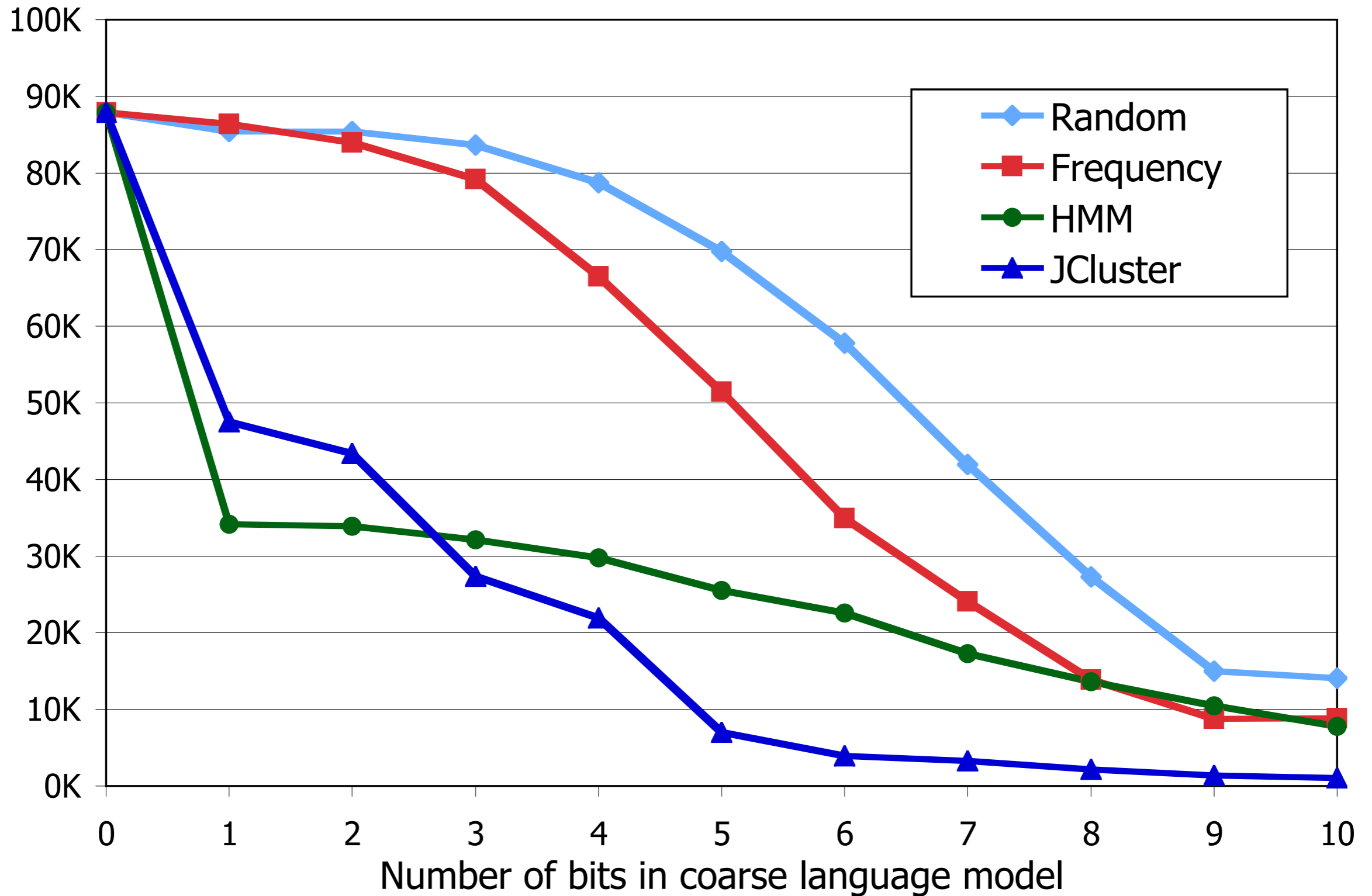
Cluster Quality (Perplexity)



Cluster Quality (Perplexity)



Cluster Quality (Perplexity)





Hierarchical Clustering

Hierarchical Clustering

fine
the
a
they
we
states
report
also
in

Hierarchical Clustering

fine	1-bit
the	0
a	0
they	0
we	0
states	1
report	1
also	1
in	1

Hierarchical Clustering

fine	1-bit	2-bit
the	0	00
a	0	00
they	0	01
we	0	01
states	1	10
report	1	10
also	1	11
in	1	11

Hierarchical Clustering

fine	1-bit	2-bit	3-bit
the	0	00	000
a	0	00	001
they	0	01	010
we	0	01	011
states	1	10	100
report	1	10	101
also	1	11	110
in	1	11	111

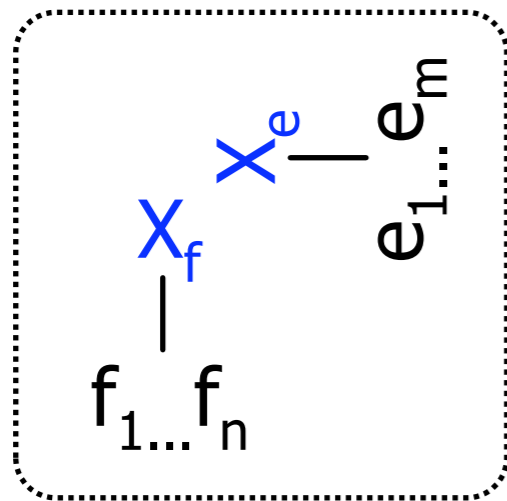


Inversion Transduction Grammars

[Wu '97]

Inversion Transduction Grammars

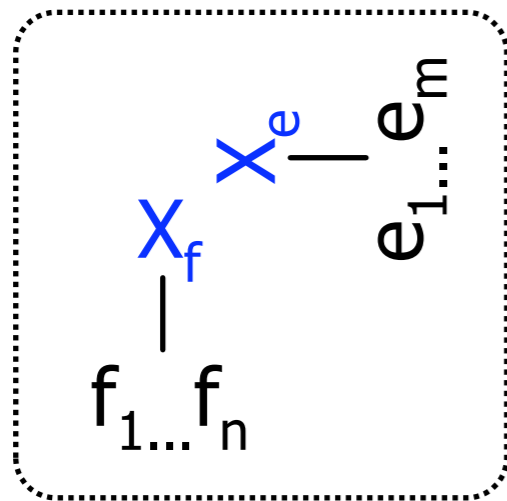
Generation



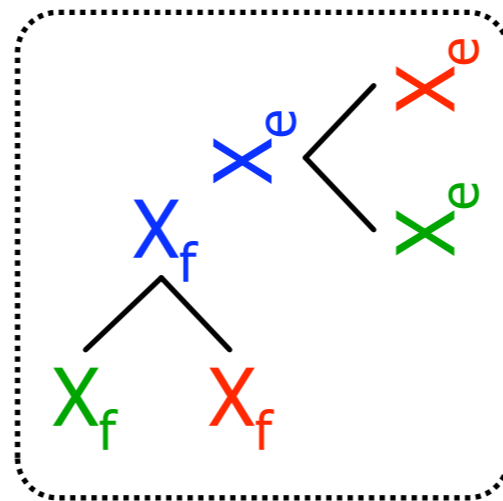
[Wu '97]

Inversion Transduction Grammars

Generation



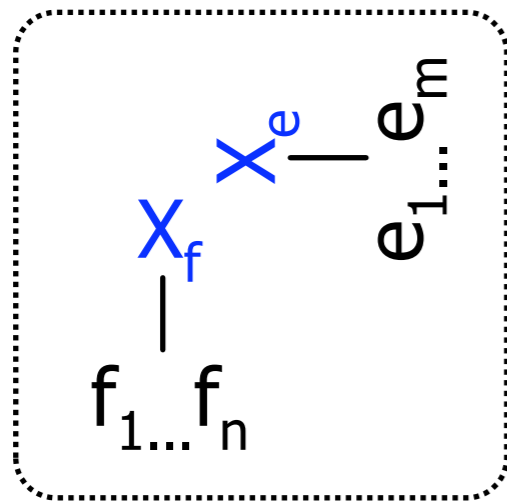
Monotonic Combination



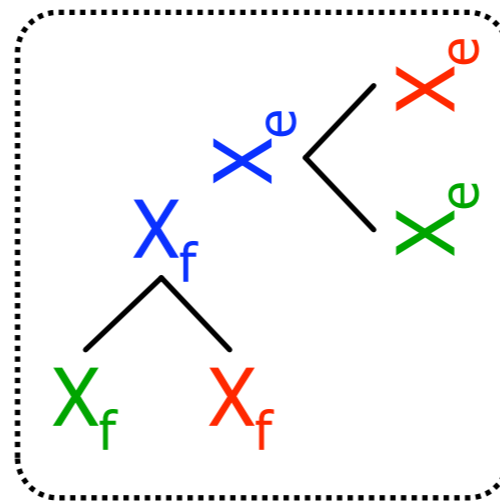
[Wu '97]

Inversion Transduction Grammars

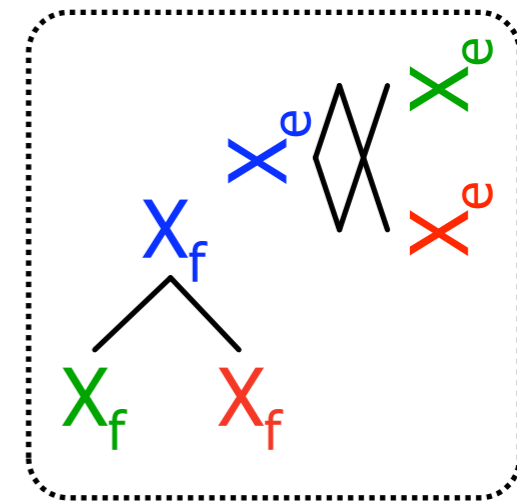
Generation



Monotonic Combination



Inverted Combination



[Wu '97]

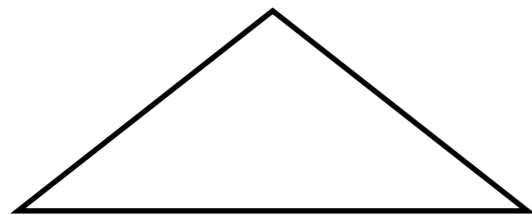


Translating with ITG (w/o LM)

reanudación del período de sesiones

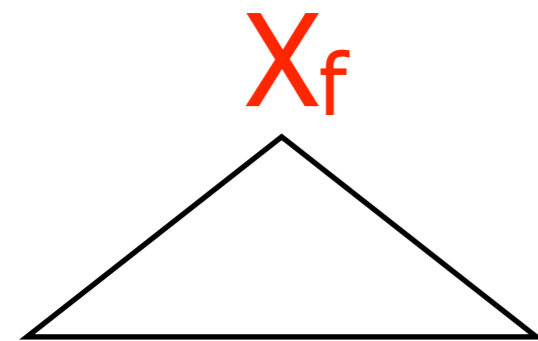
Translating with ITG (w/o LM)

X_f

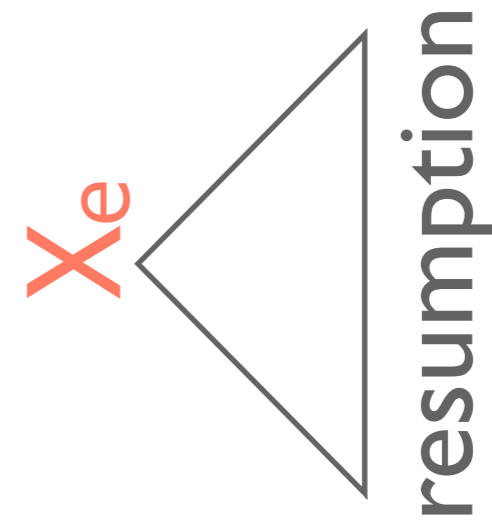


reanudación del período de sesiones

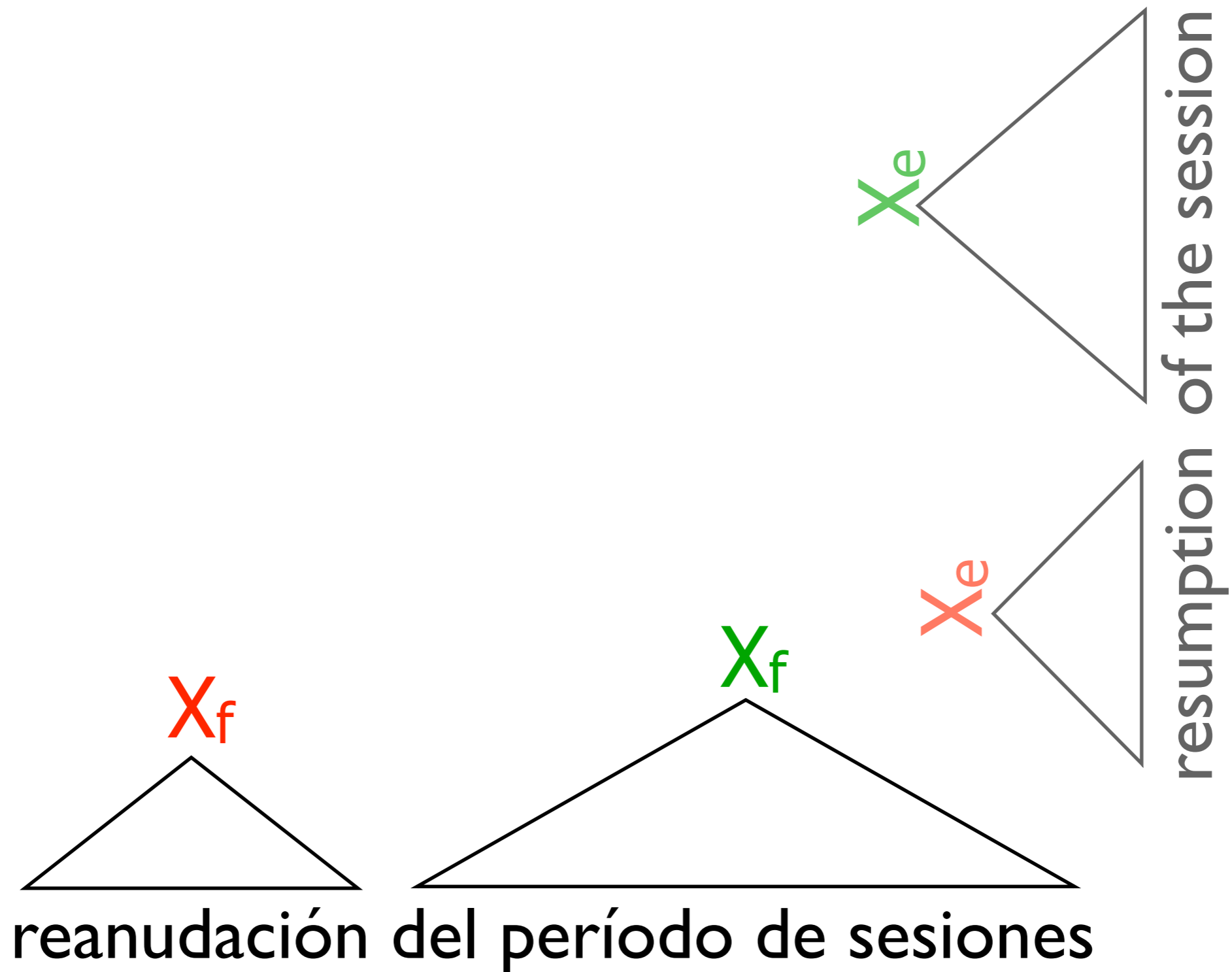
Translating with ITG (w/o LM)



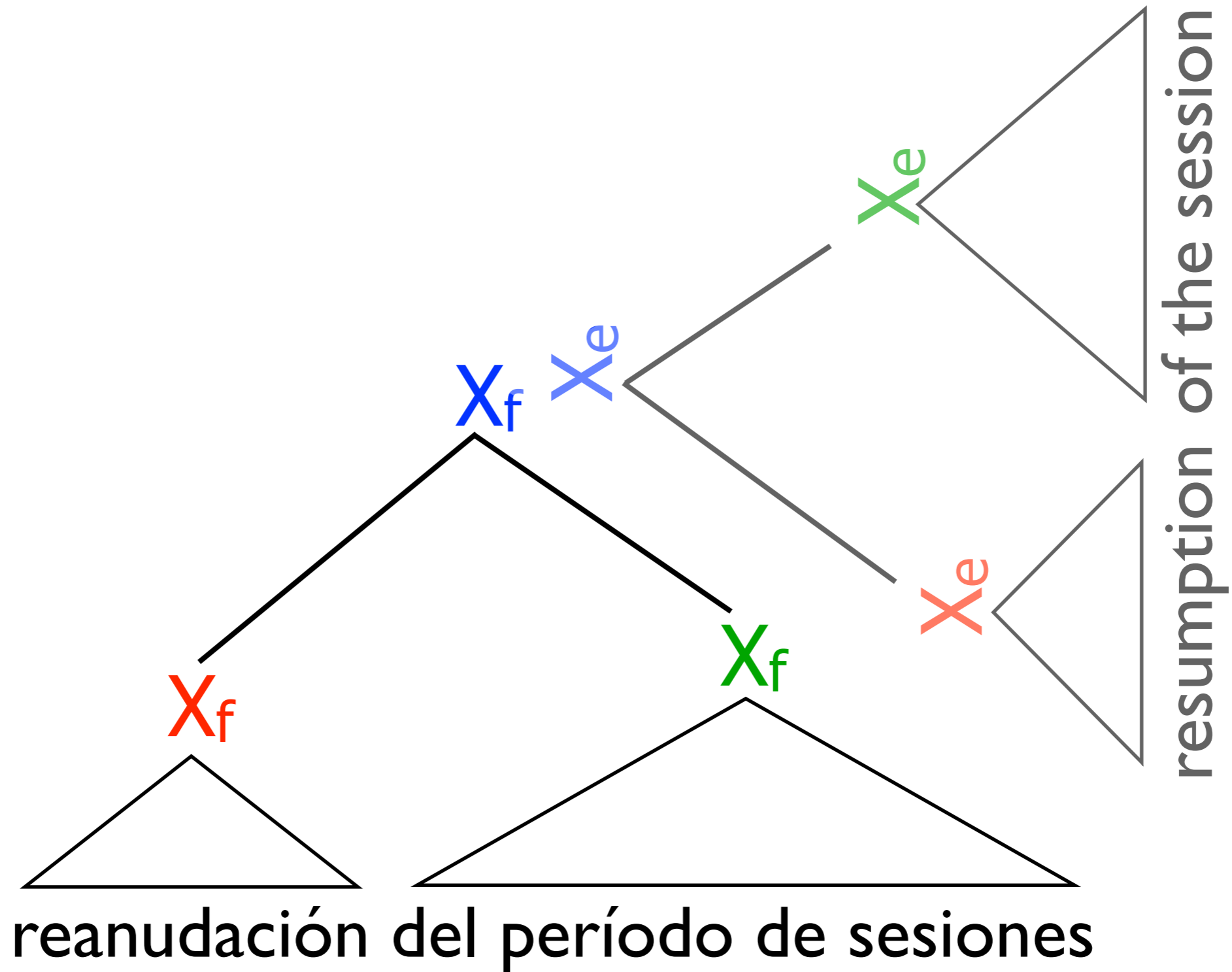
reanudación del período de sesiones



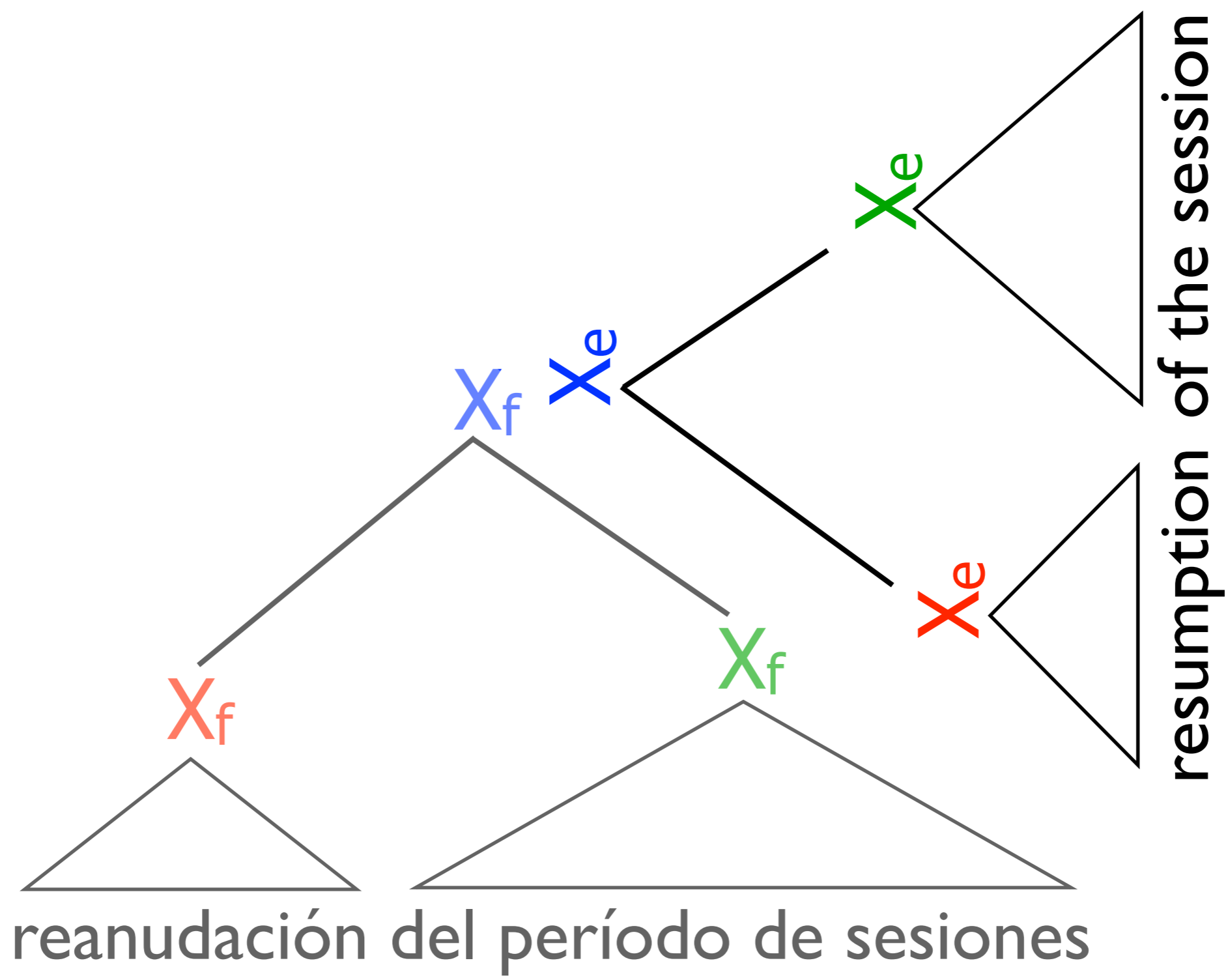
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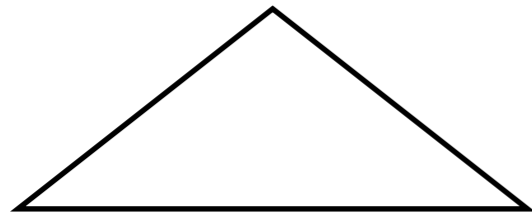


Syntactic MT w/ Language Model

reanudación del período de sesiones

Syntactic MT w/ Language Model

resumption- X_f -resumption



reanudación del período de sesiones

resumption

Syntactic MT w/ Language Model

resumption- X_f -resumption of- X_f -session

reanudación del período de sesiones

resumption of the session

Syntactic MT w/ Language Model

Score with
Language Model

resumption- X_f -resumption of- X_f -session

reanudación del período de sesiones

resumption of the session

Syntactic MT w/ Language Model

resumption- X_f -resumption of- X_f -session

reanudación del período de sesiones

resumption of the session

Syntactic MT w/ Language Model

resumption- X_f -session

reanudación del período de sesiones

resumption of the session

Syntactic MT w/ Language Model

Lexical Explosion

resumption, of- X_f -the, session

reanudación del período de sesiones

resumption of the session

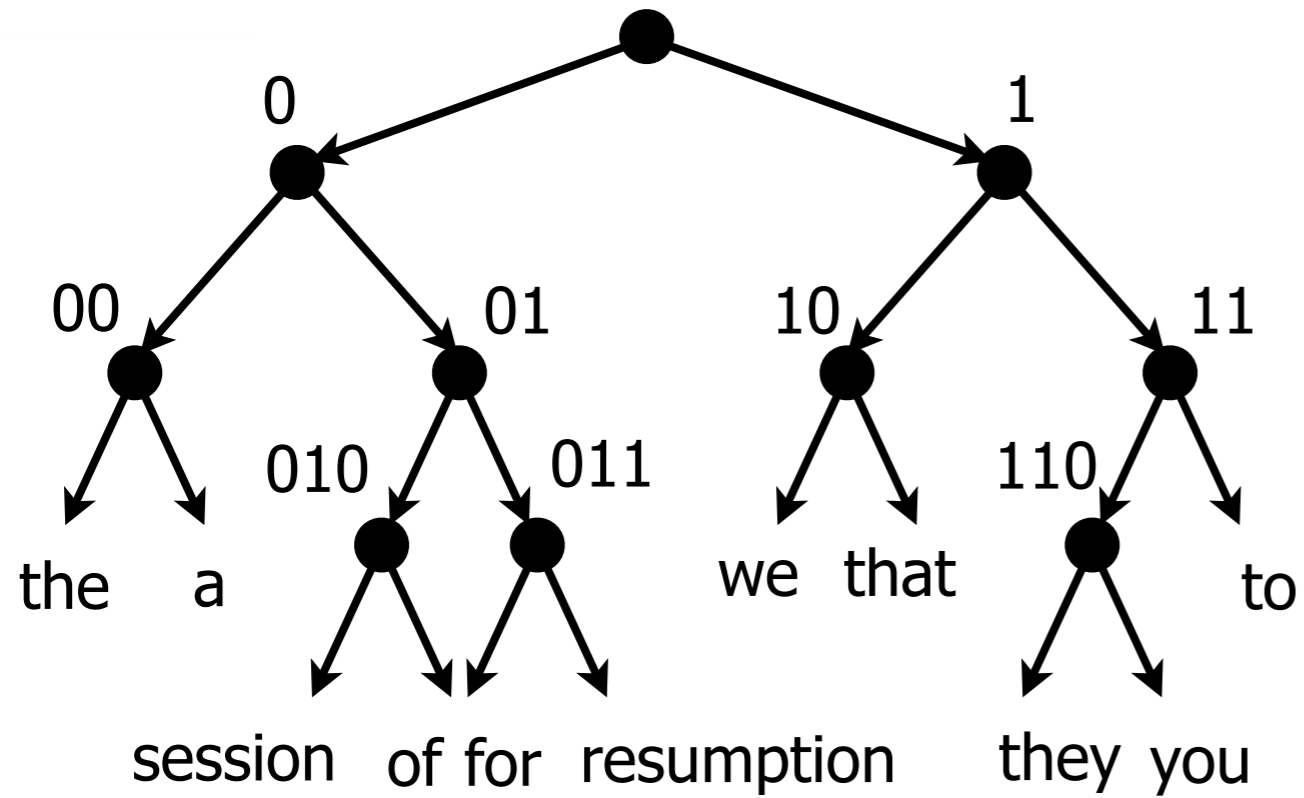


State Space Projections



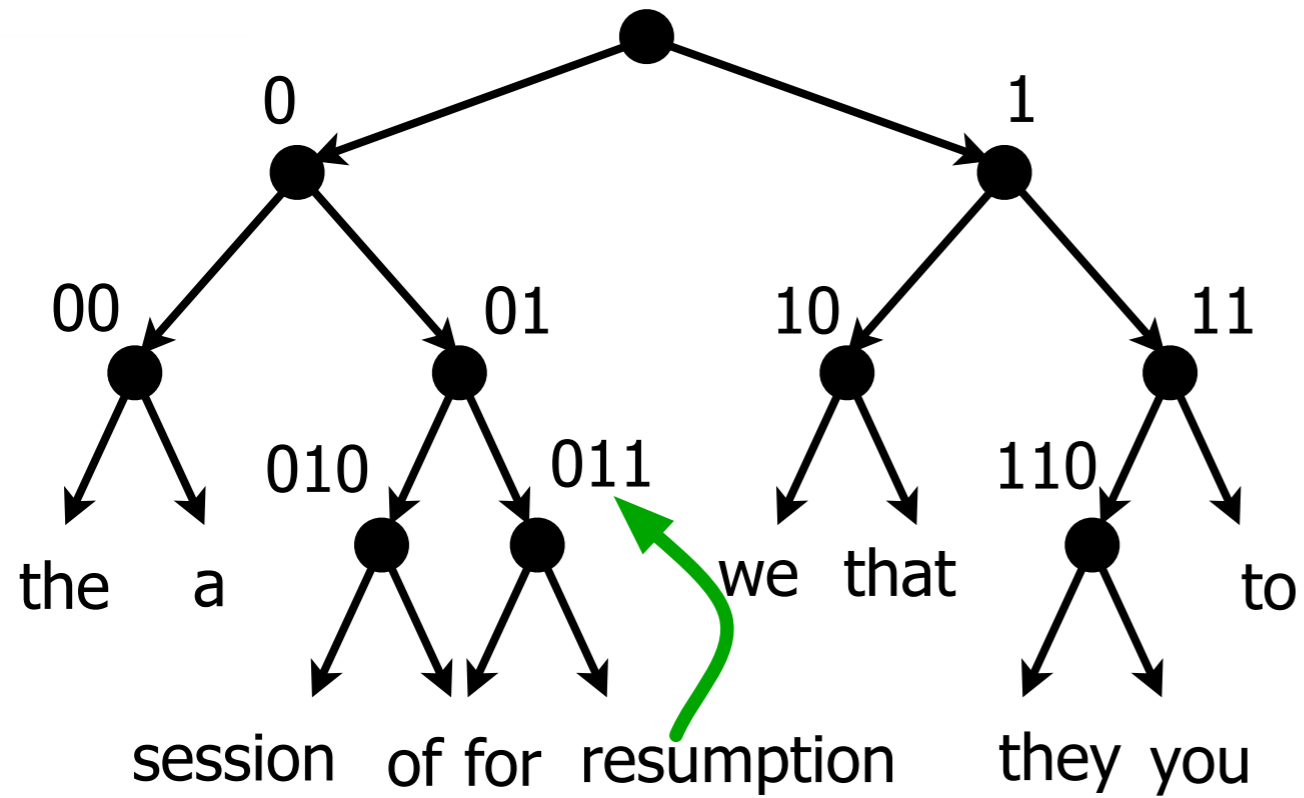
resumption-X-session

State Space Projections



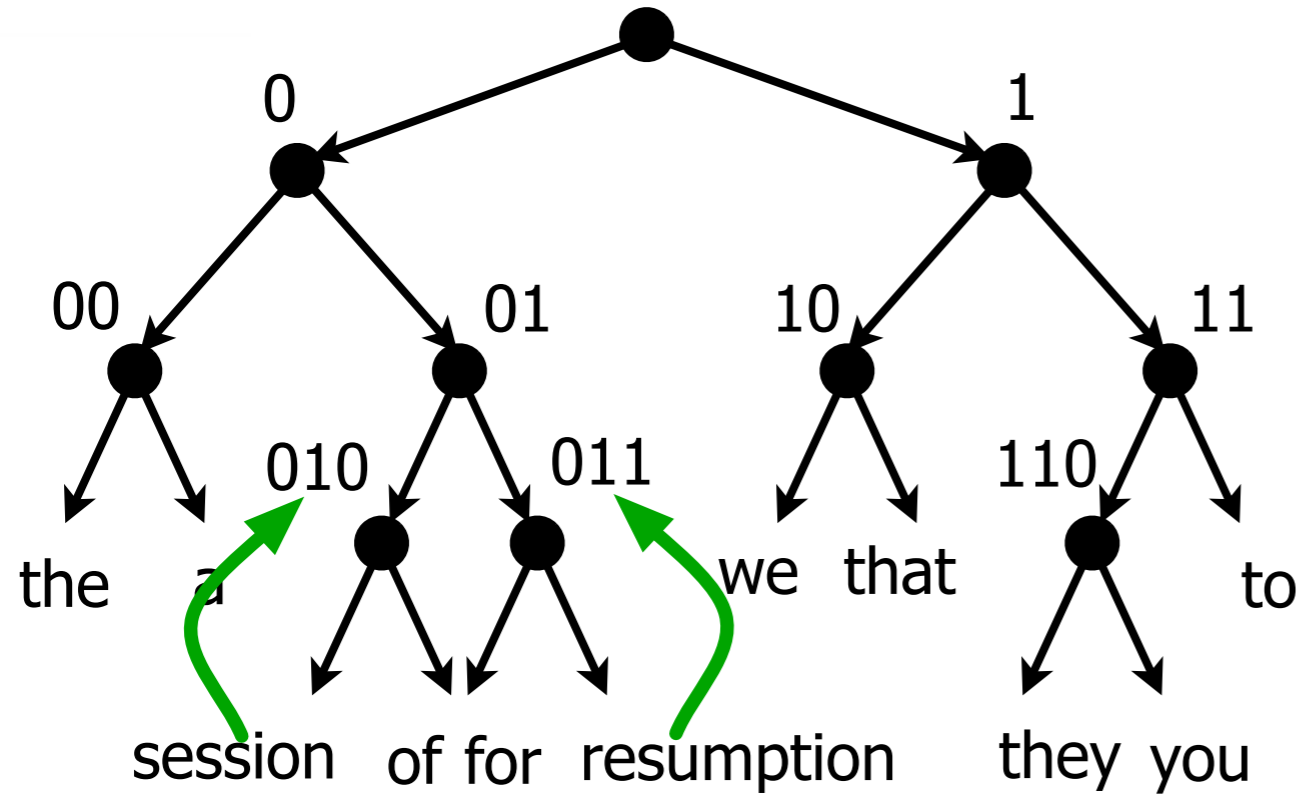
●
resumption-X-session

State Space Projections



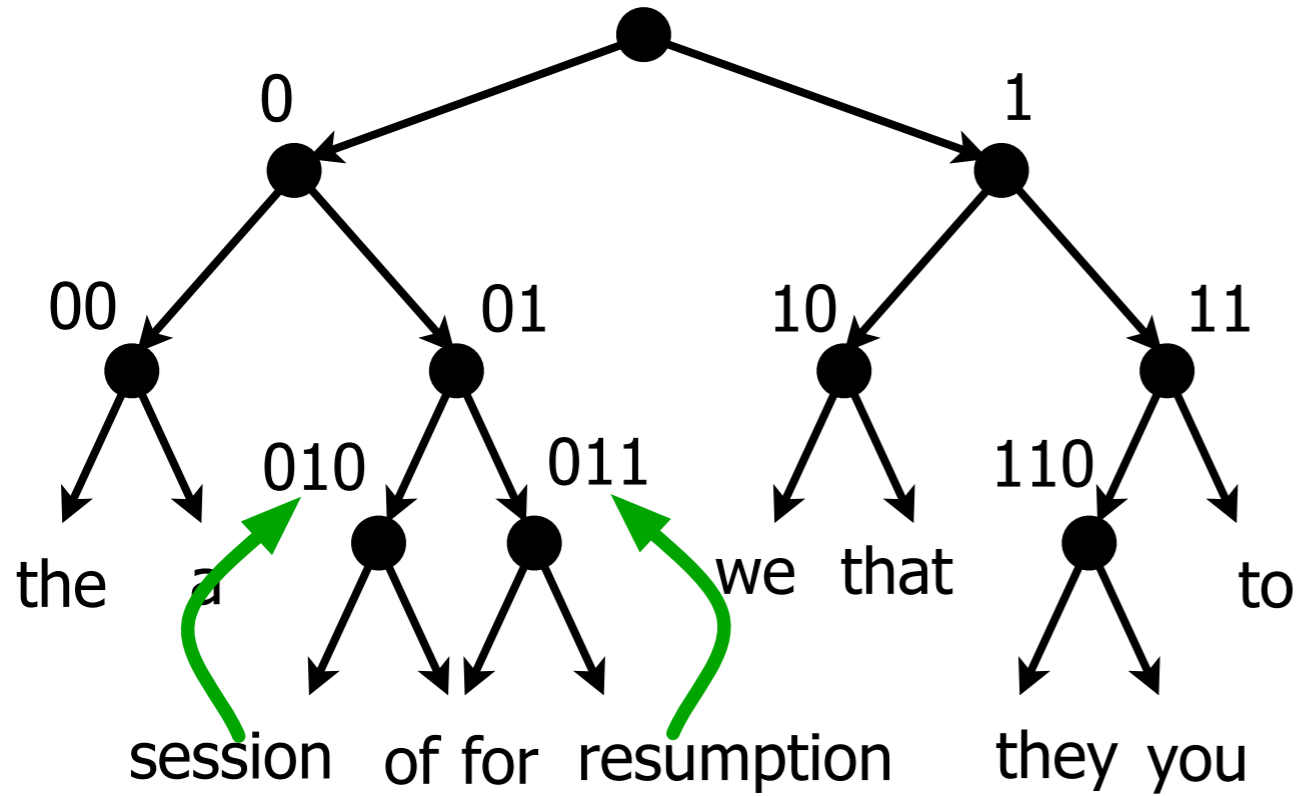
●
resumption-X-session

State Space Projections



●
 resumption-X-session

State Space Projections



●
-X-

●
resumption-X-session

State Space Projections





State Space Projections



011-X-010



resumption-X-session

State Space Projections

●
01-X-01

●
011-X-010

●
resumption-X-session

State Space Projections



0-X-0



01-X-01

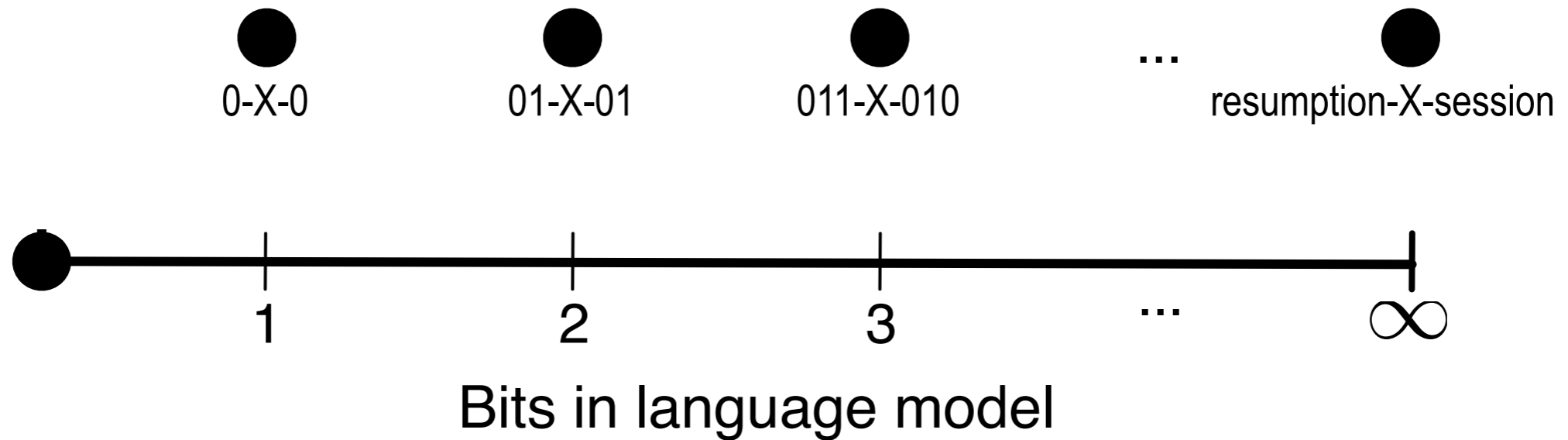


011-X-010

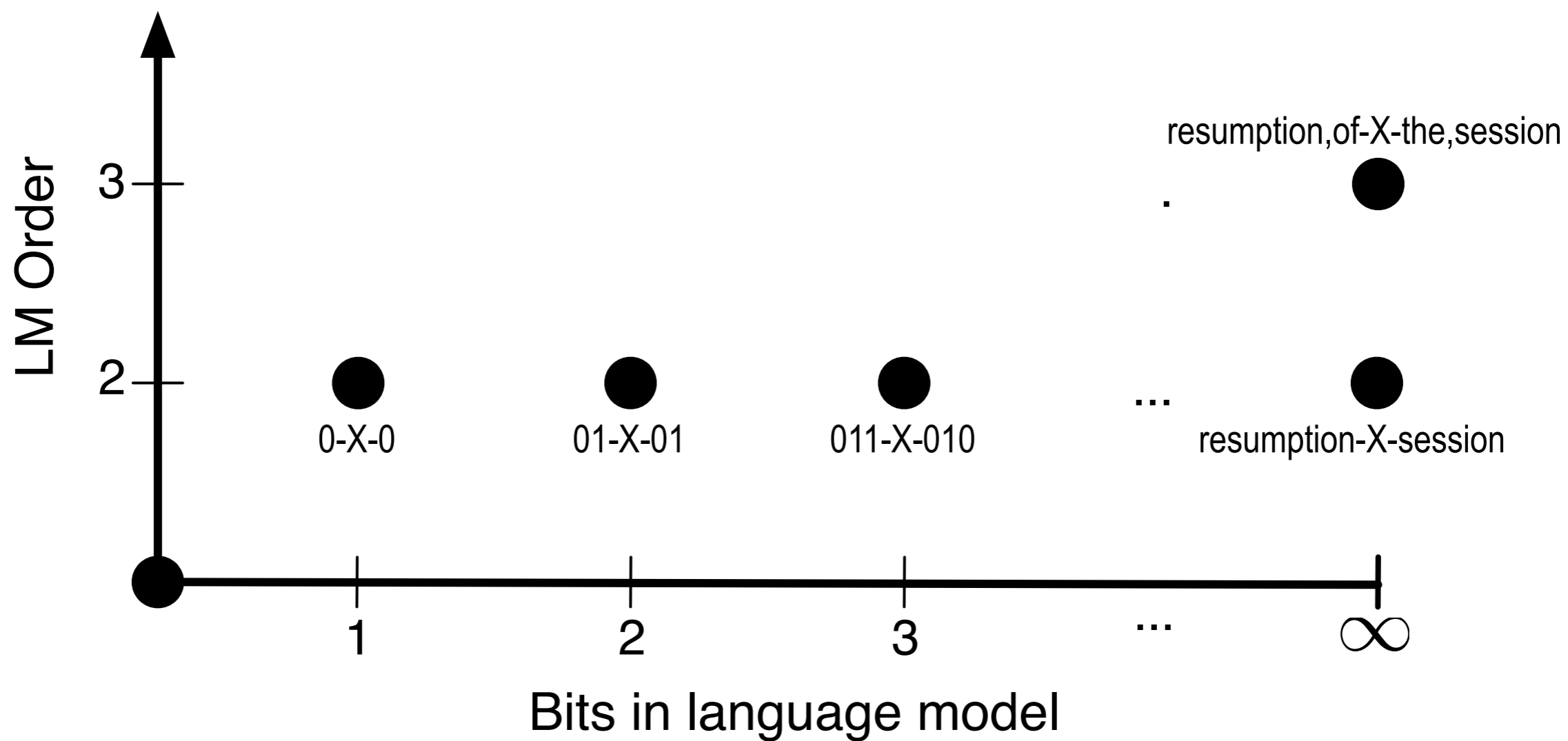


resumption-X-session

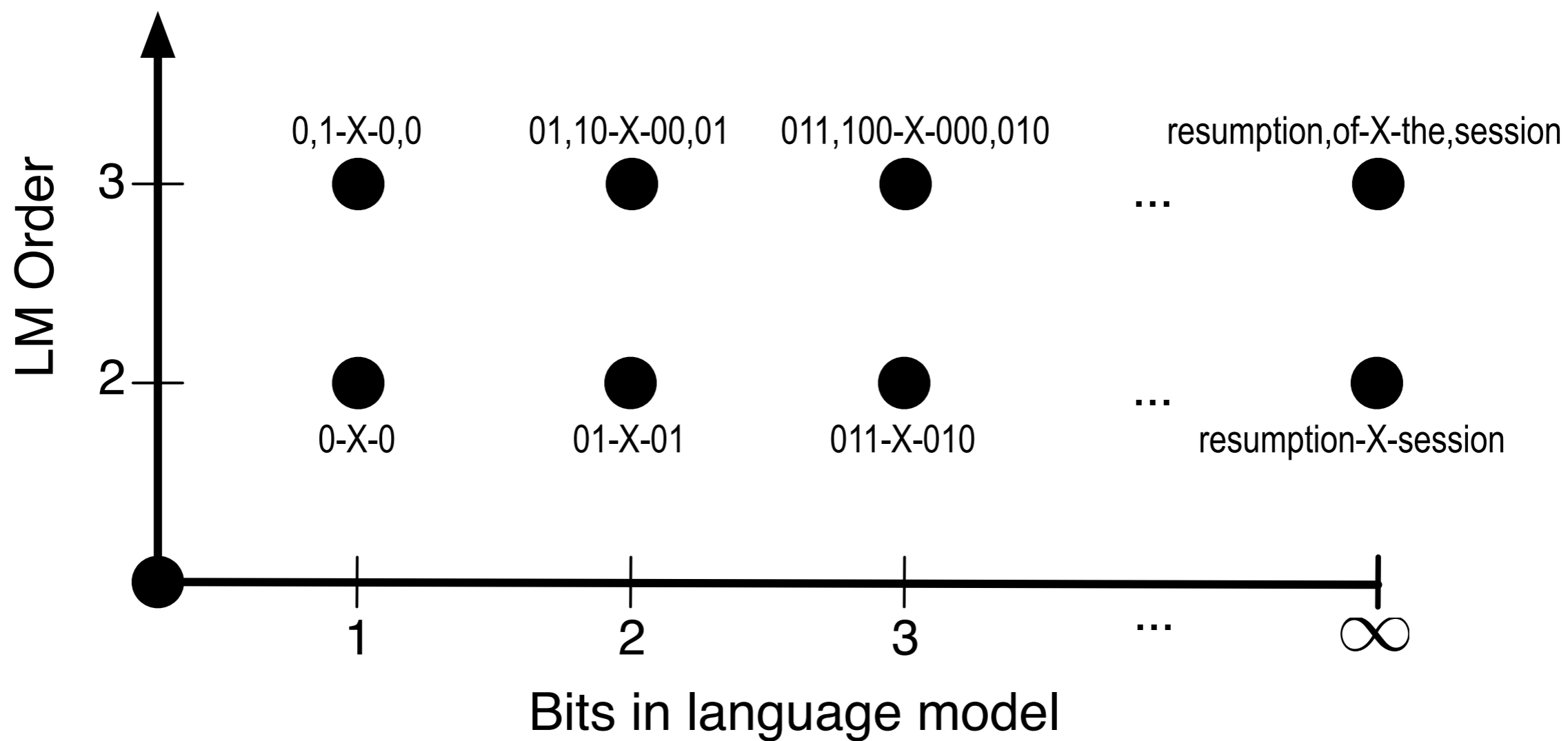
State Space Projections



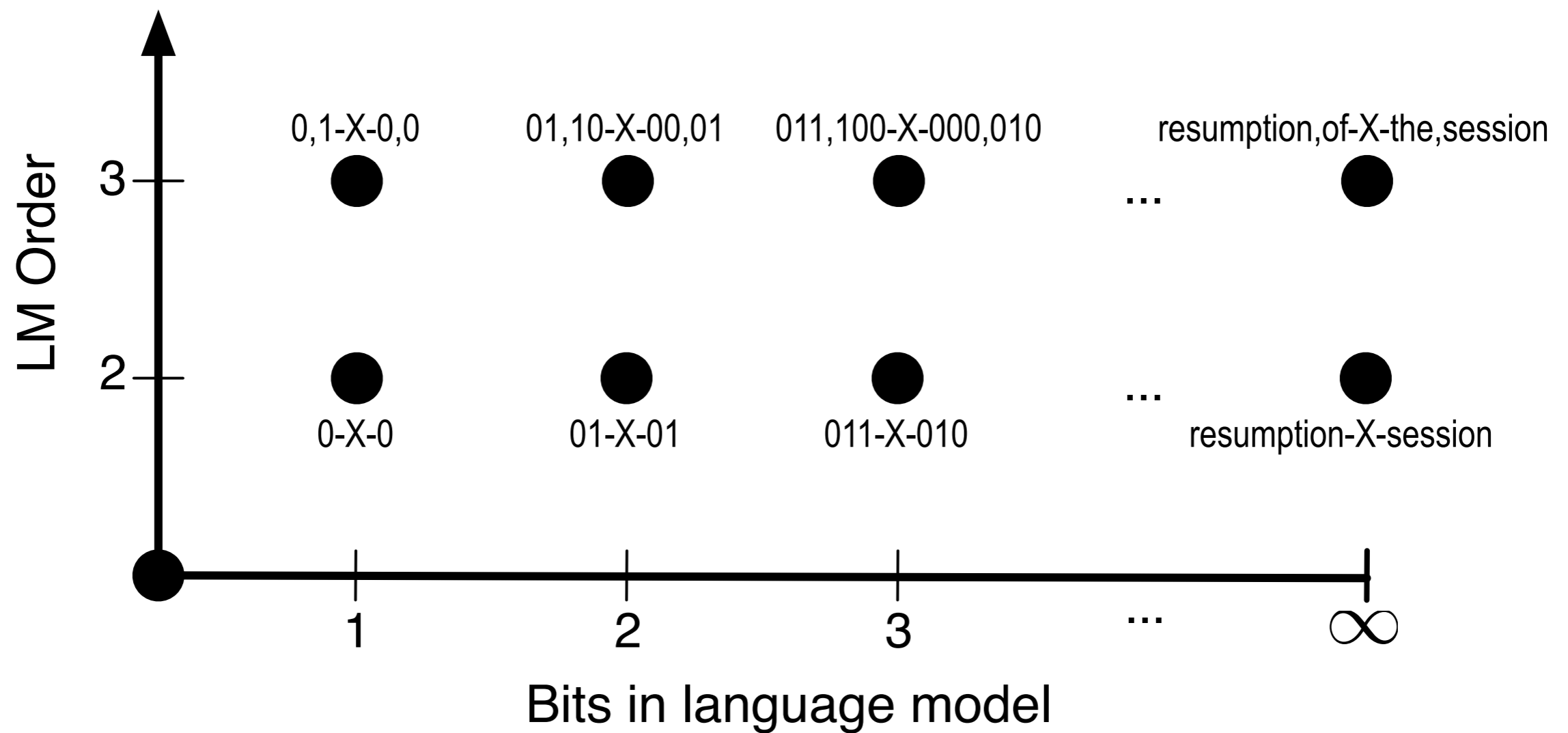
State Space Projections



State Space Projections

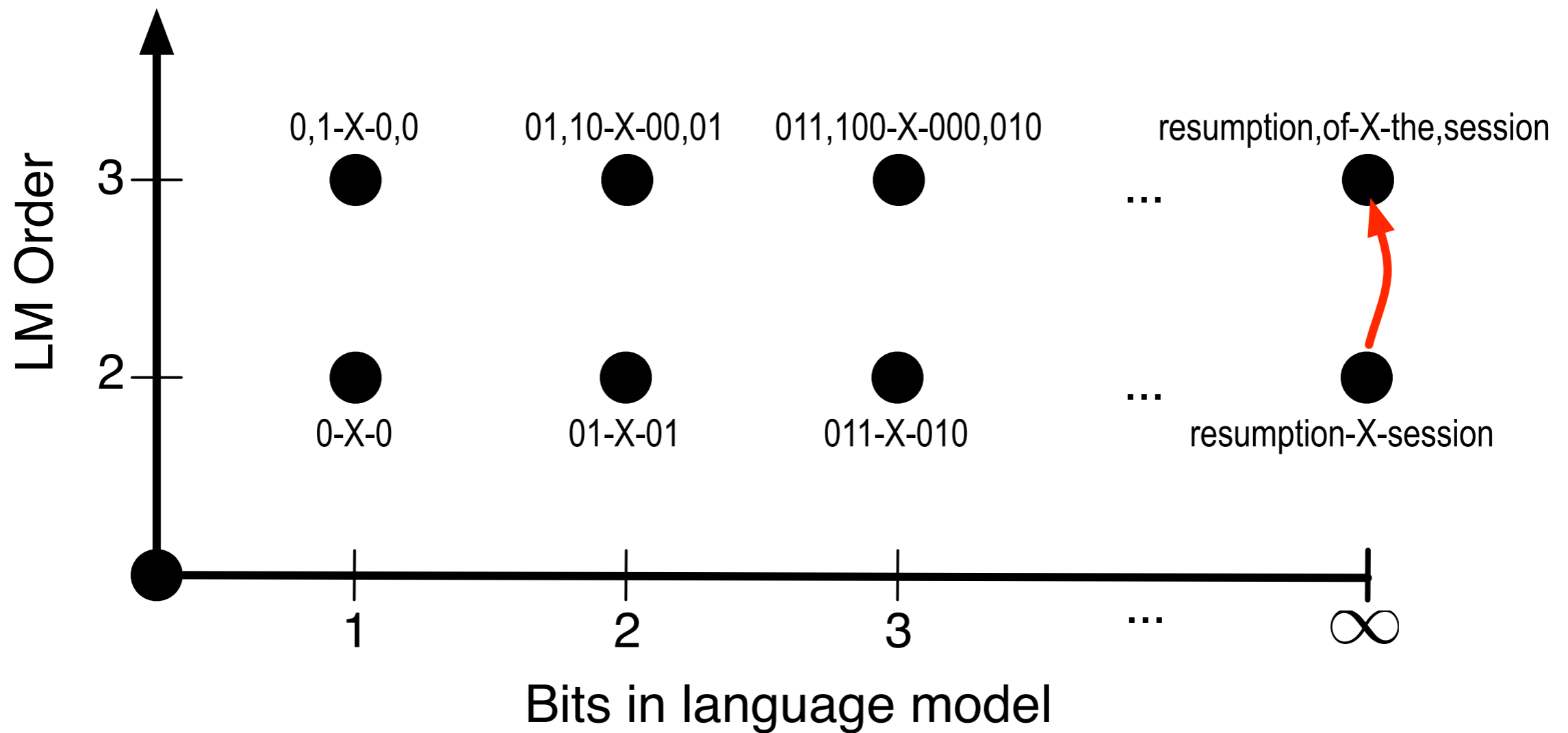


Coarse-to-Fine Inference

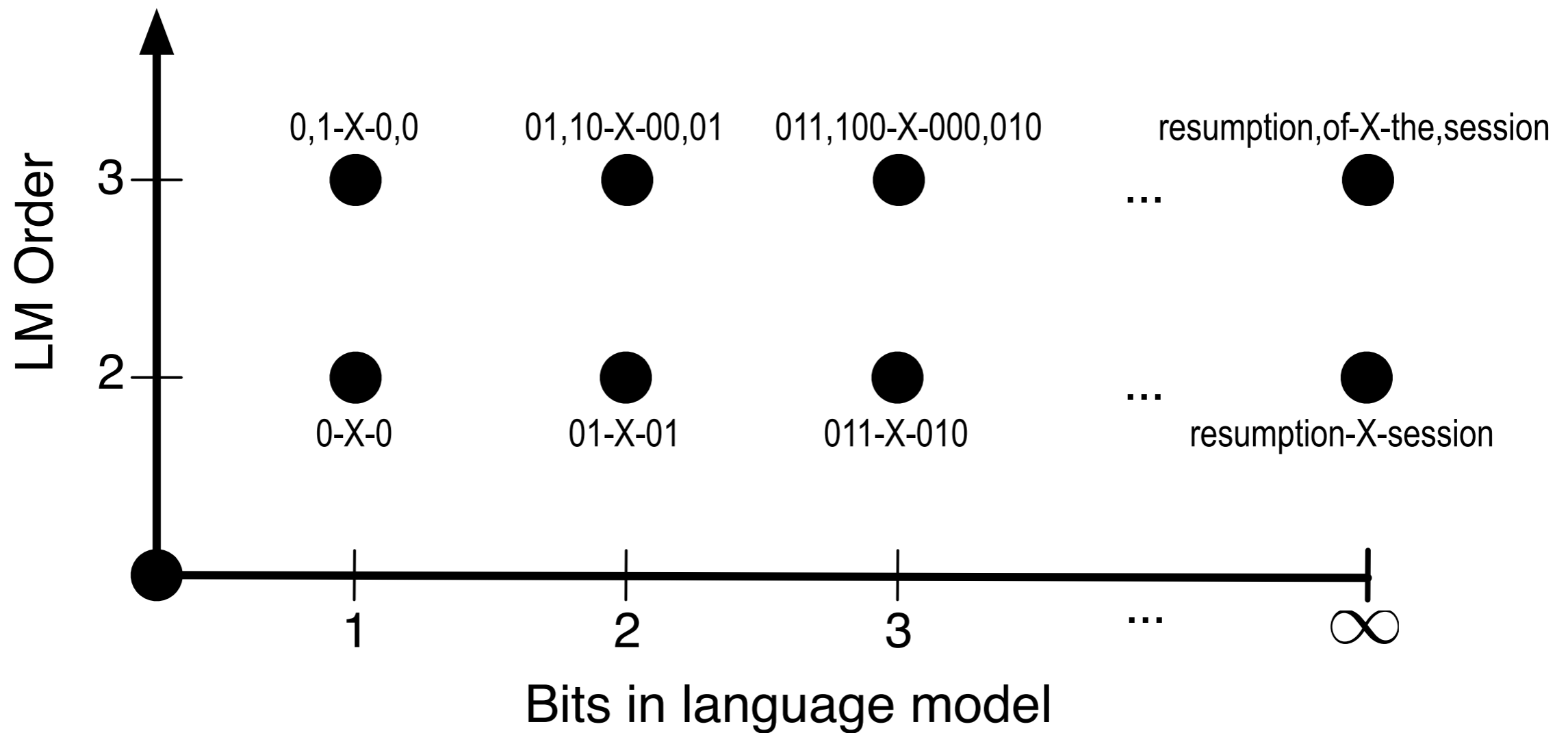


Coarse-to-Fine Inference

Order based [Zhang & Gildea '08]

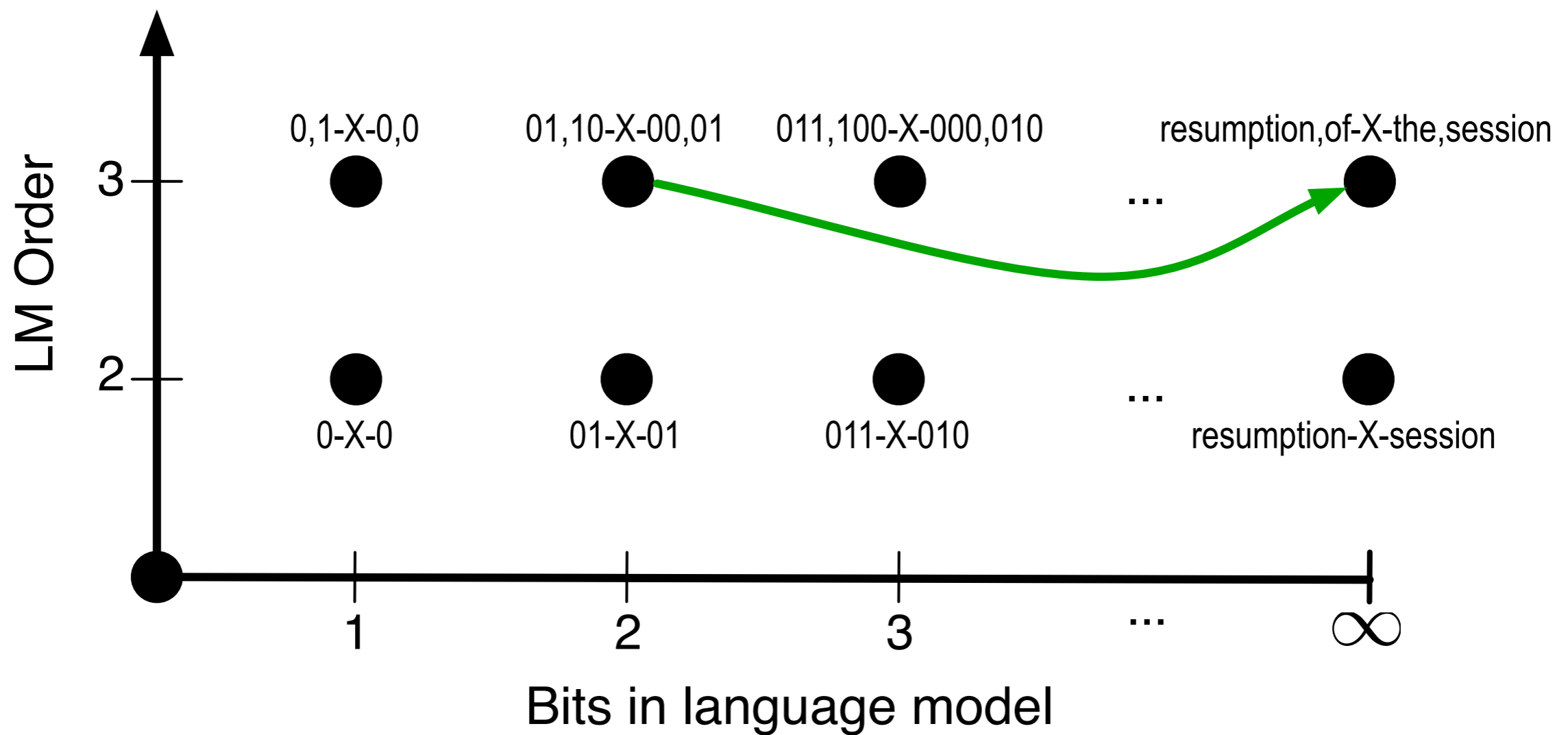


Coarse-to-Fine Inference

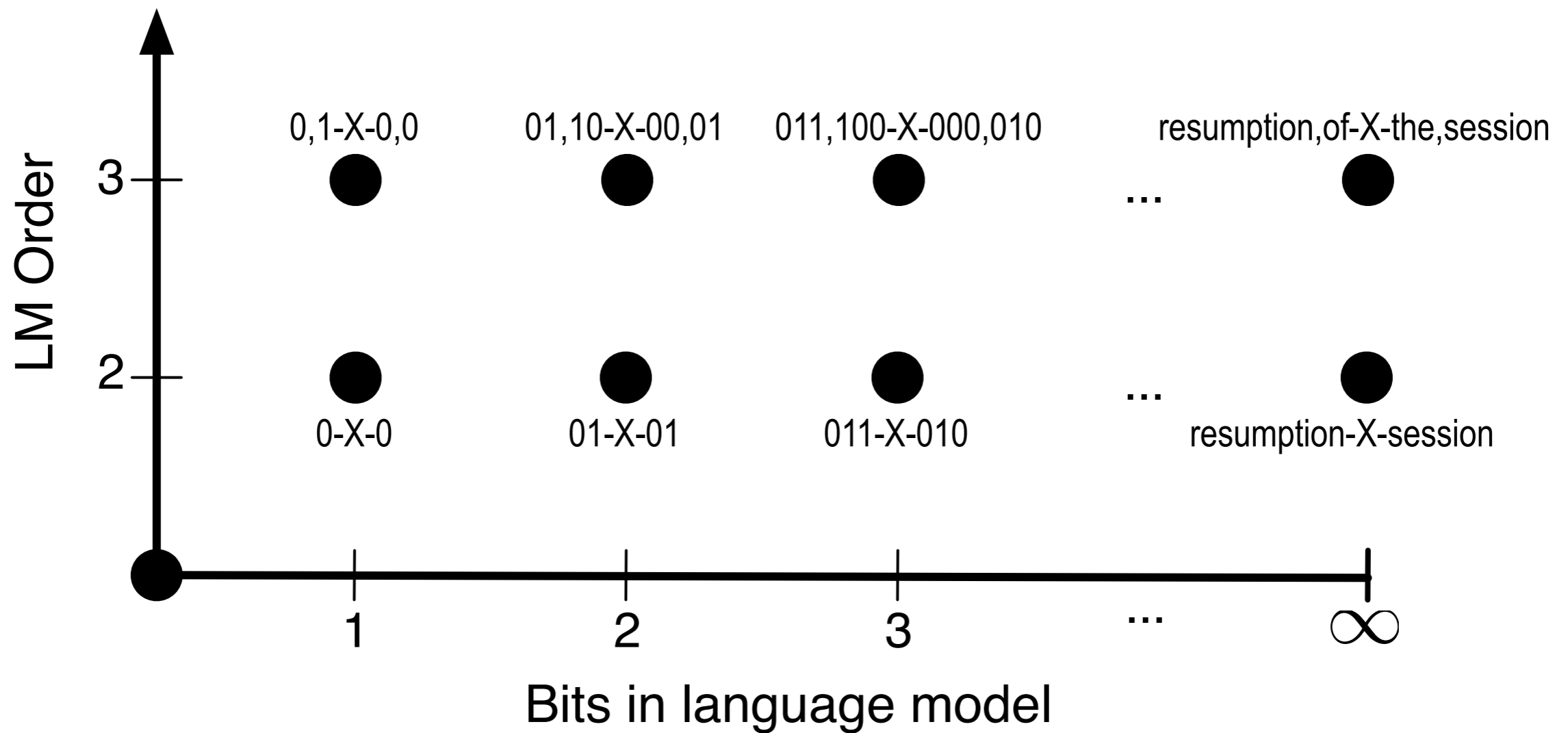


Coarse-to-Fine Inference

Encoding based

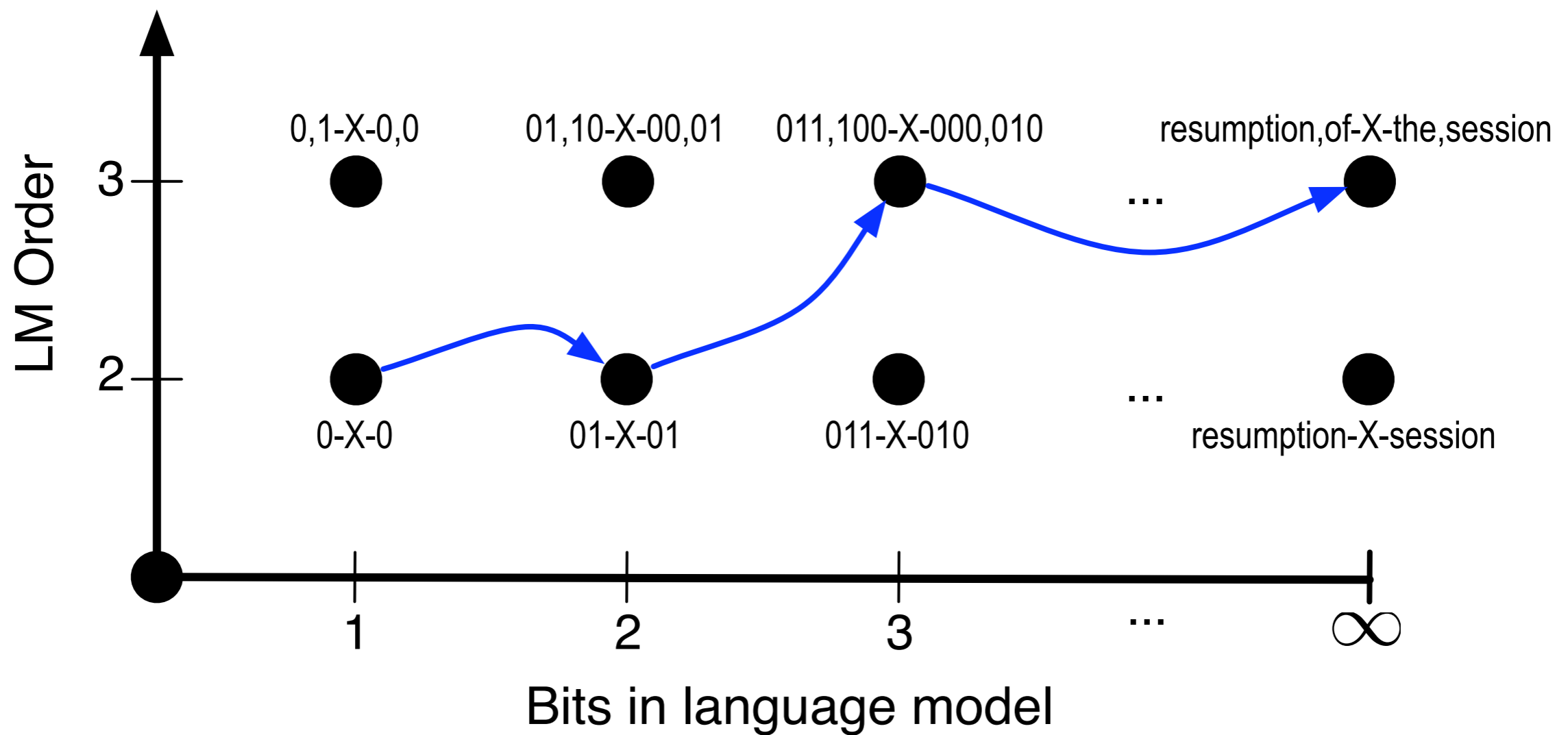


Coarse-to-Fine Inference



Coarse-to-Fine Inference

Combine both?



Experimental Setup

- **Europarl corpus:**
 - Spanish, German, French → English
 - About 8M words of parallel text
 - Average sentence length 30
 - 2K development sentences
- **Pre-Processing**
 - Standard GIZA++ word alignments and phrase extraction
 - SRILM trigram language model
- **ITG translation model**
 - Extracted phrases / scores used as ITG terminal productions
 - Simplistic distortion model: 1 monotonic and 1 inverted rule
 - Less than 1 BLEU point below Moses on same setup

Coarse-to-Fine Pruning

For each span:

Coarse Pass:

--	--	--	--

Coarse-to-Fine Pruning

For each span:

Coarse Pass:

0-X-0	0-X-1	1-X-0	1-X-1
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Coarse-to-Fine Pruning

For each span:

Coarse Pass:

Posteriors

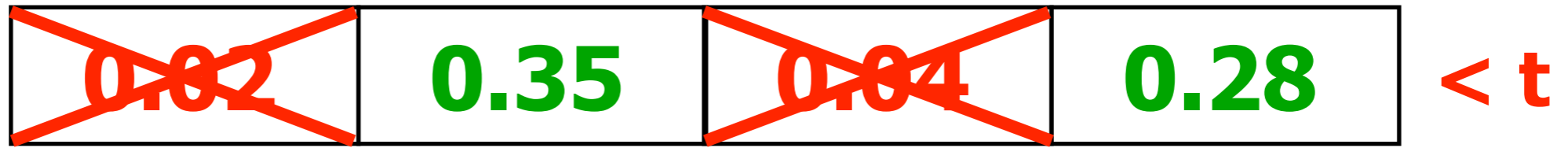
0.02	0.35	0.04	0.28
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Coarse-to-Fine Pruning

For each span:

Coarse Pass:

Posteriors

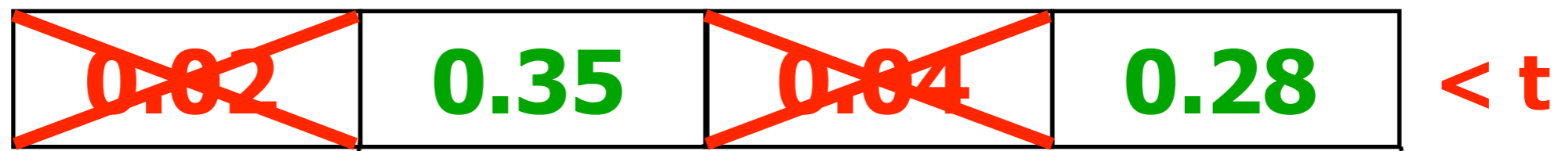


Coarse-to-Fine Pruning

For each span:

Coarse Pass:

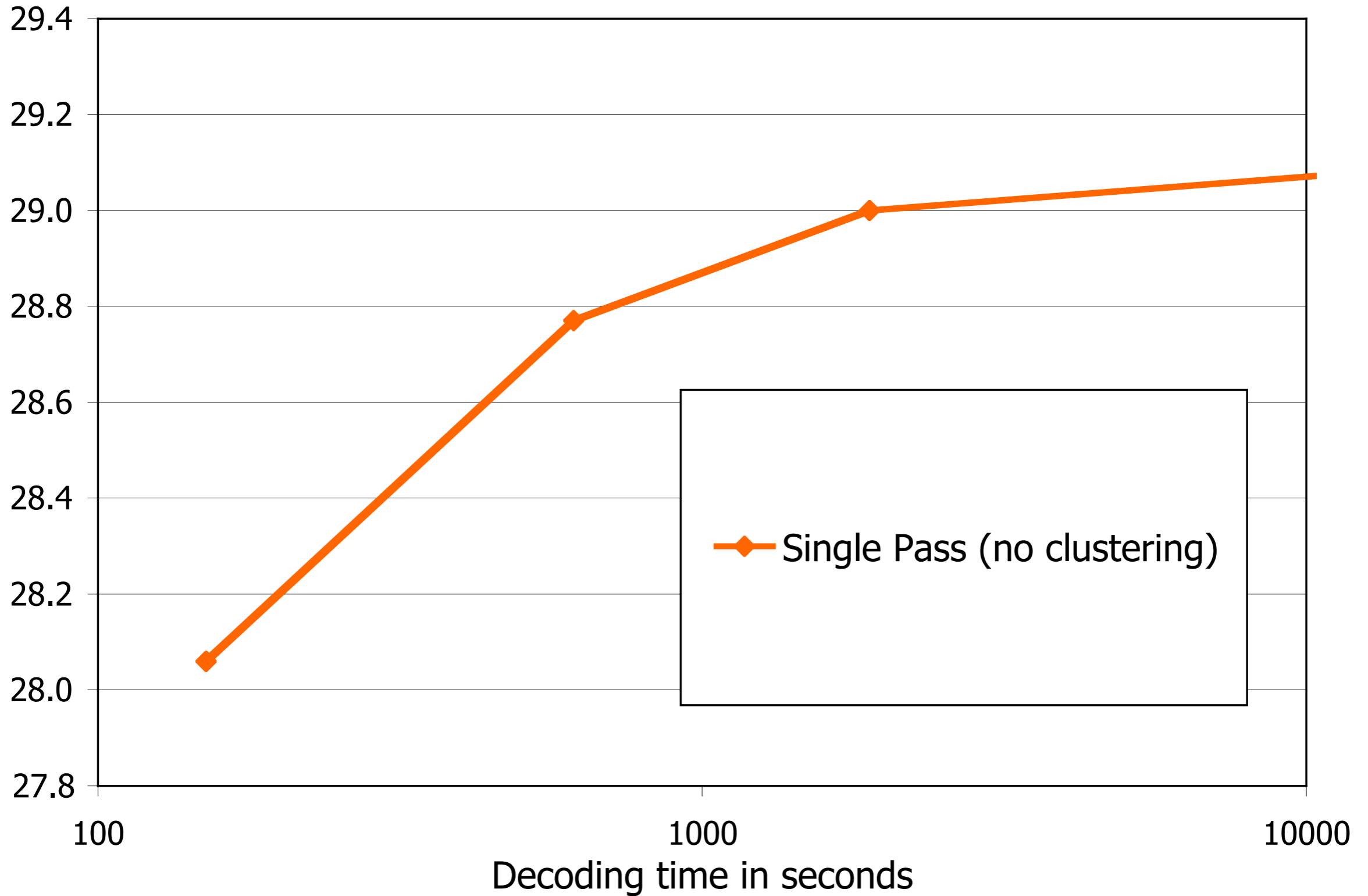
Posteriors



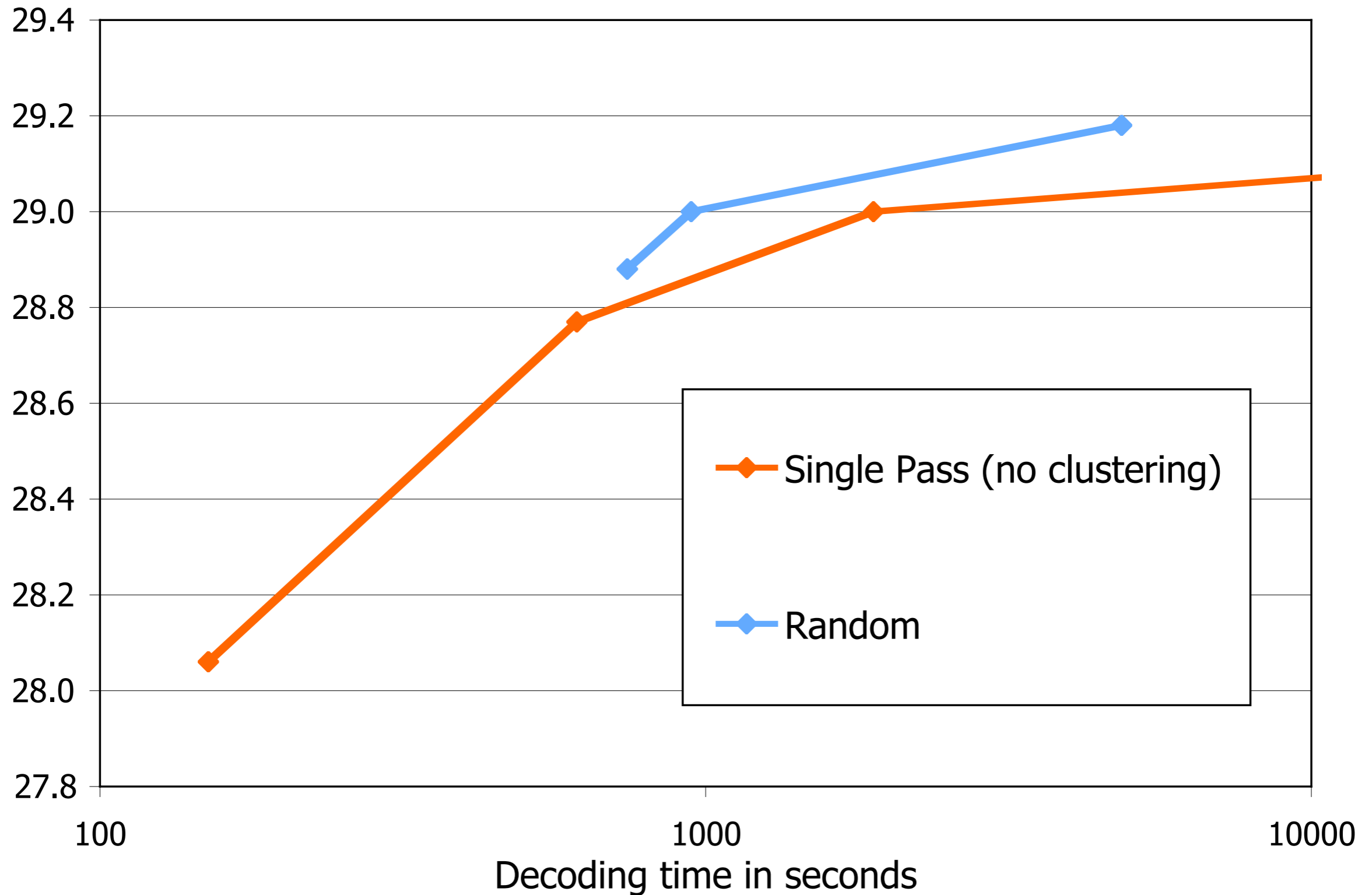
Finer Pass:



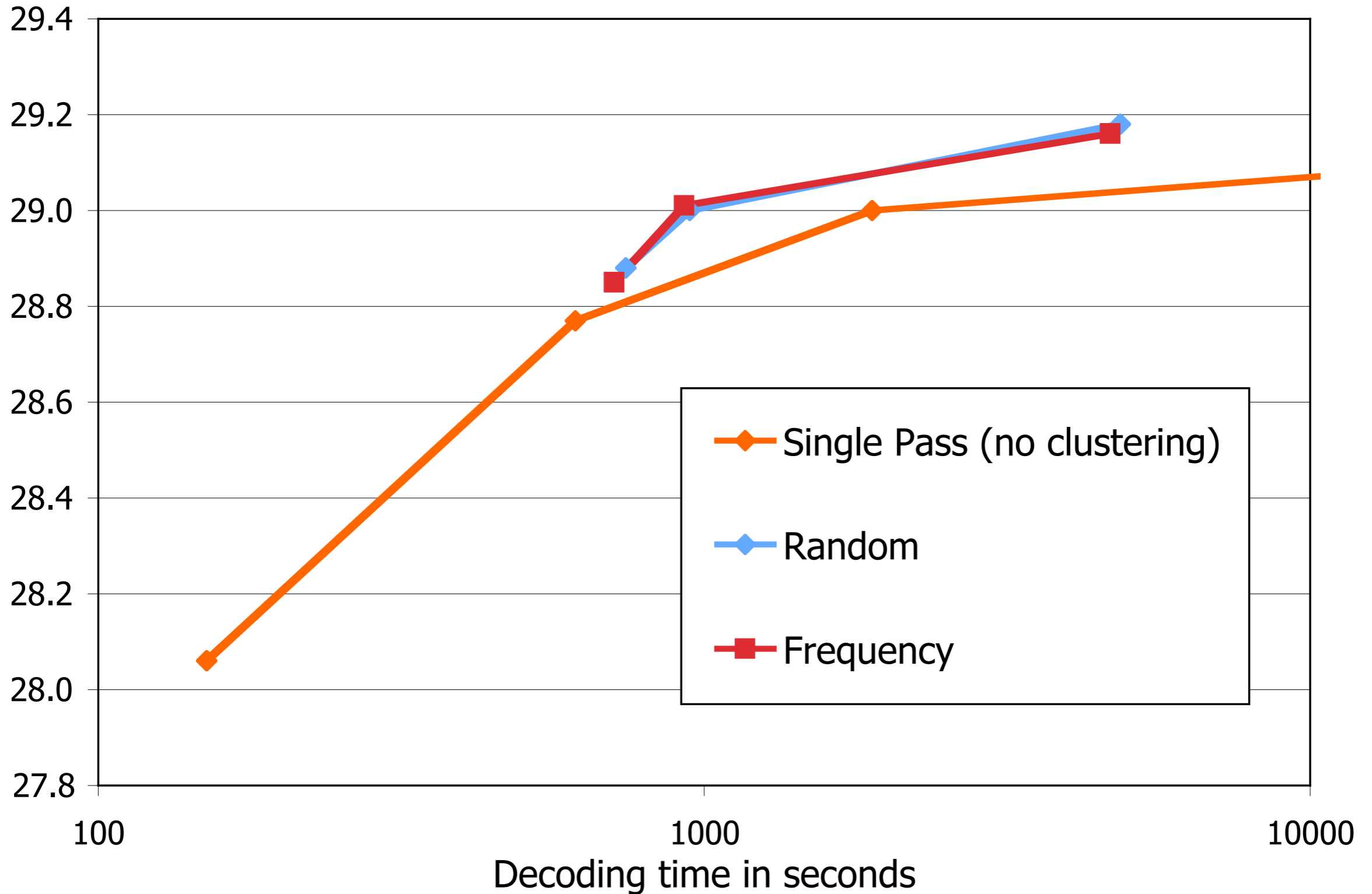
Cluster Quality (BLEU)



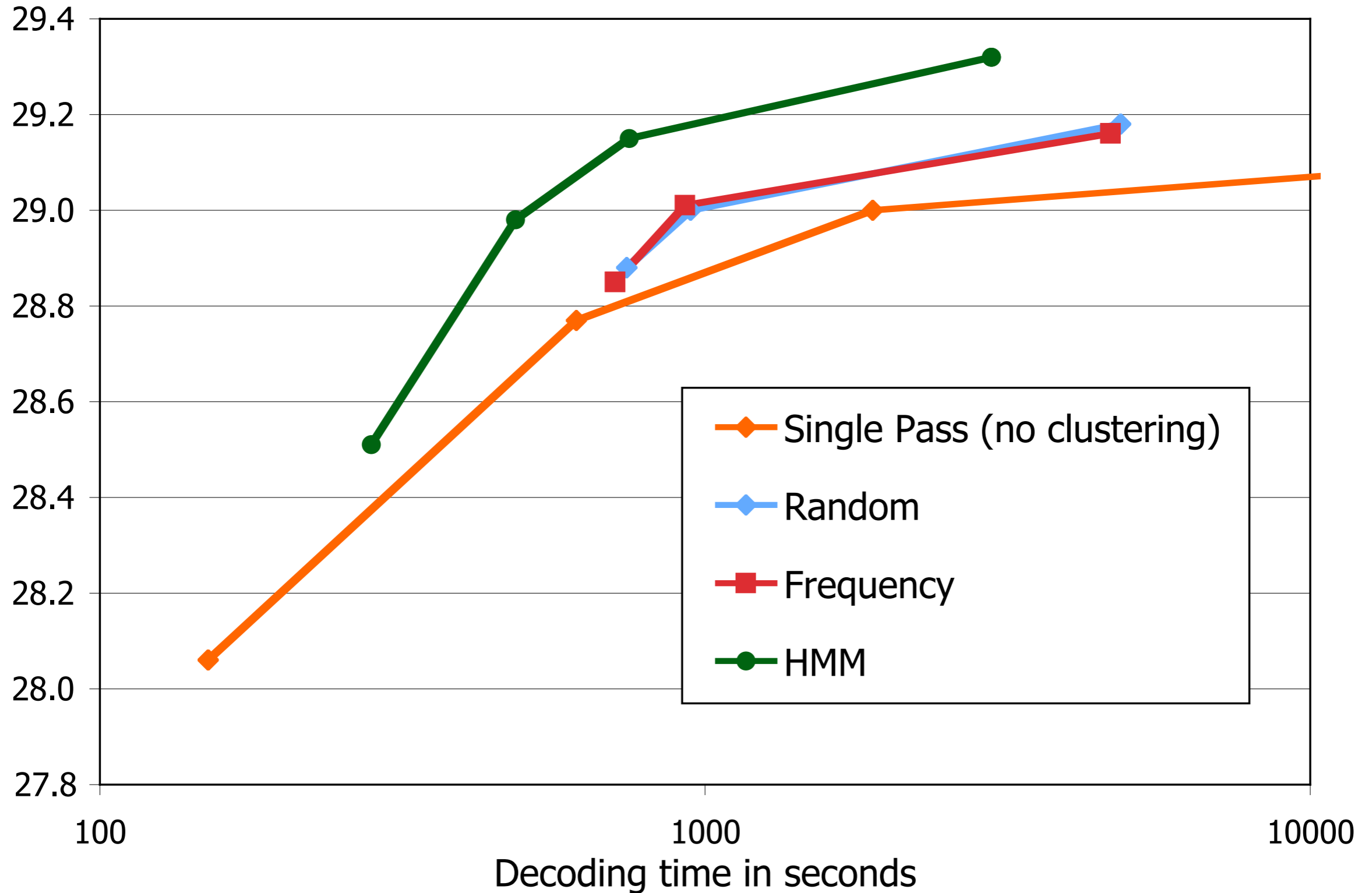
Cluster Quality (BLEU)



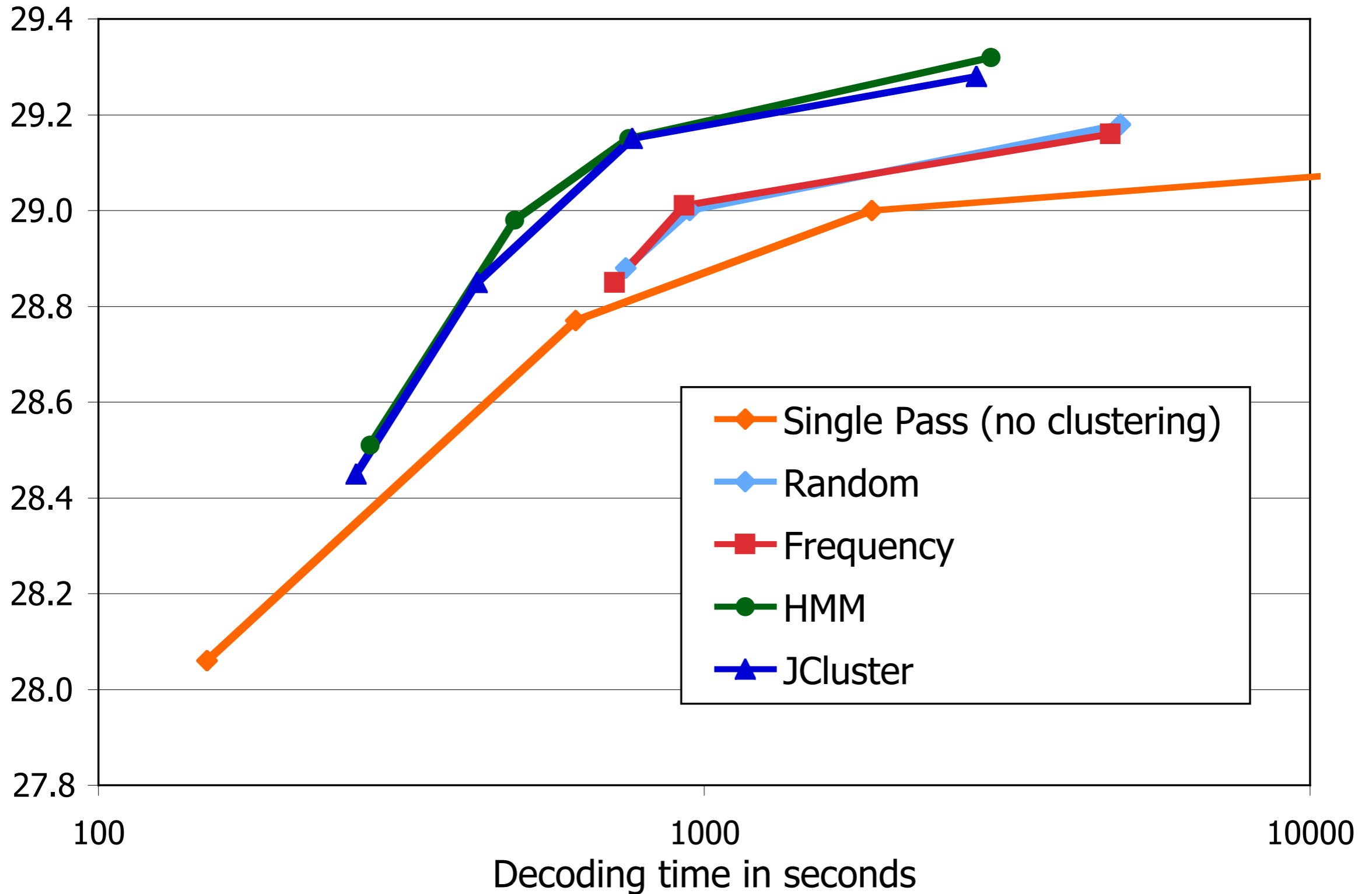
Cluster Quality (BLEU)



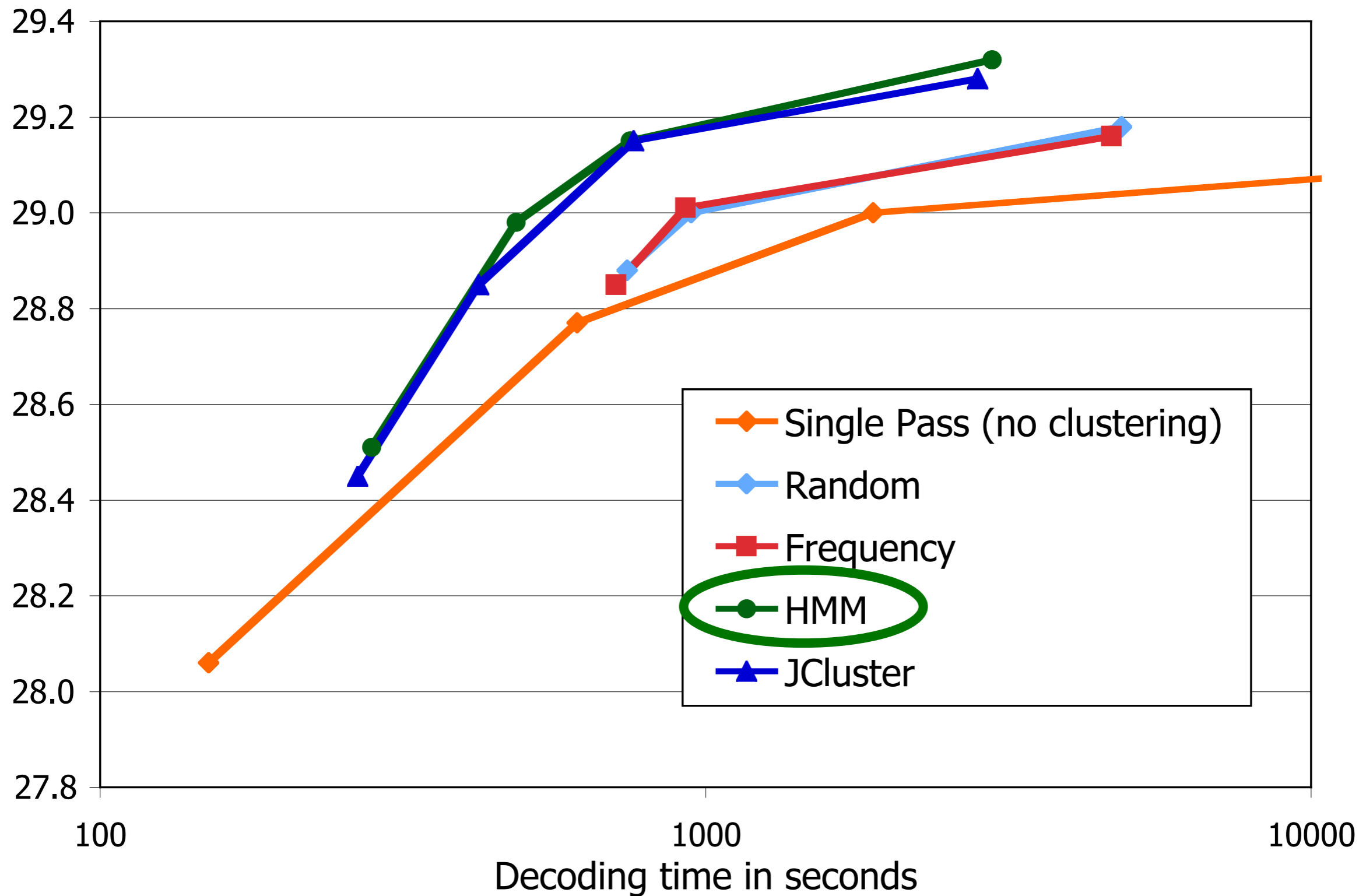
Cluster Quality (BLEU)



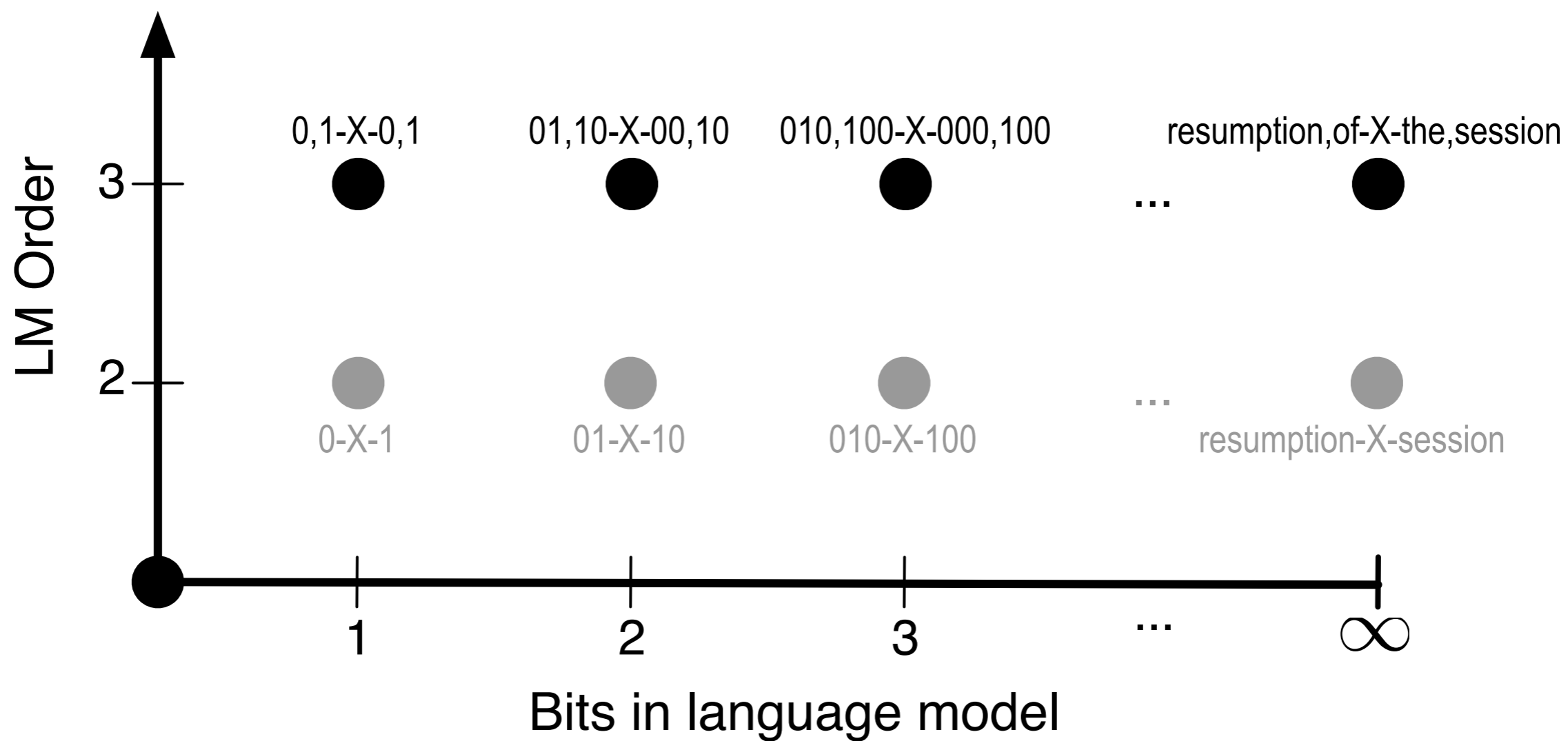
Cluster Quality (BLEU)



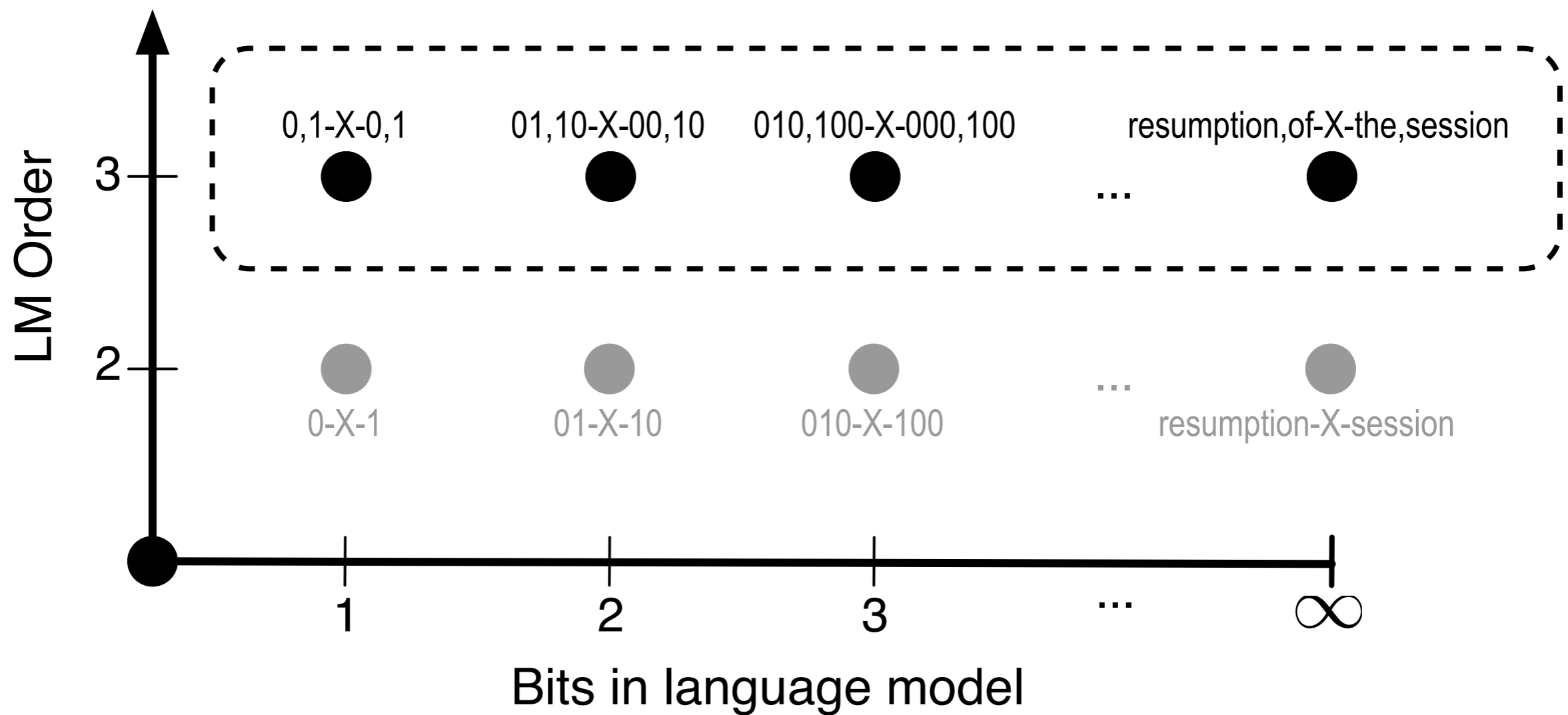
Cluster Quality (BLEU)



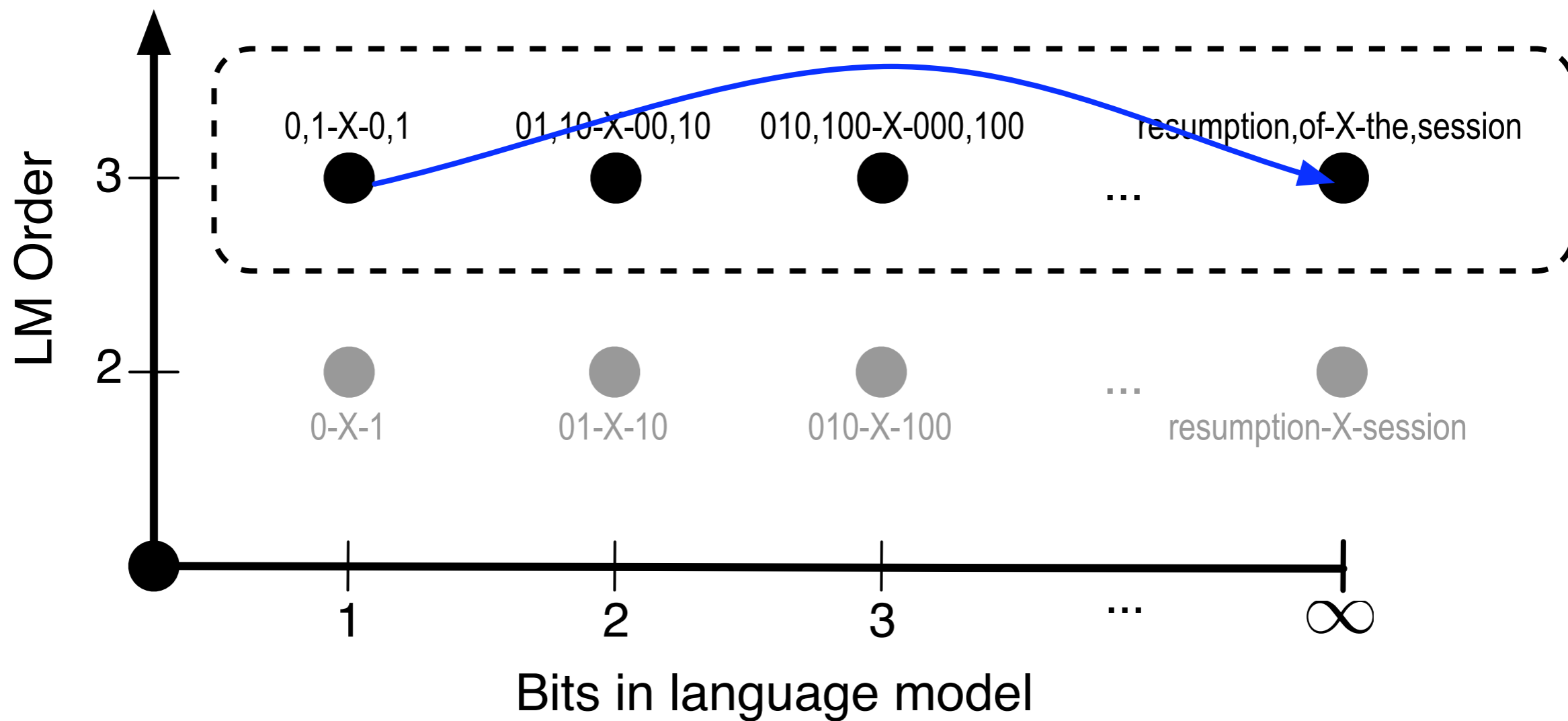
Where to start?



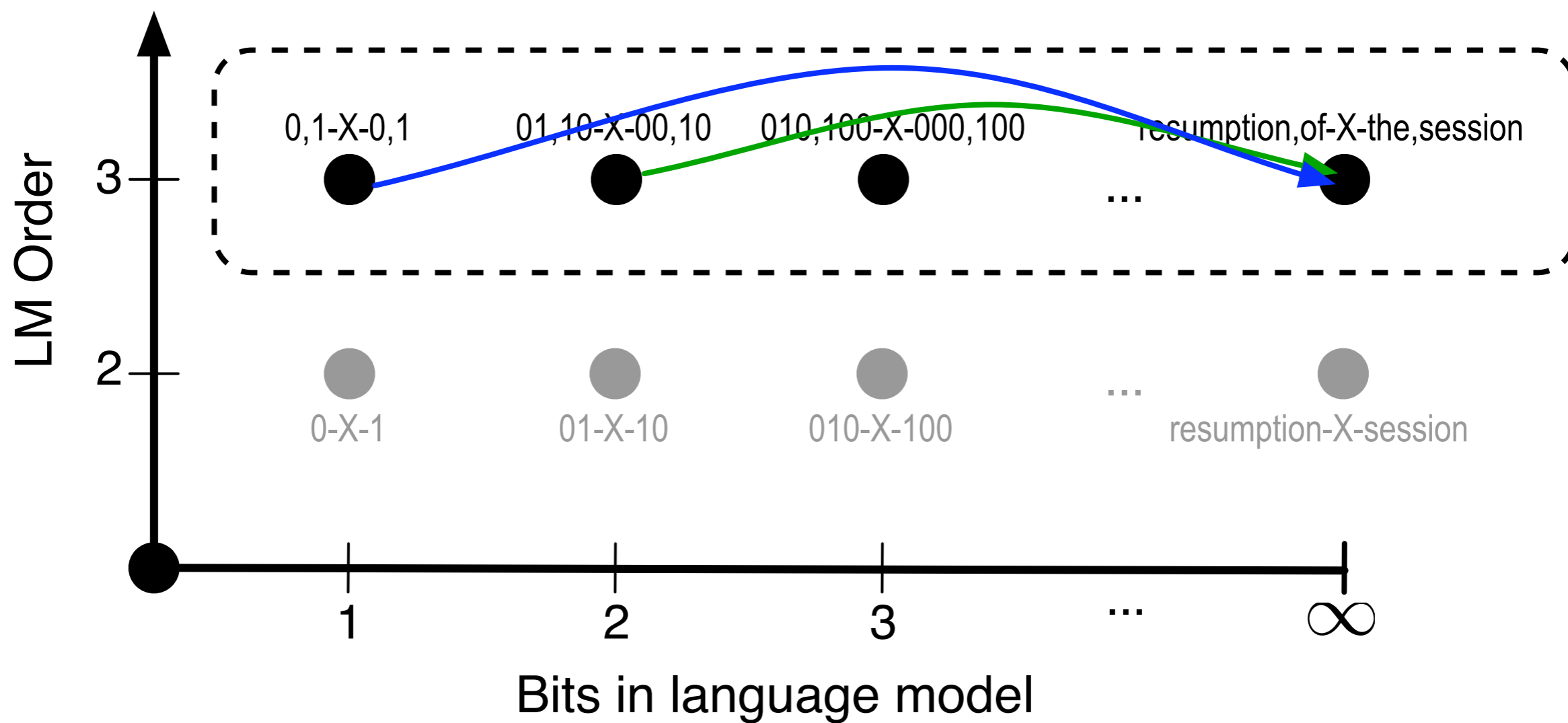
Where to start?



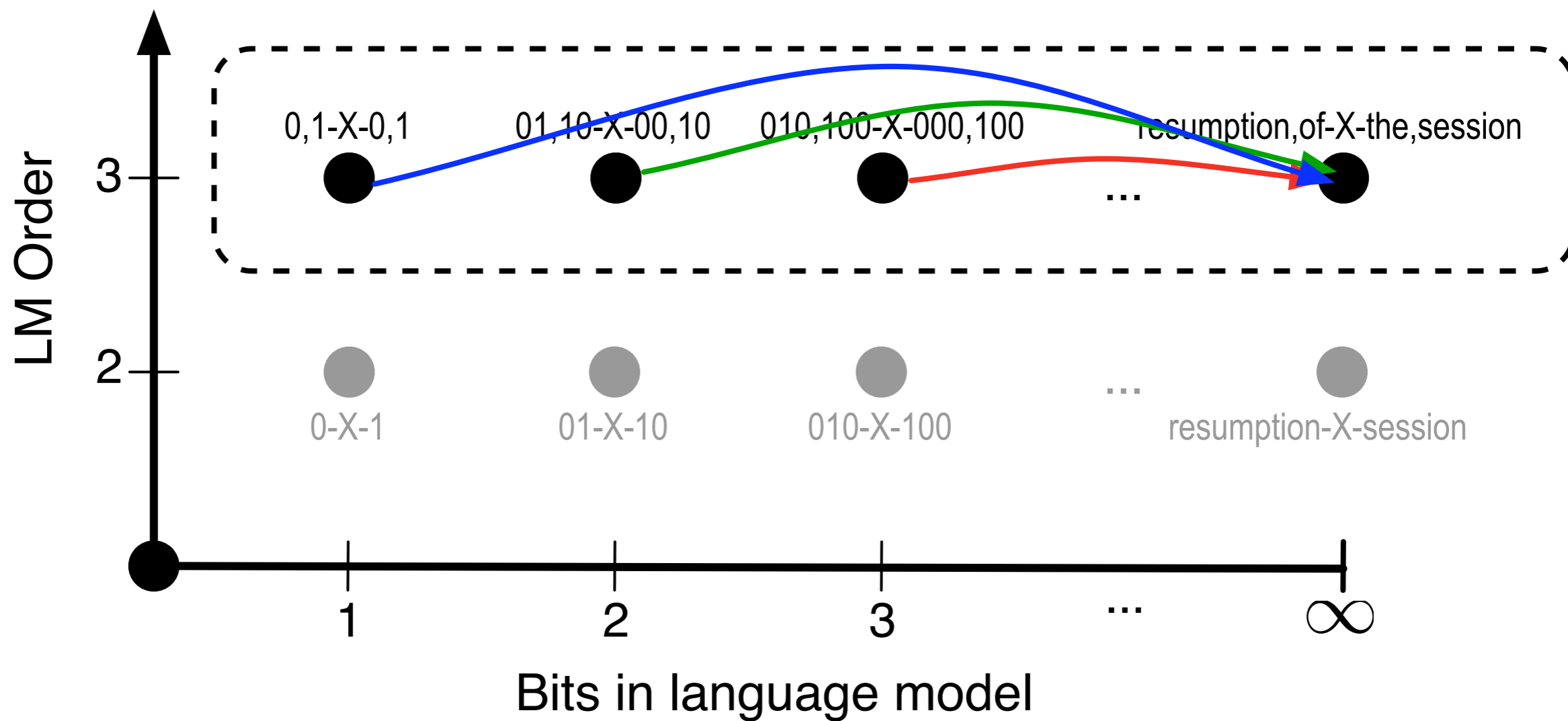
Where to start?






Where to start?



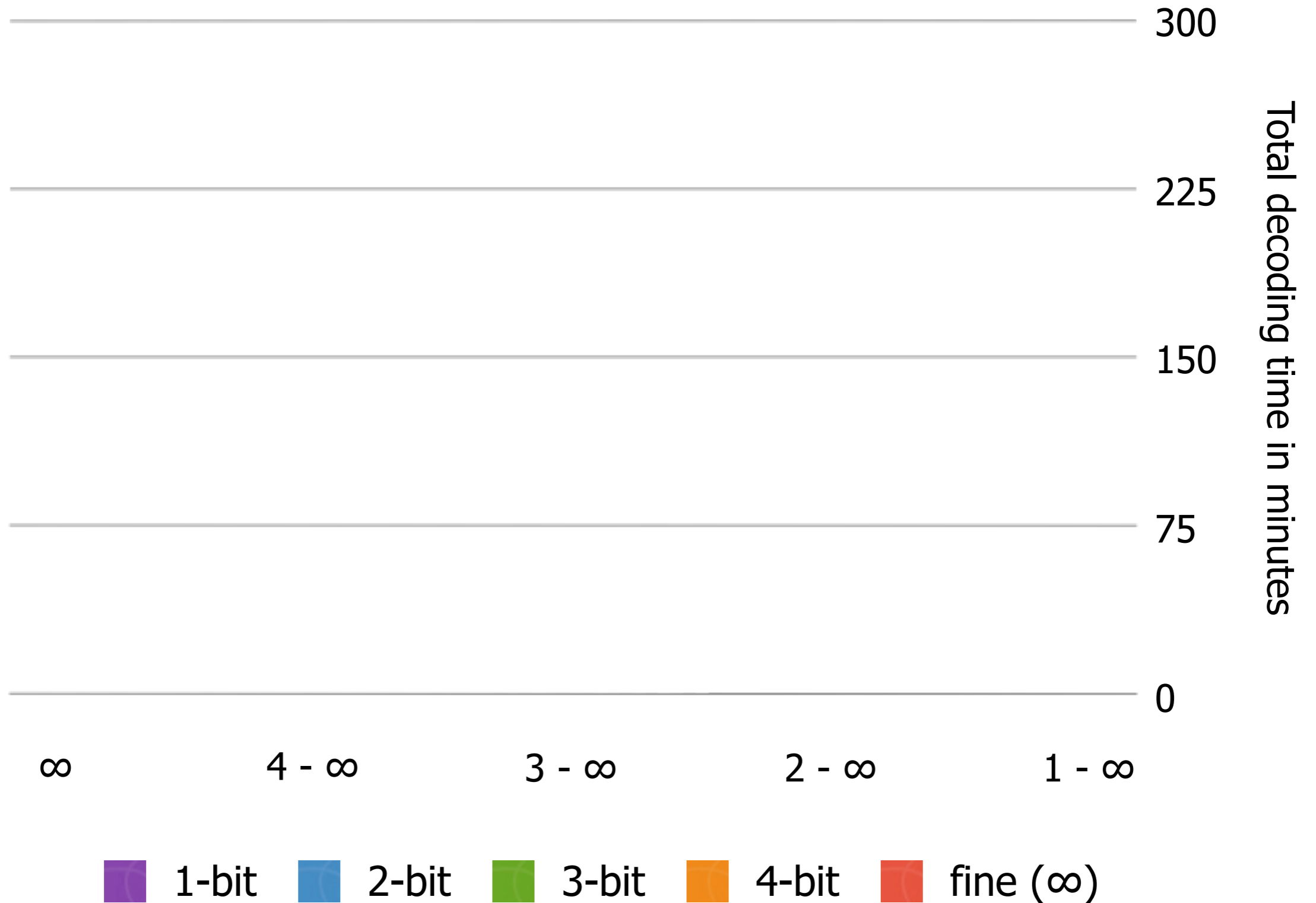
Where to start?



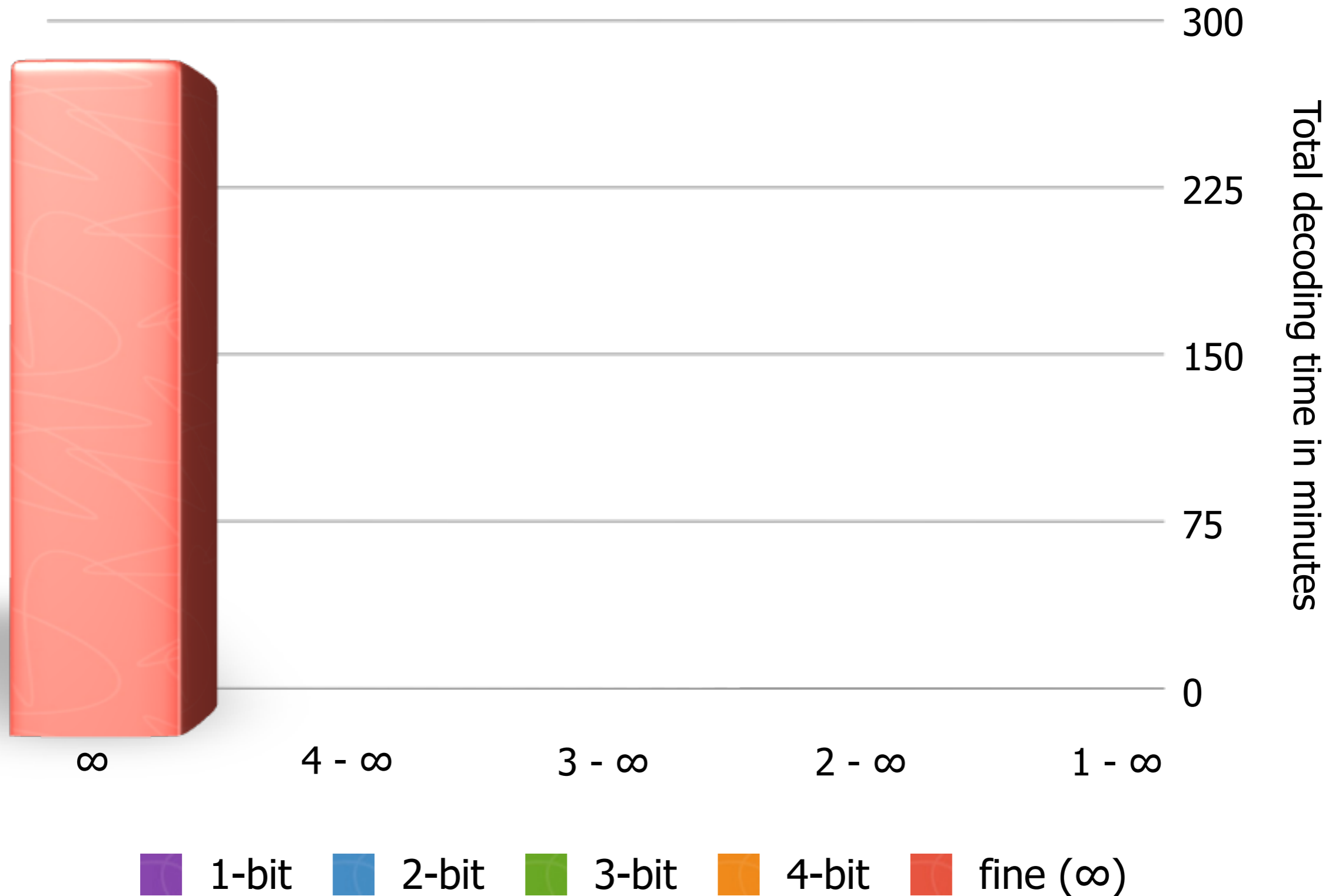
Where to start?

 1-bit  2-bit  3-bit  4-bit  fine (∞)

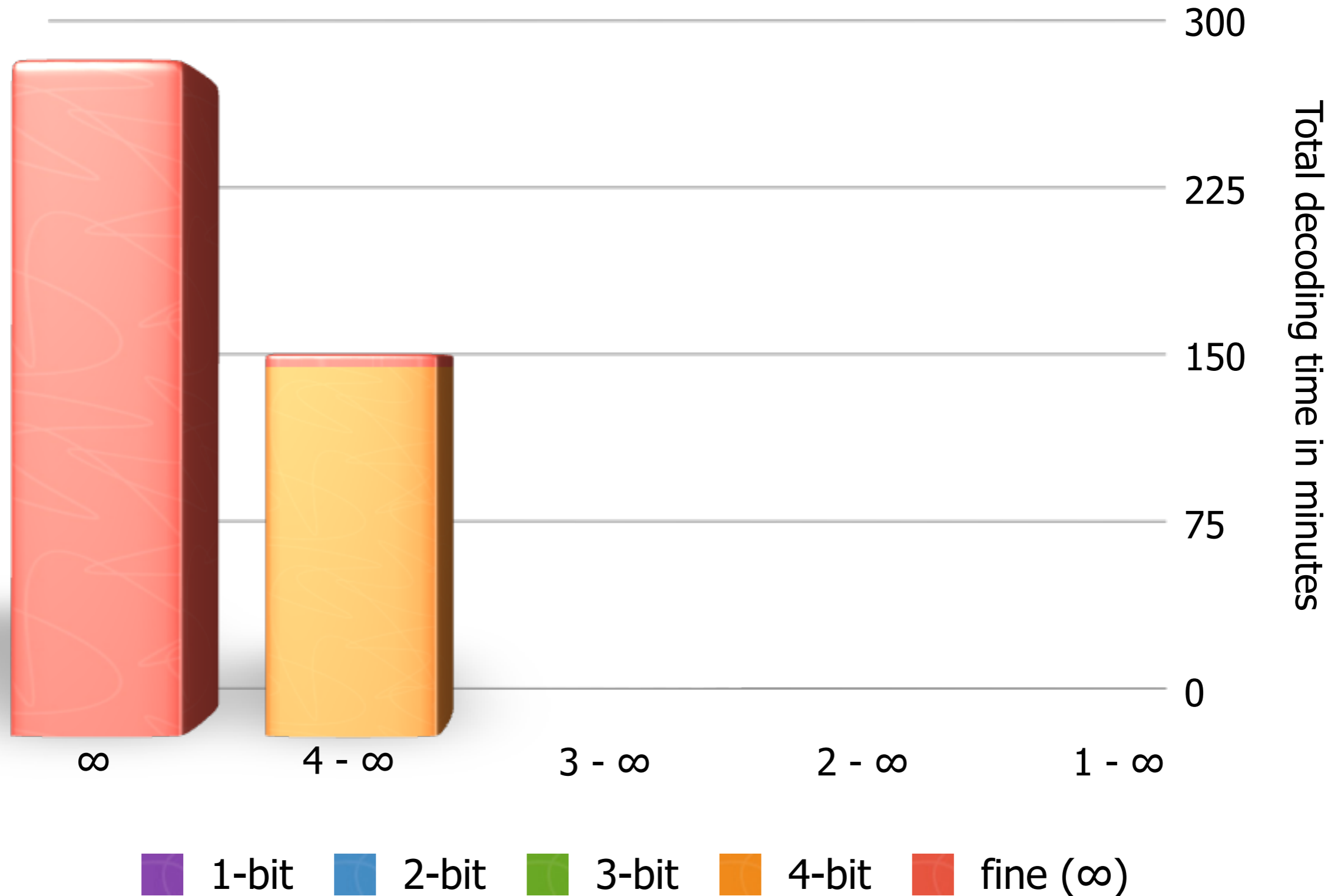
Where to start?



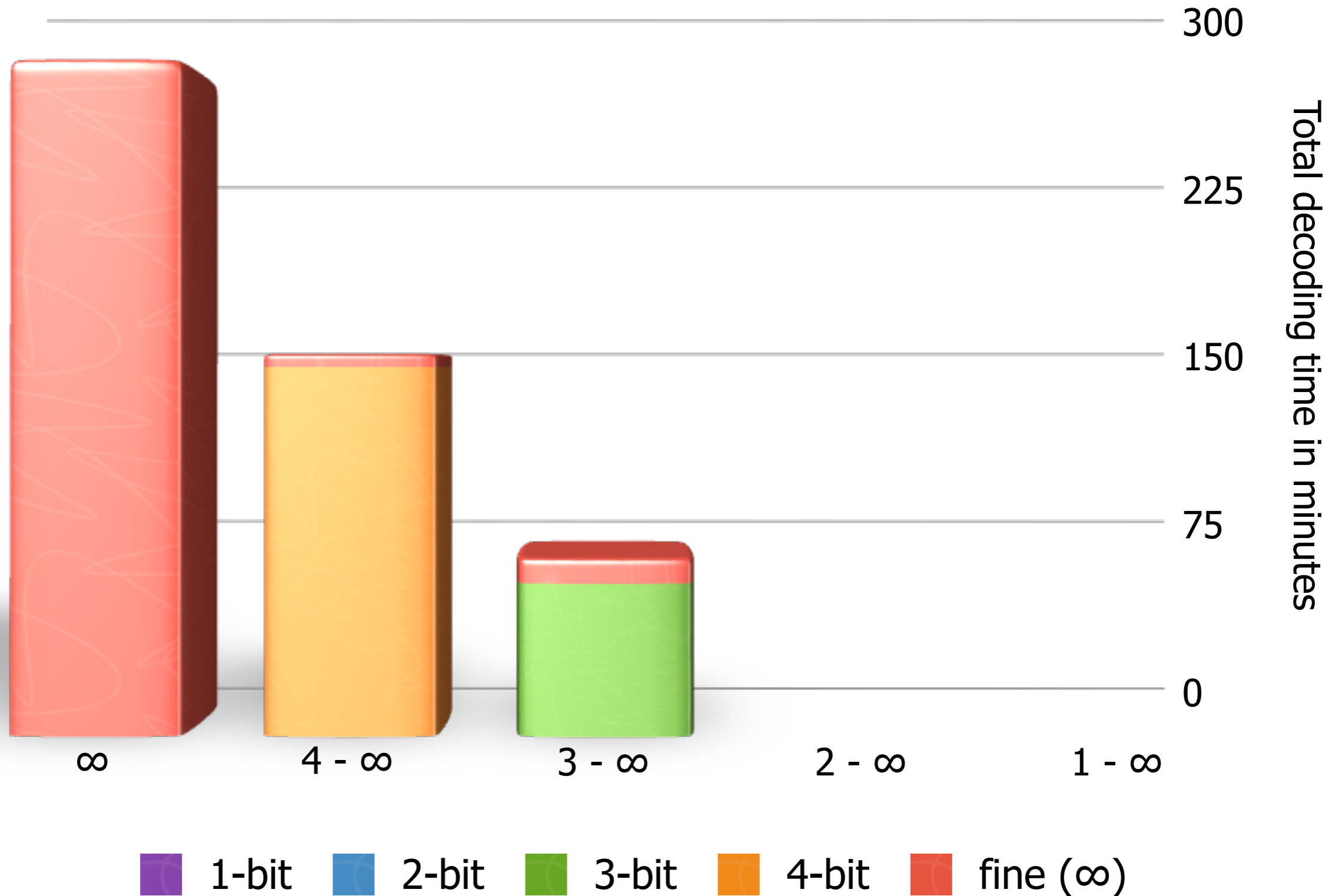
Where to start?



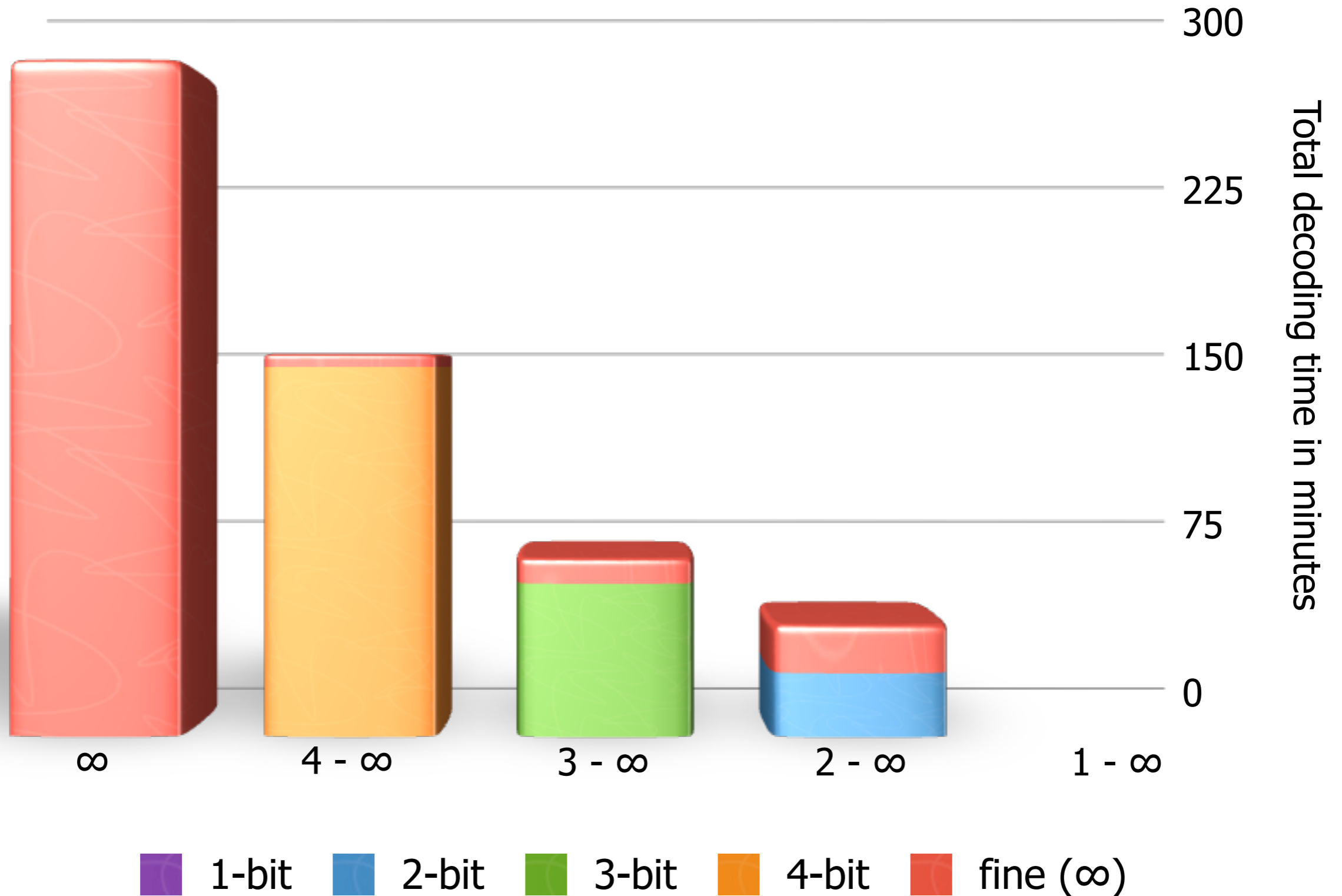
Where to start?



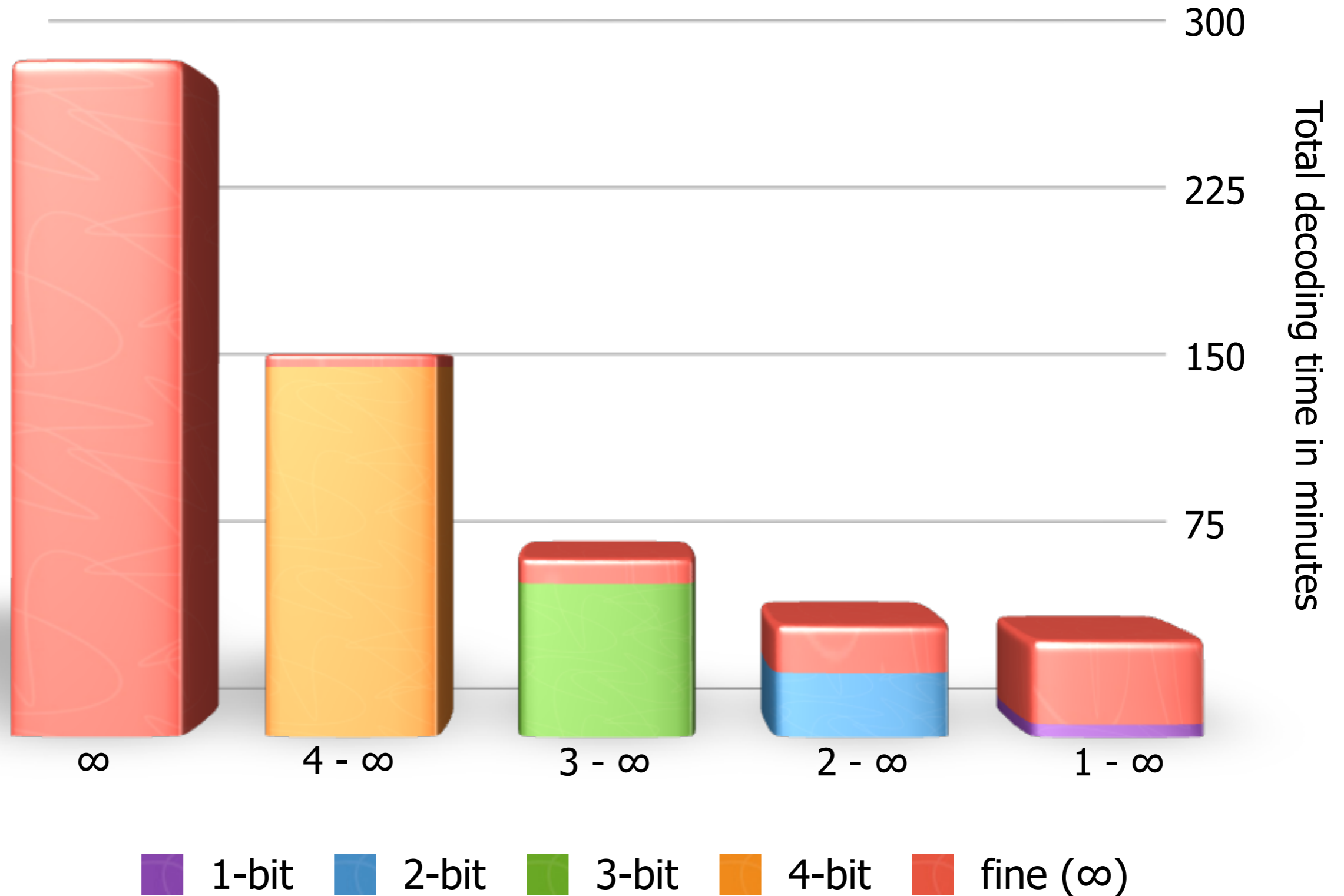
Where to start?



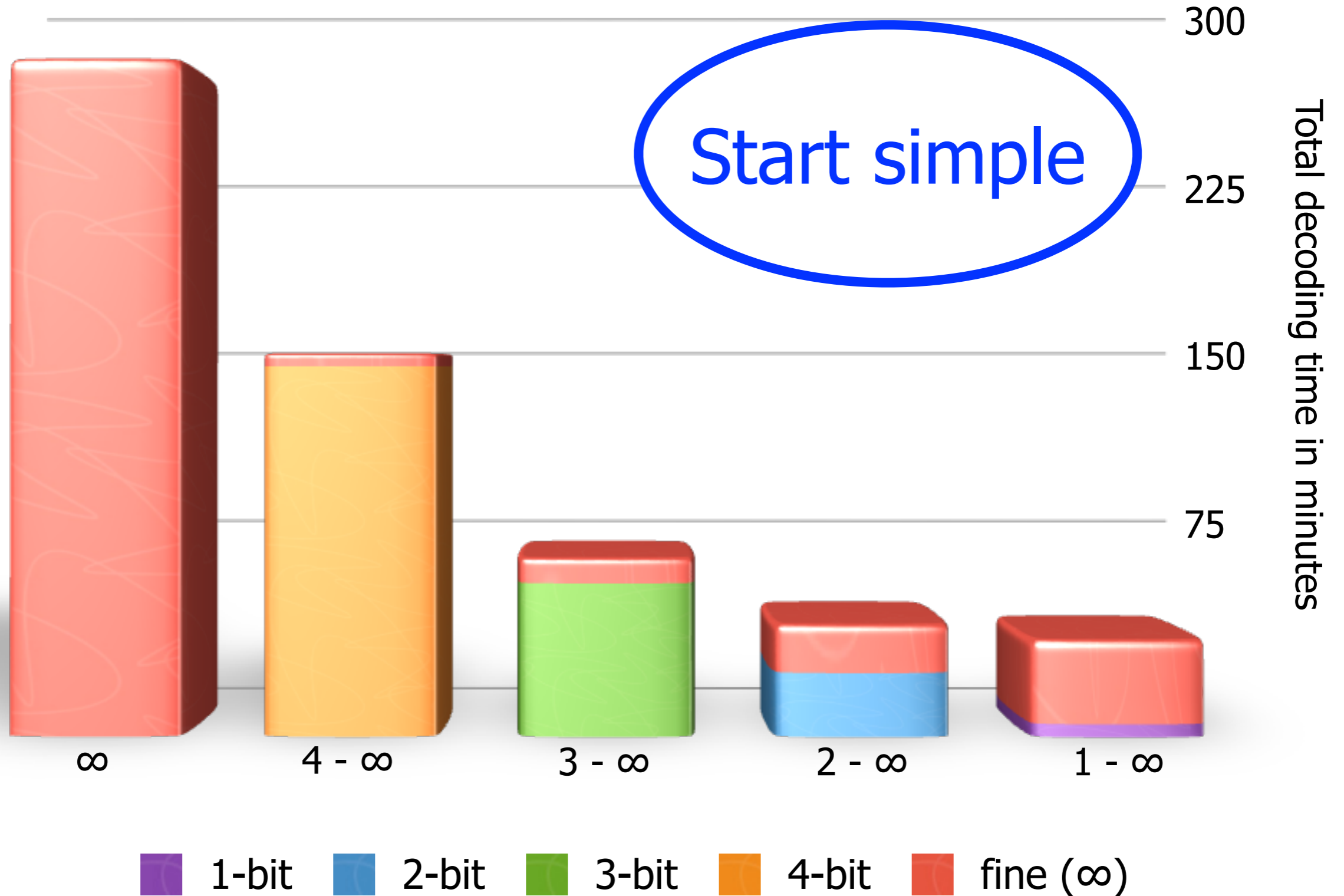
Where to start?



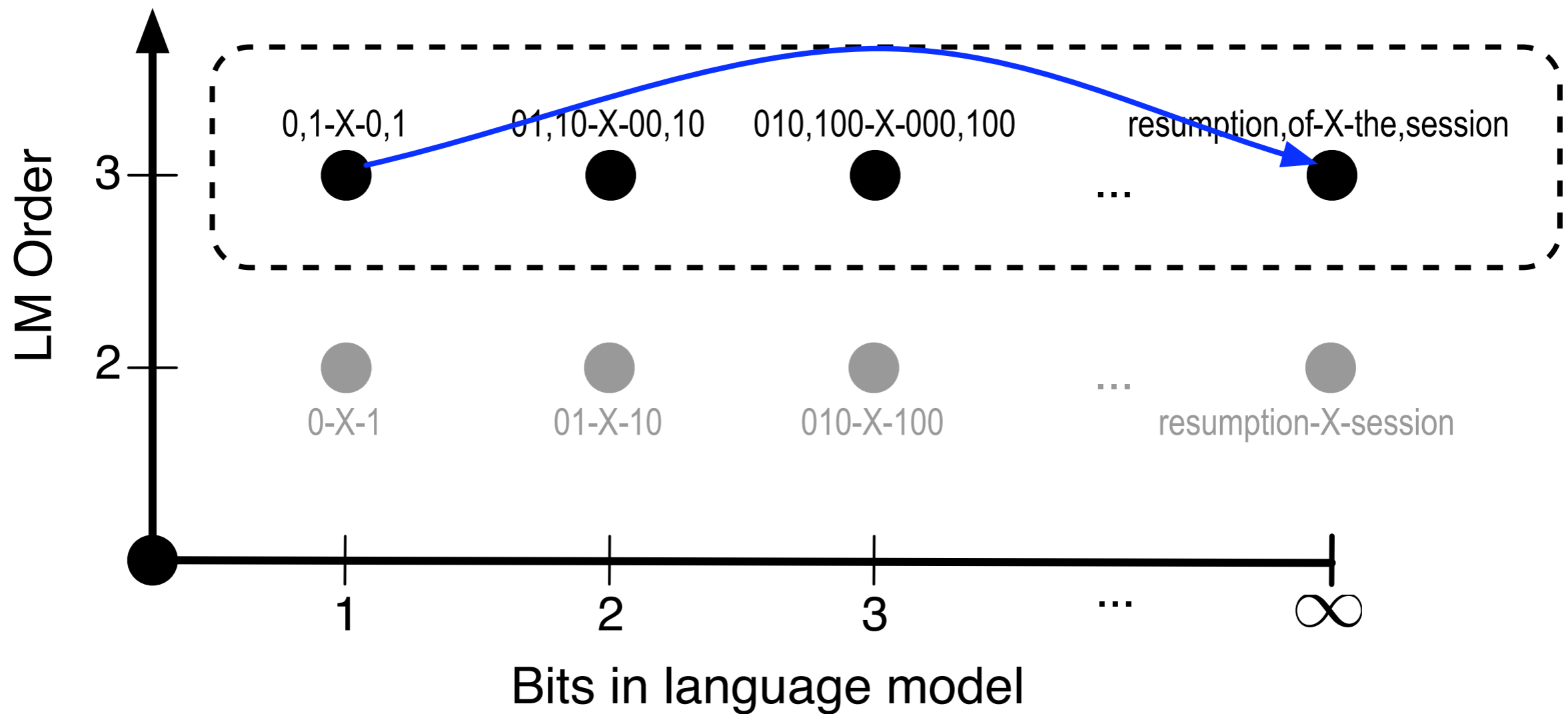
Where to start?



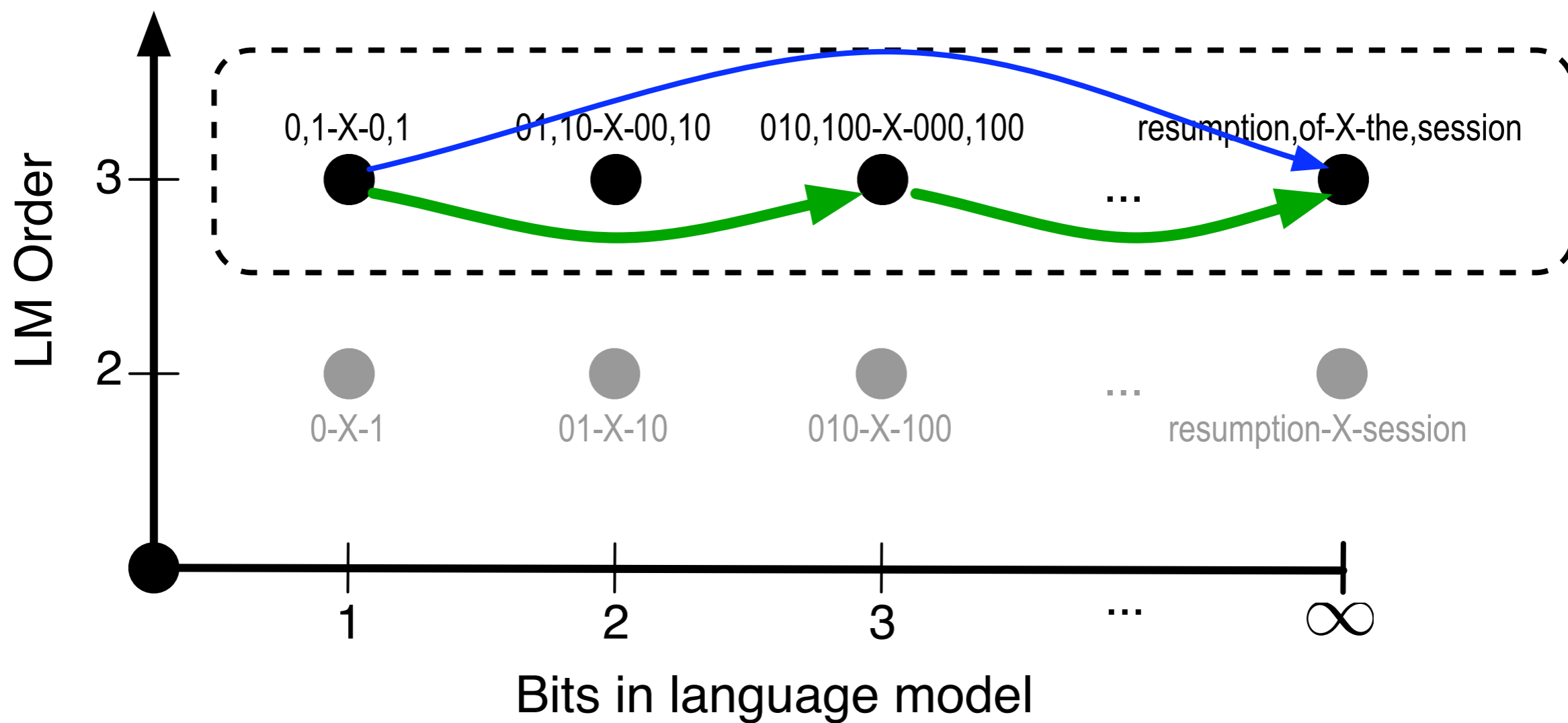
Where to start?



How many passes?

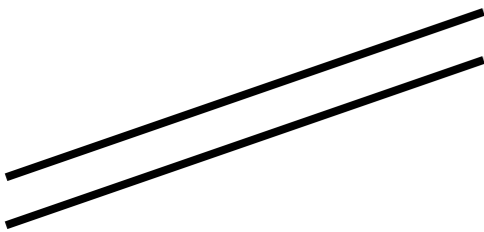


How many passes?





How many passes?

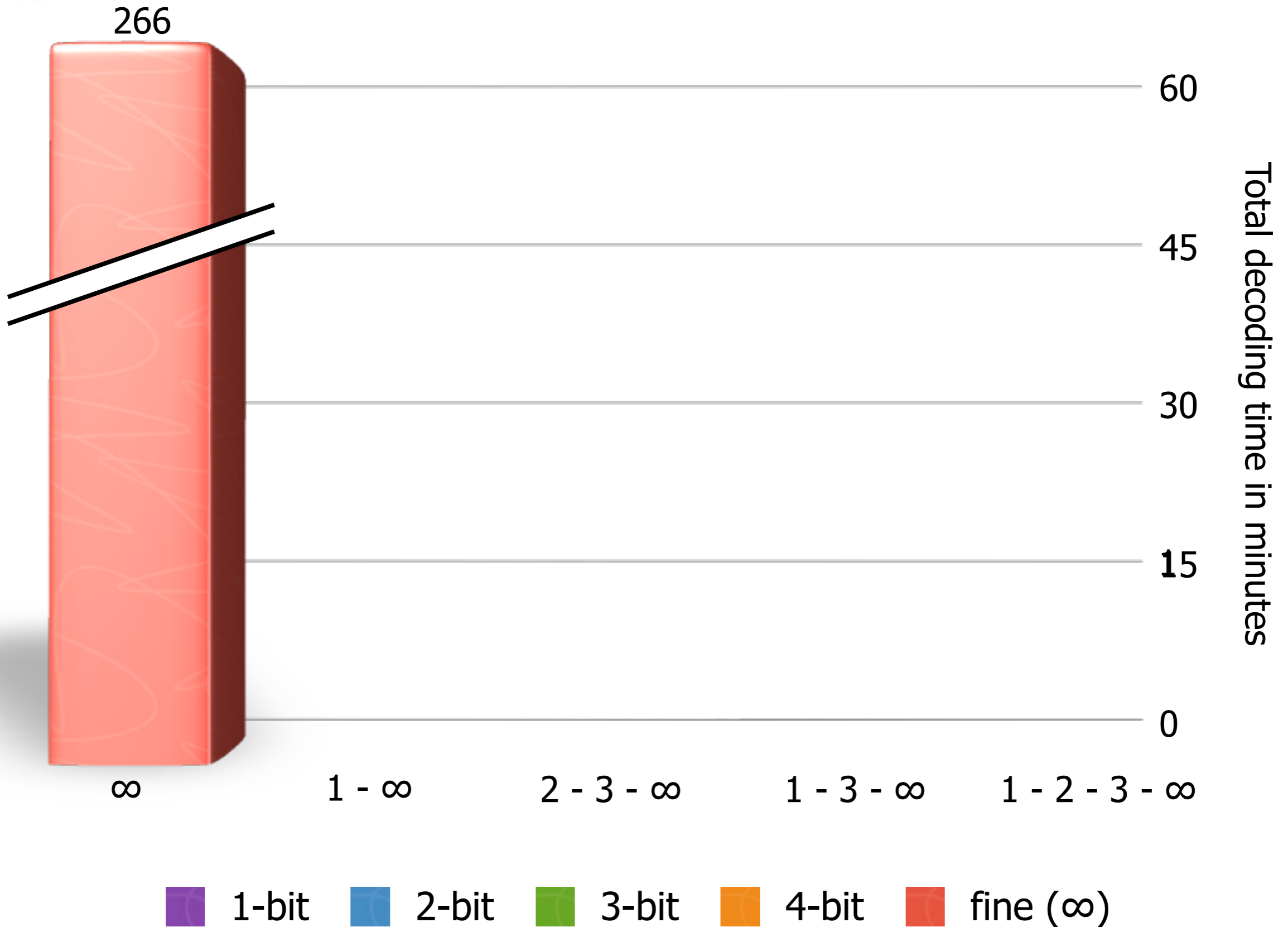
266



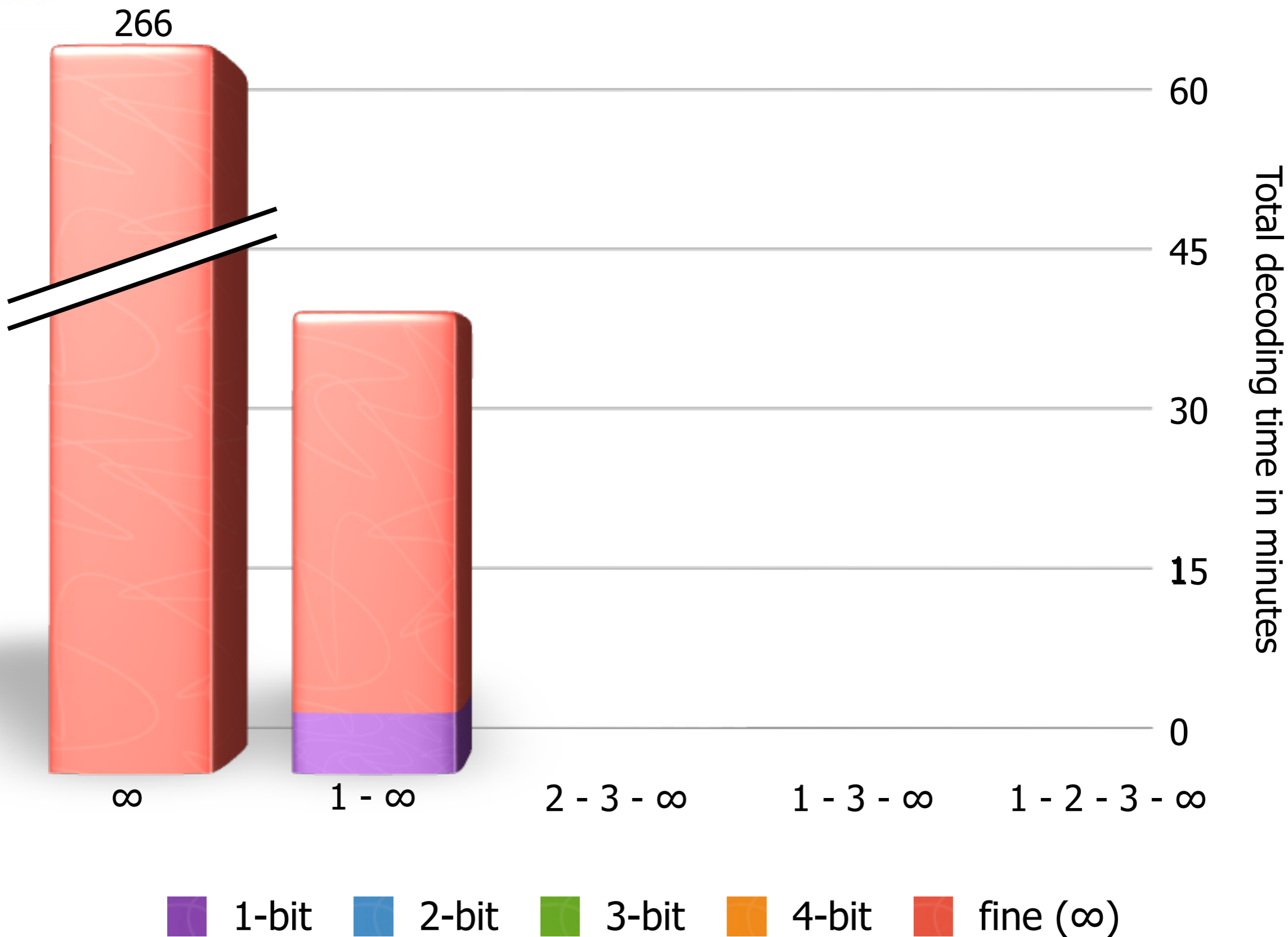
1

 1-bit  2-bit  3-bit  4-bit  fine (∞)

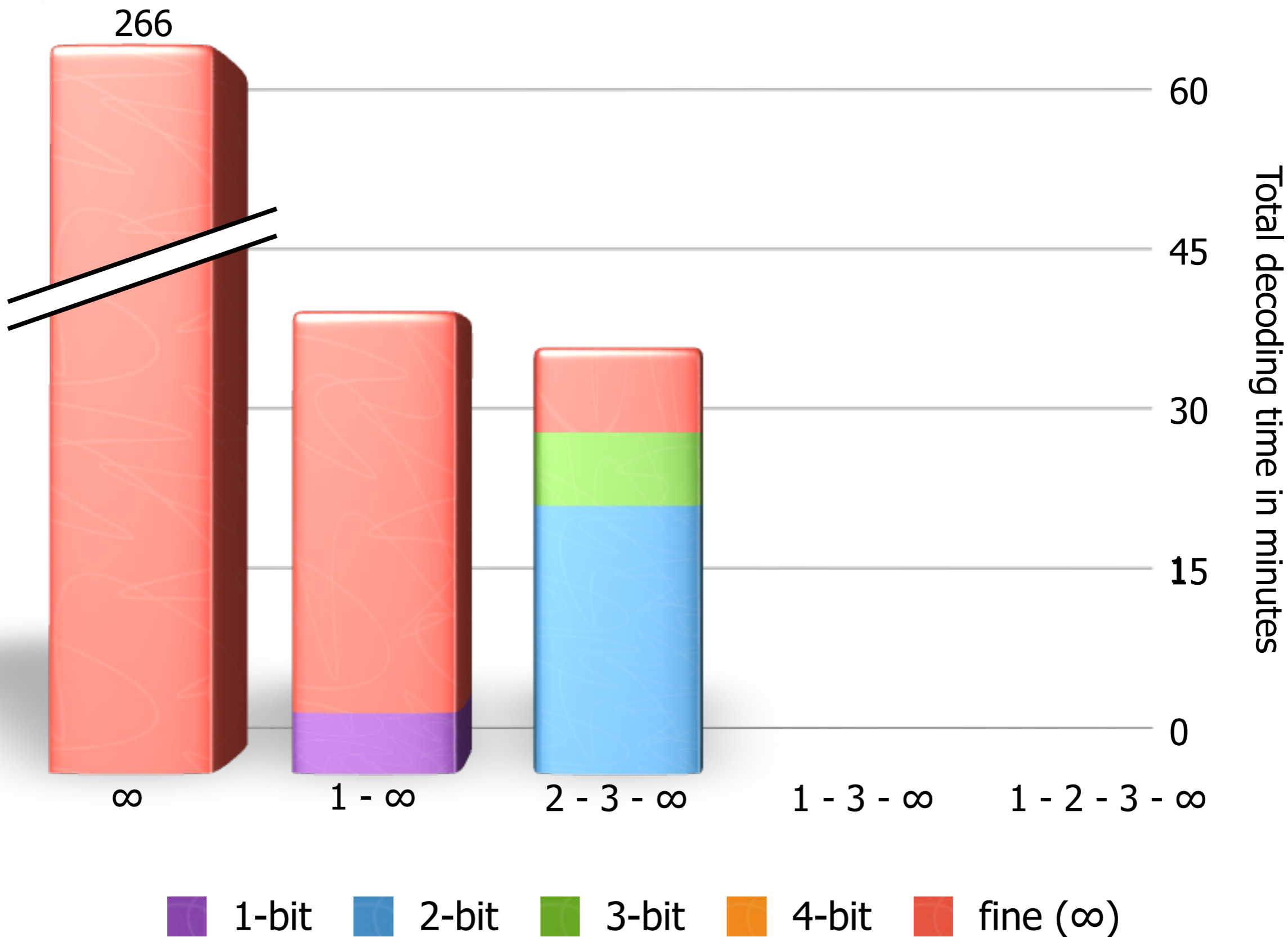
How many passes?



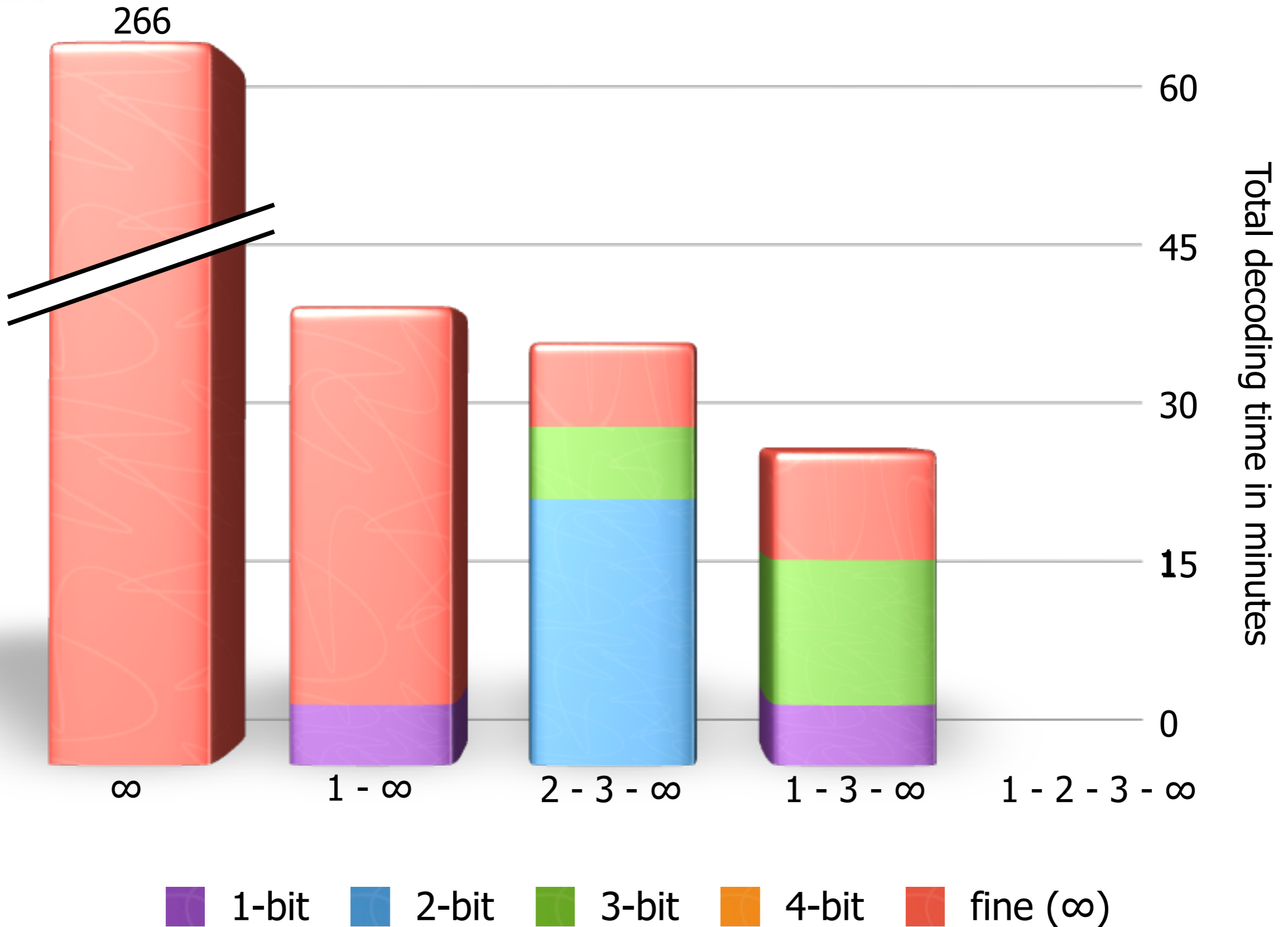
How many passes?



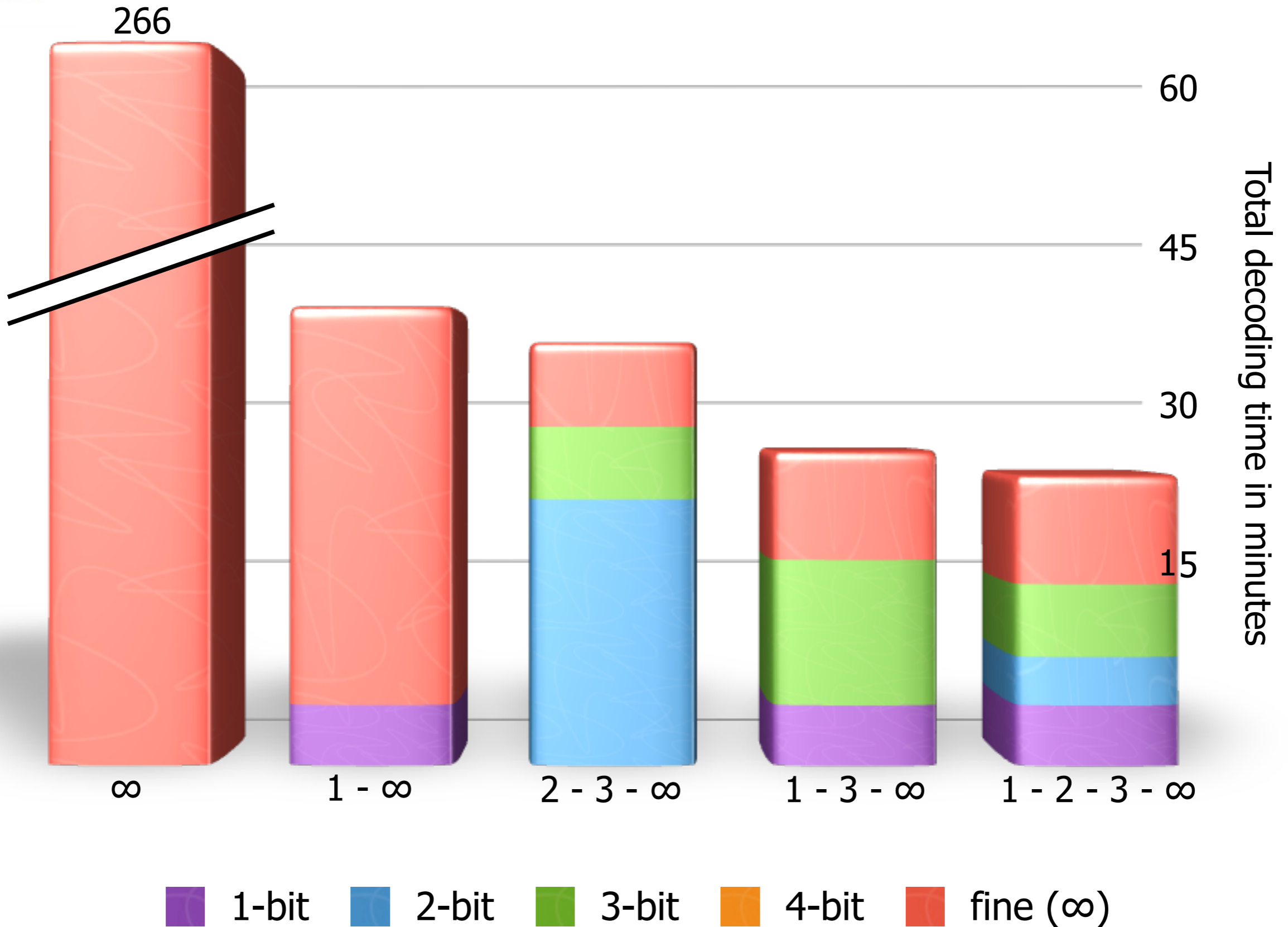
How many passes?



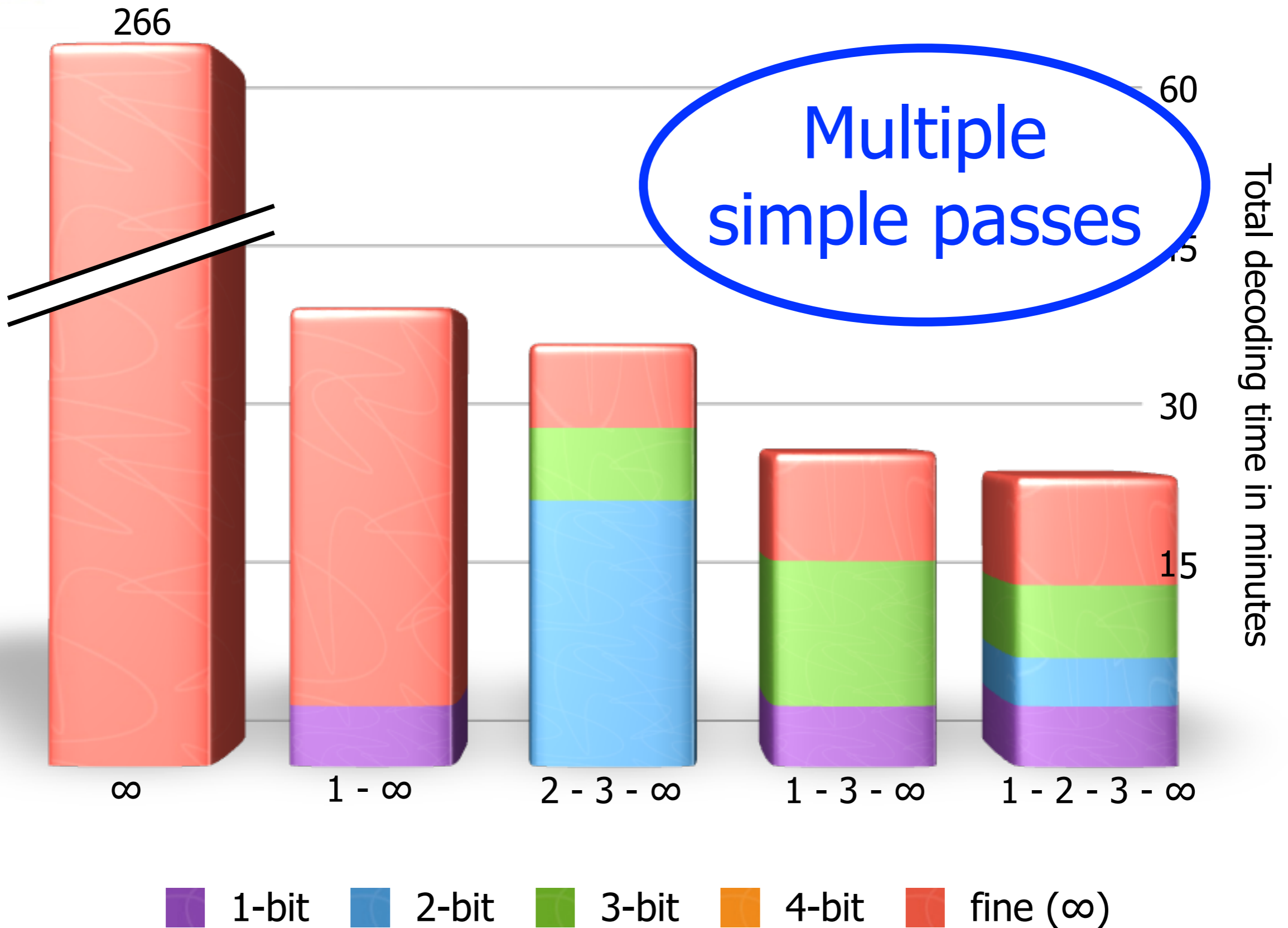
How many passes?



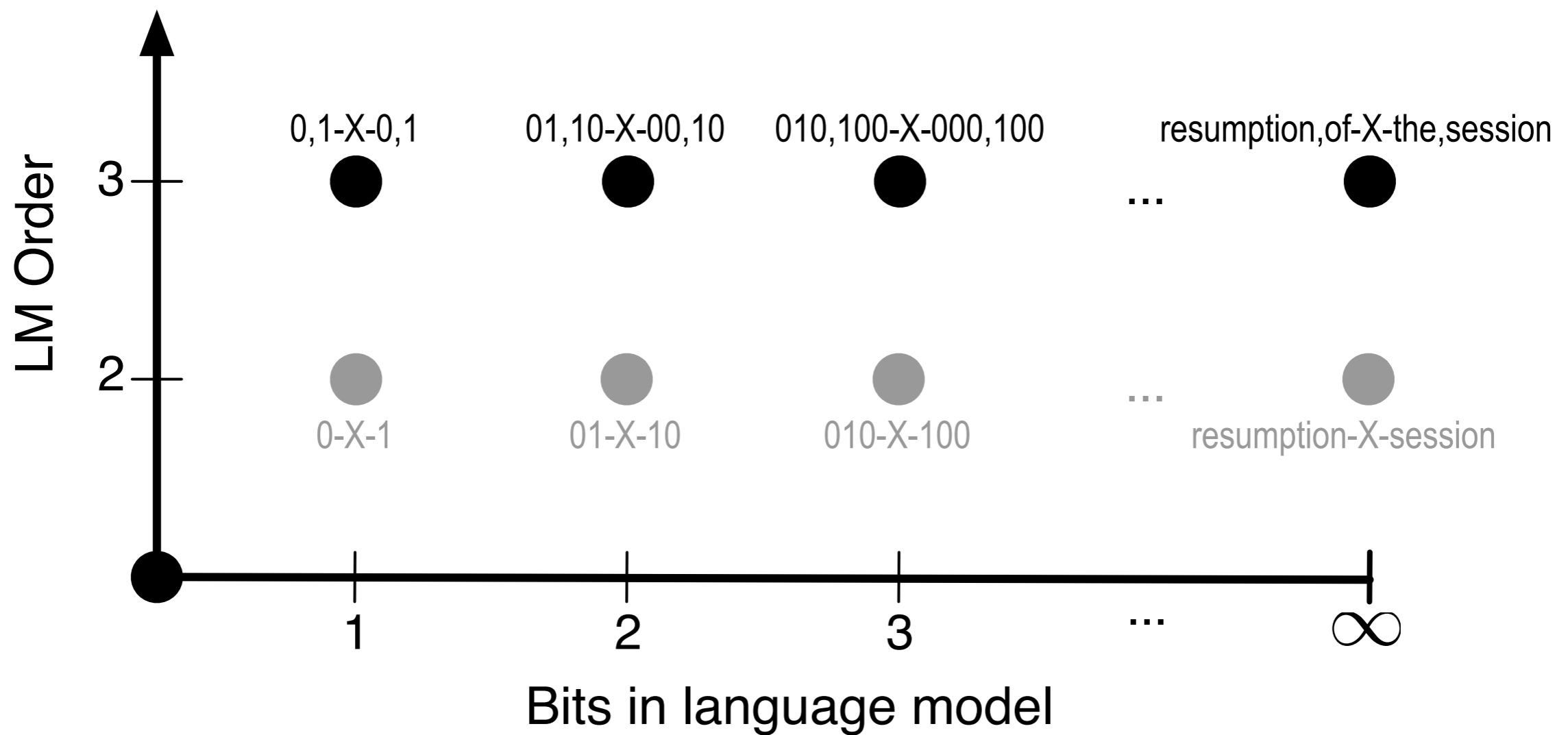
How many passes?



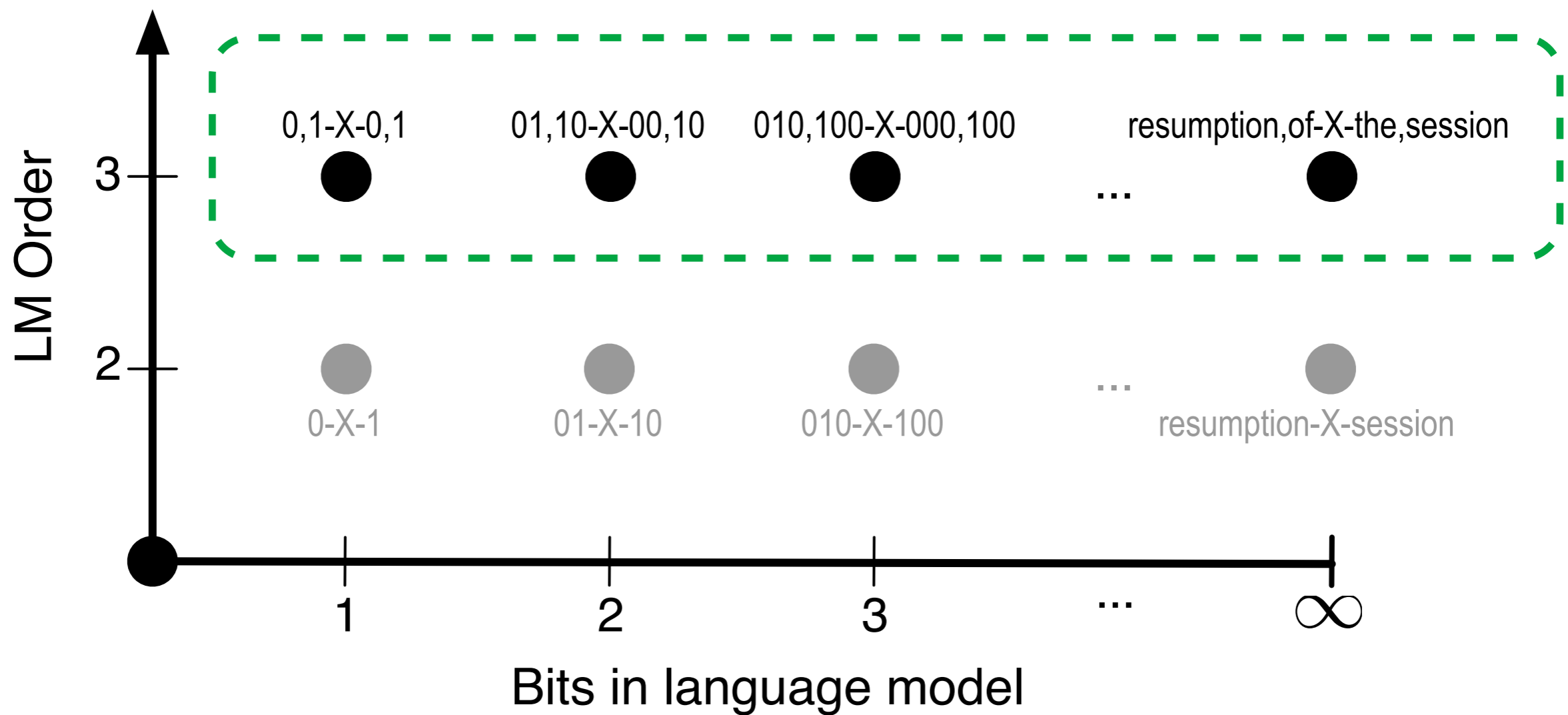
How many passes?



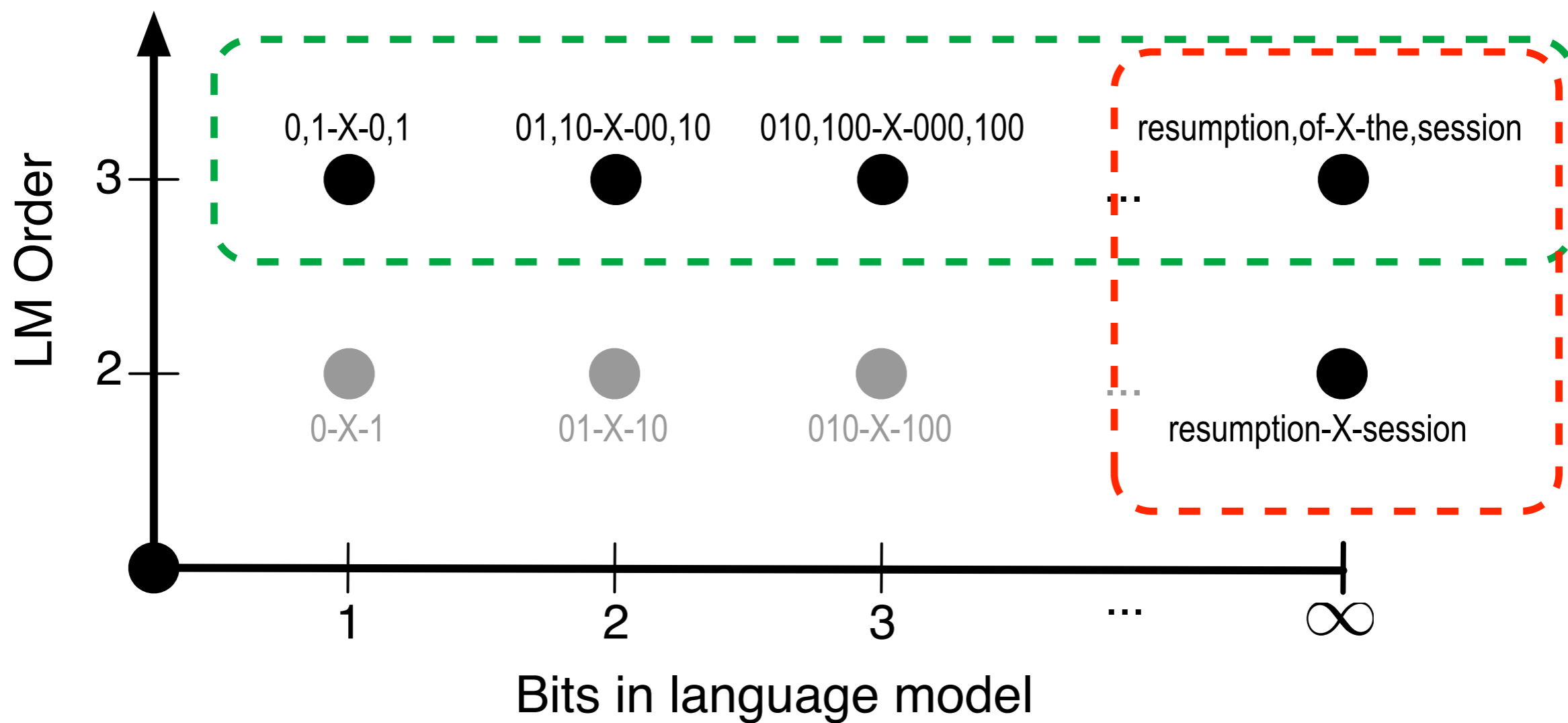
Use Encoding or Order?



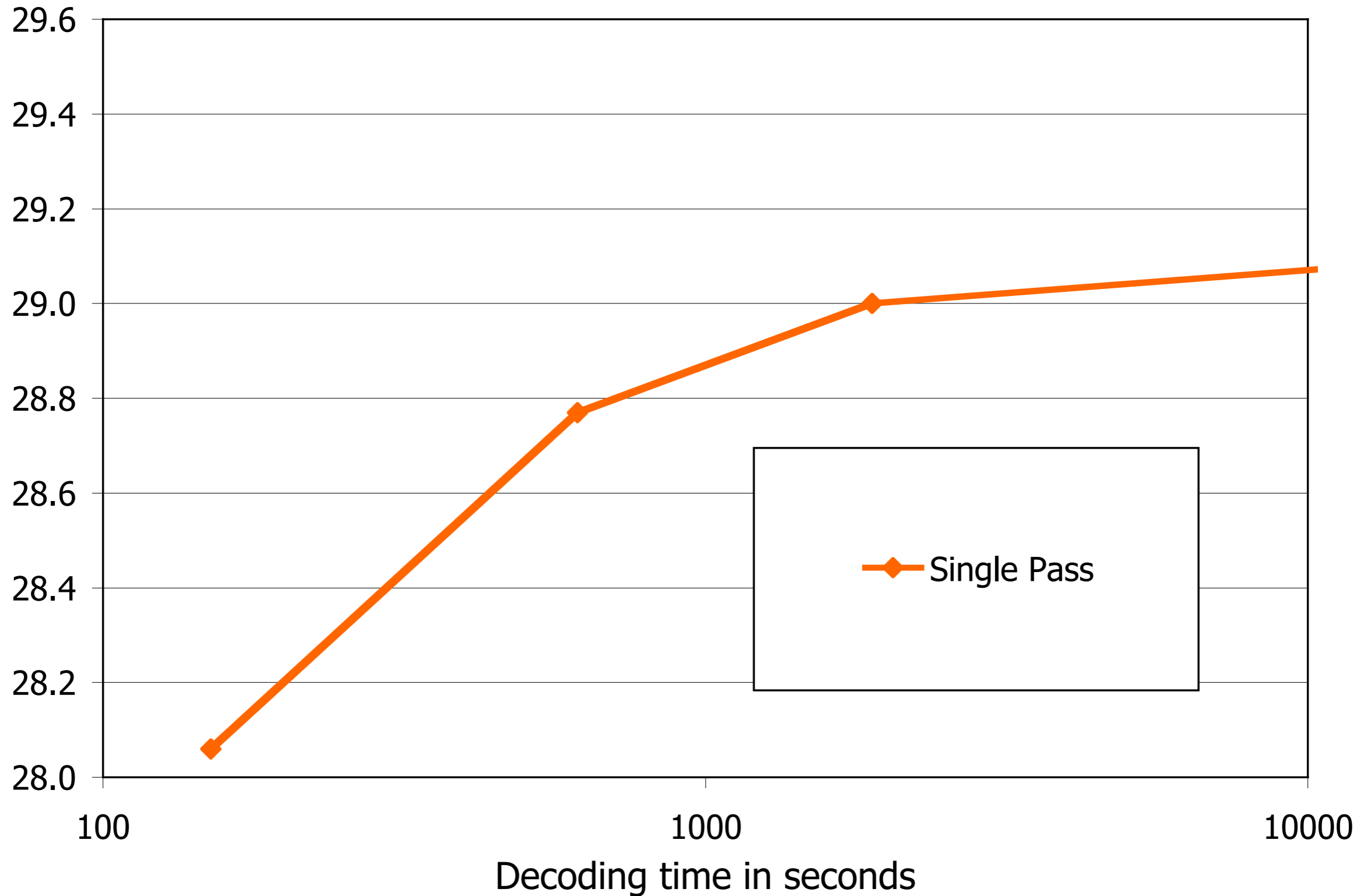
Use Encoding or Order?



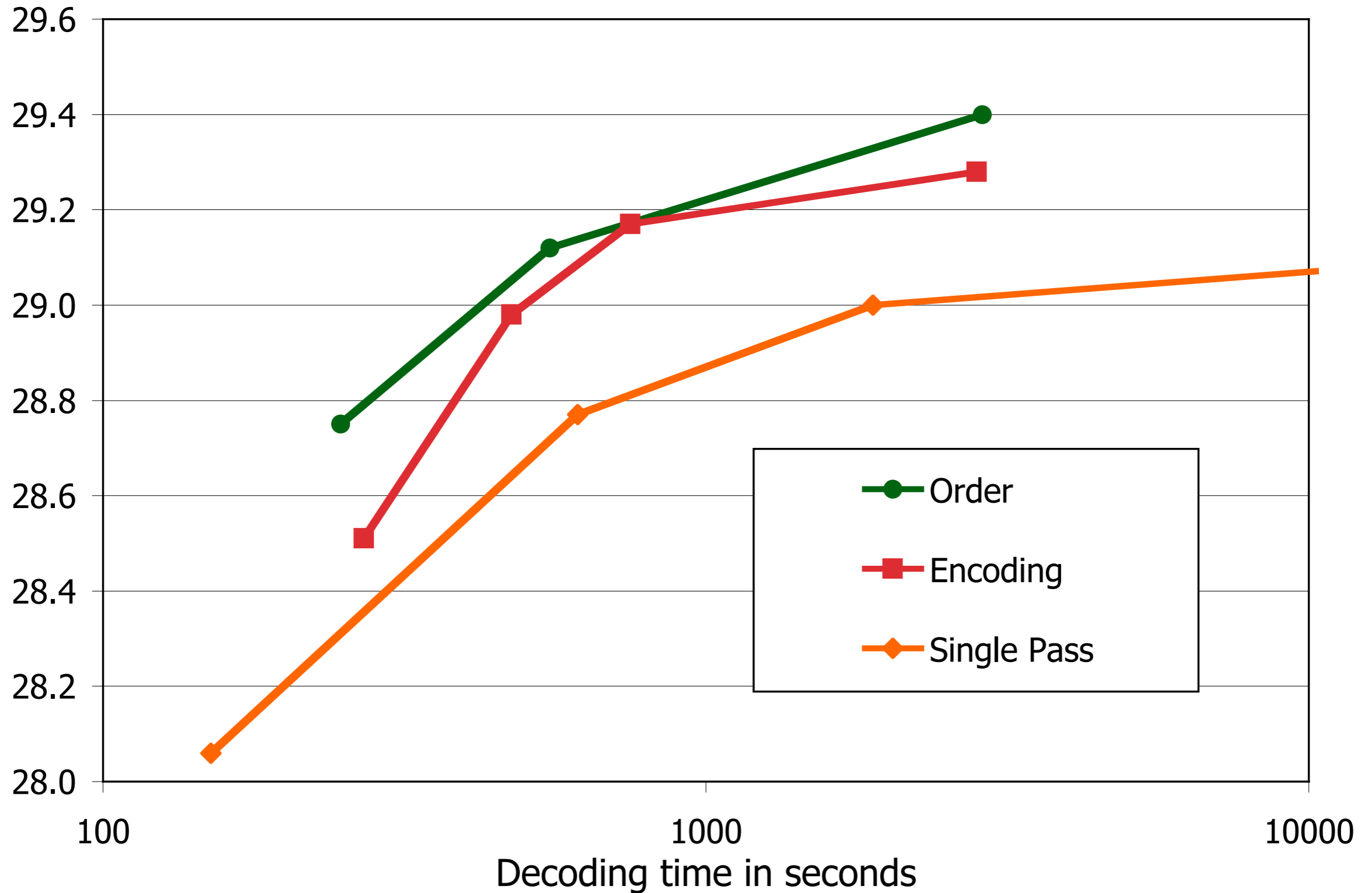
Use Encoding or Order?



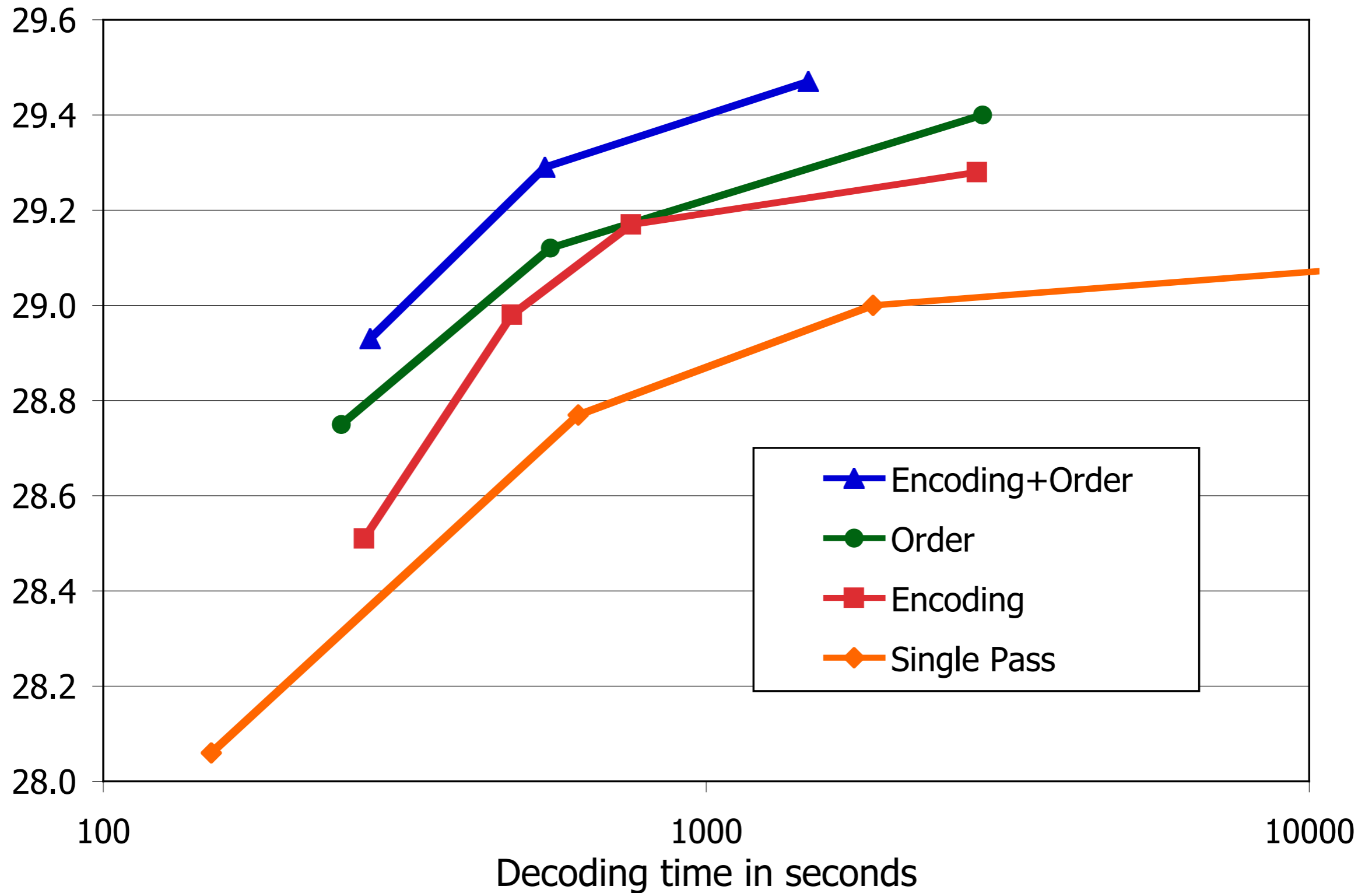
Use Encoding or Order?



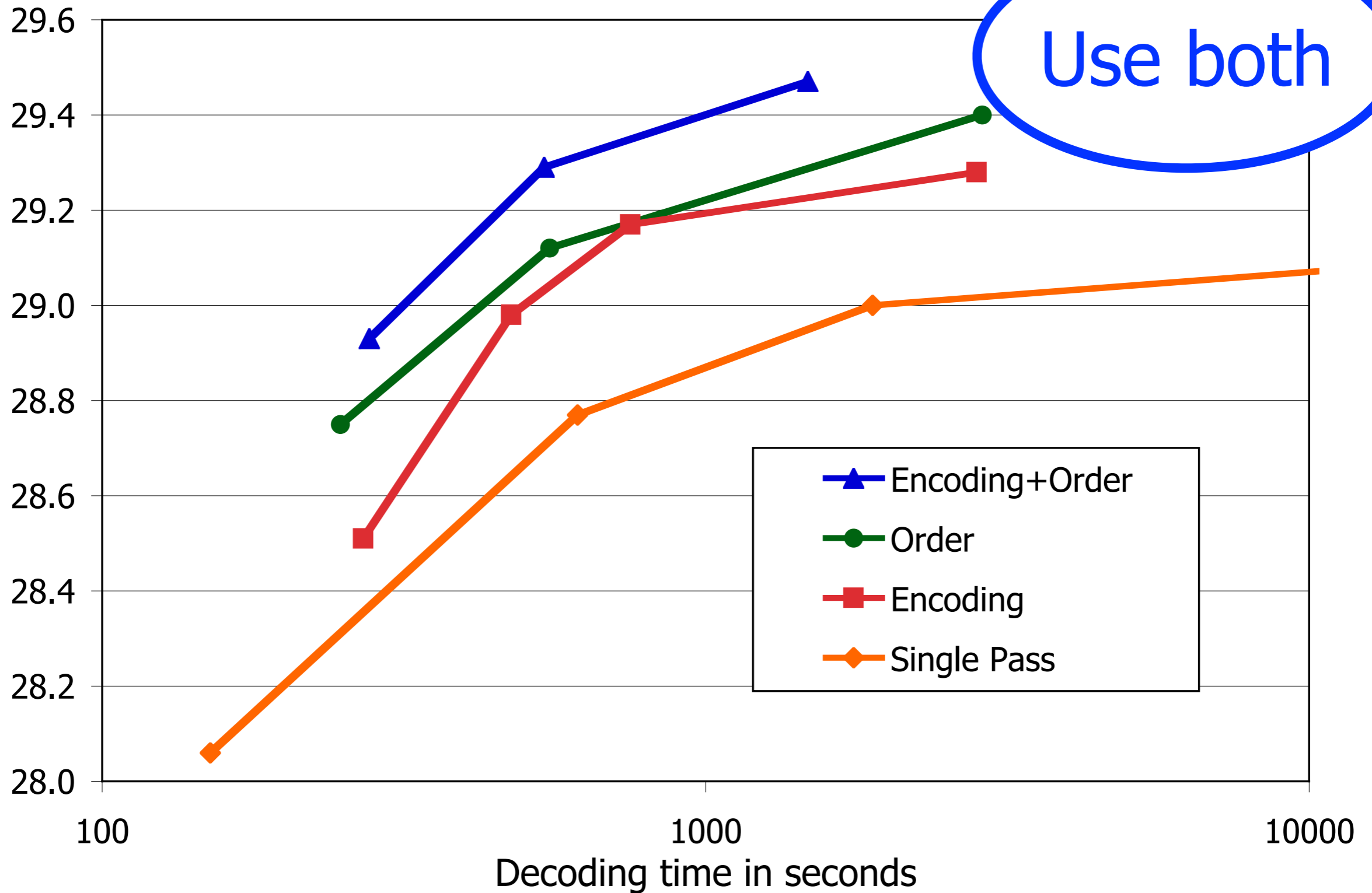
Use Encoding or Order?



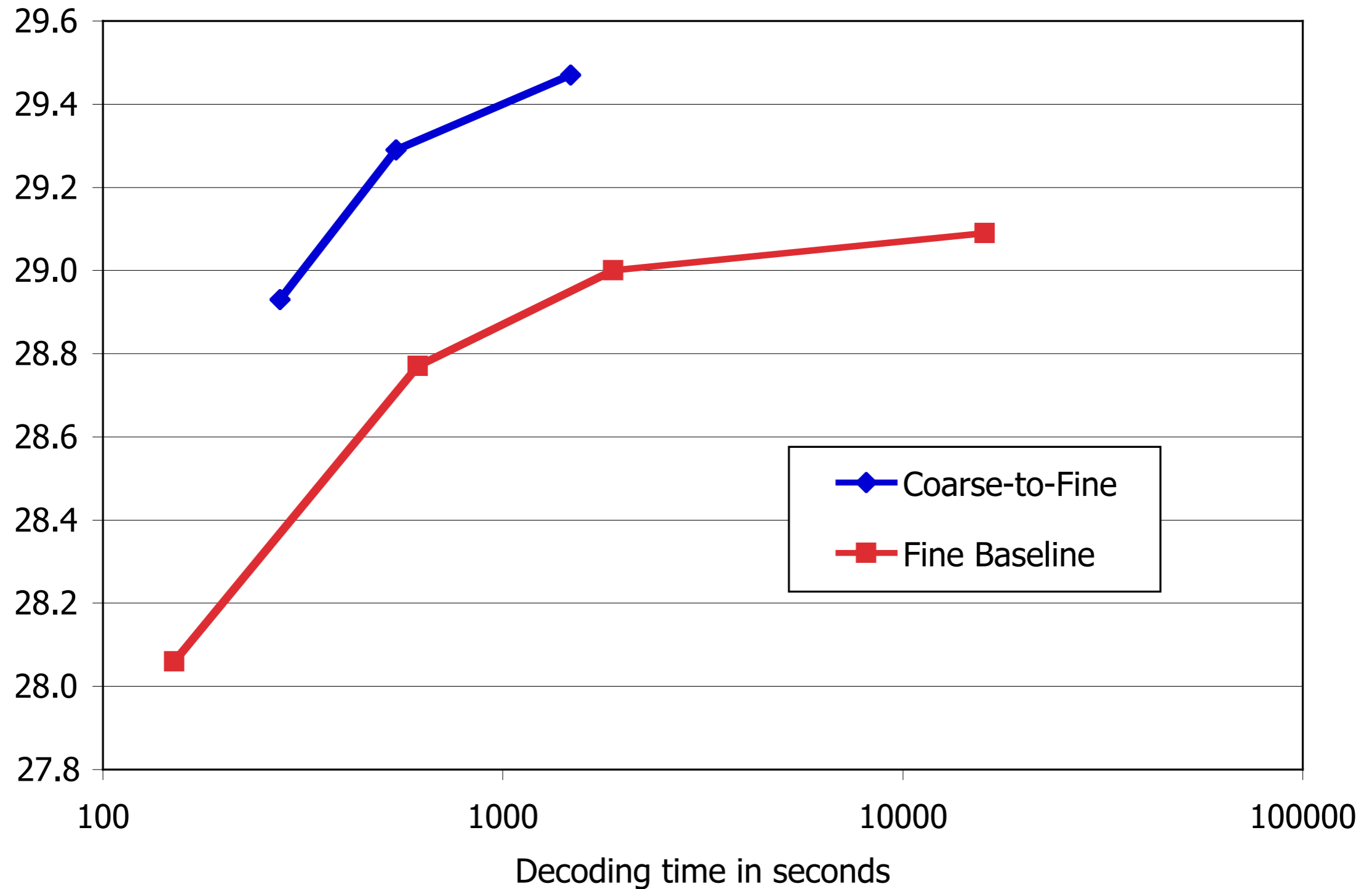
Use Encoding or Order?



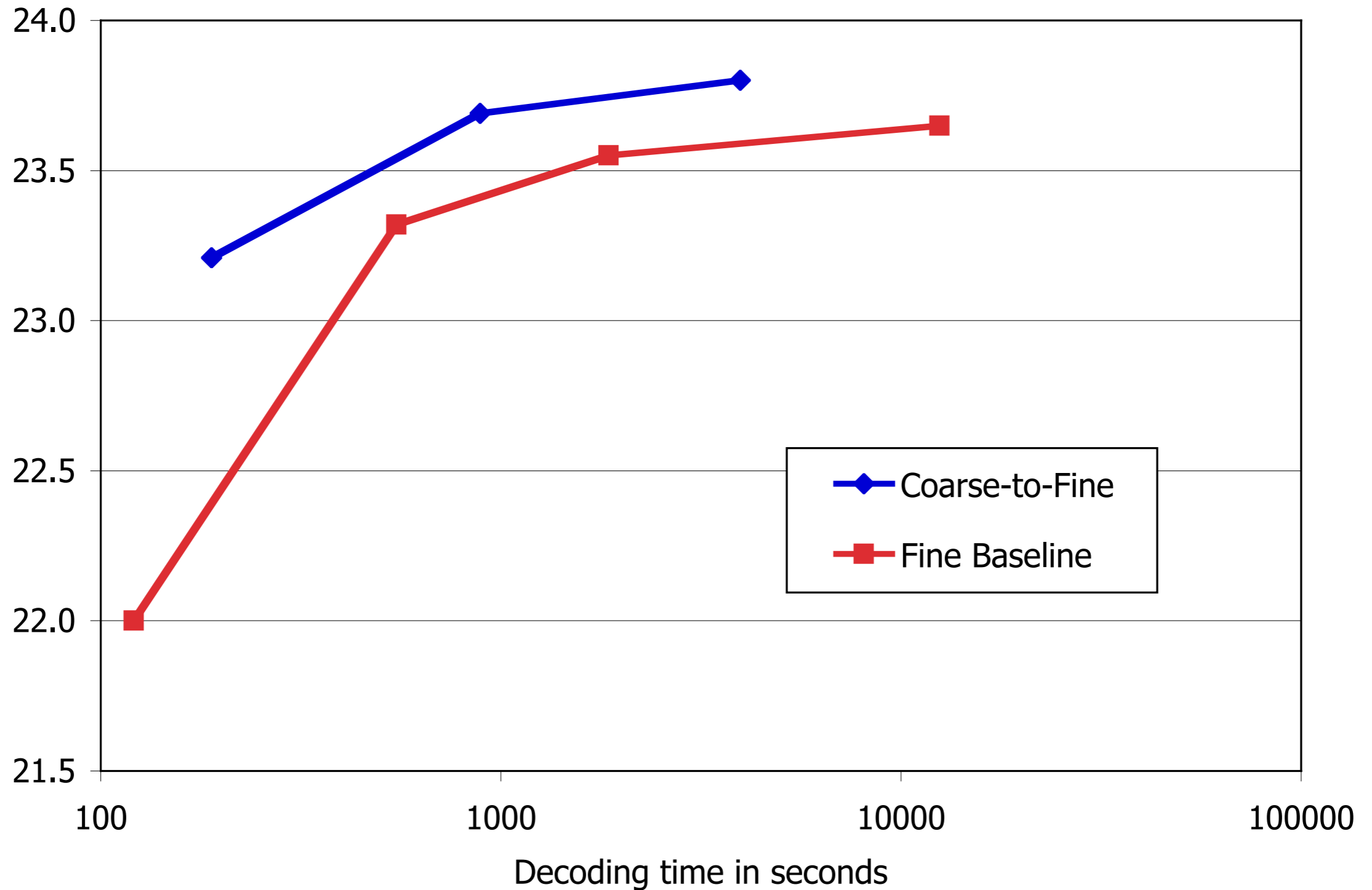
Use Encoding or Order?



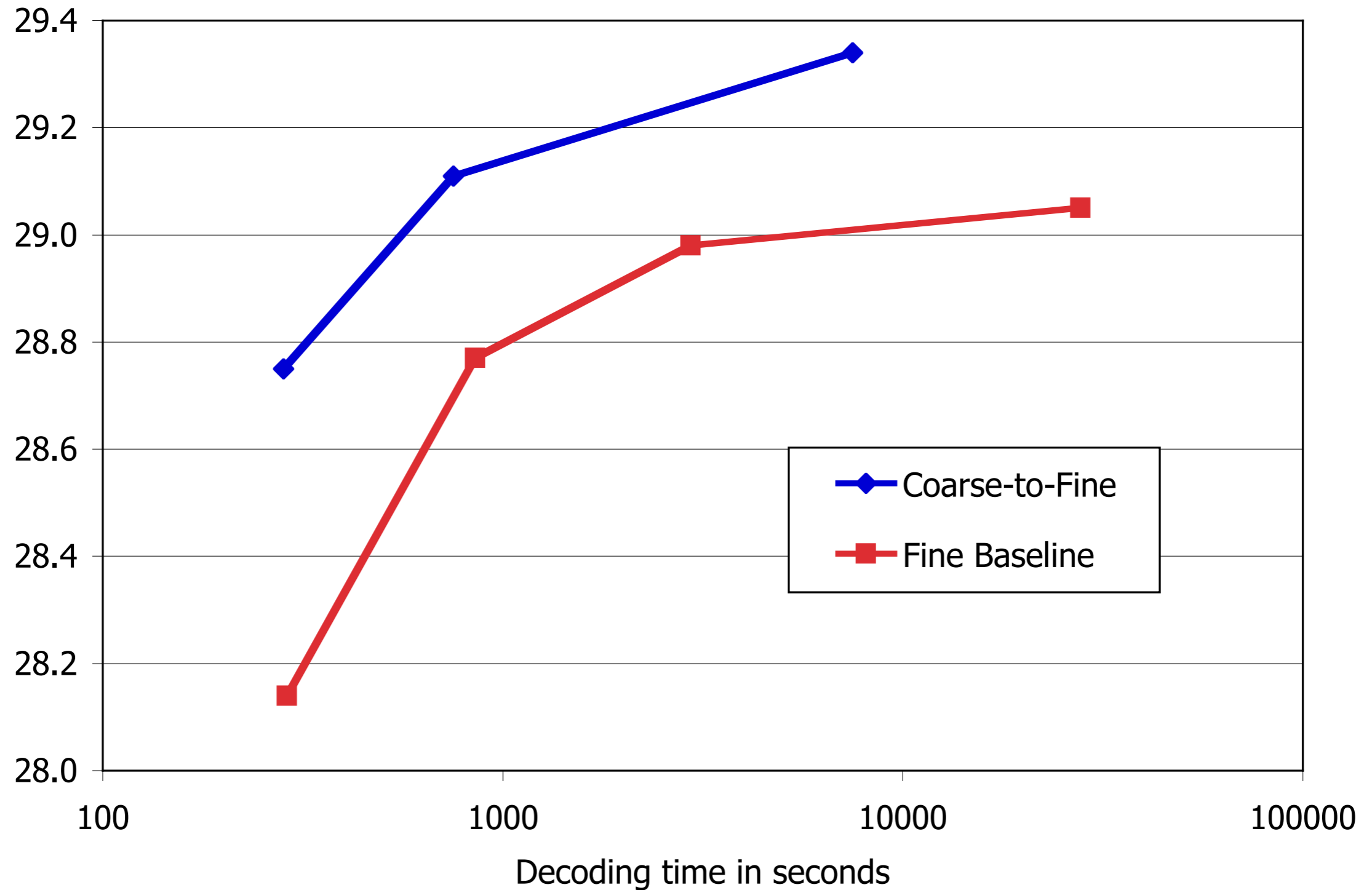
Final Results (Spanish)



Final Results (German)



Final Results (French)

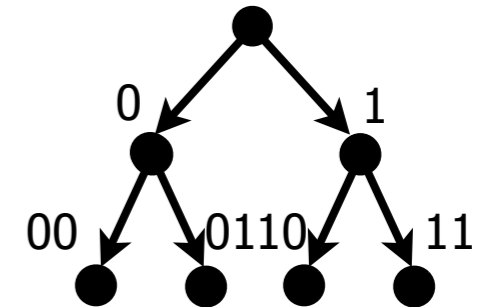




Conclusions

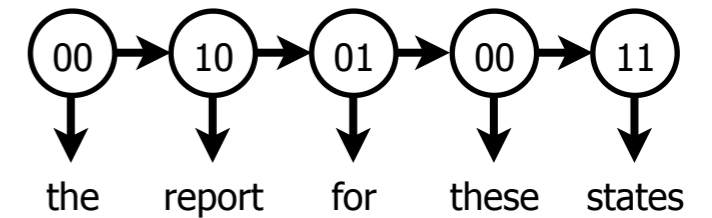
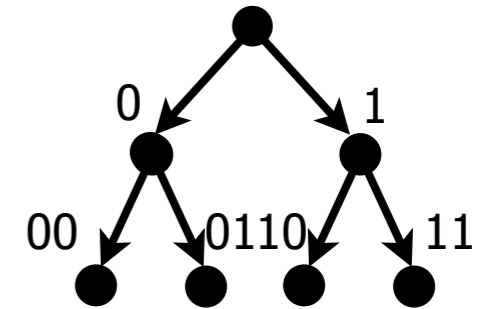
Conclusions

- Hierarchical clusters enable fast coarse-to-fine translation by reducing the vocabulary size



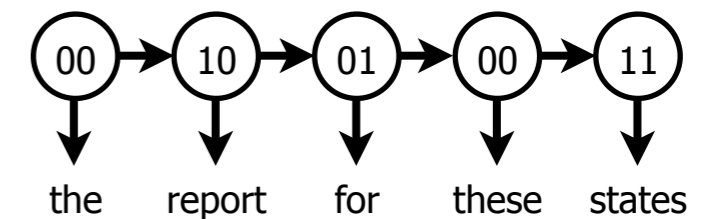
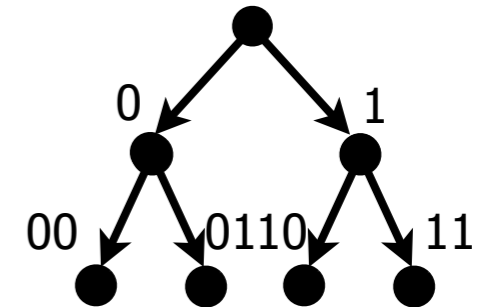
Conclusions

- Hierarchical clusters enable fast coarse-to-fine translation by reducing the vocabulary size
- Better clusters = better pruning



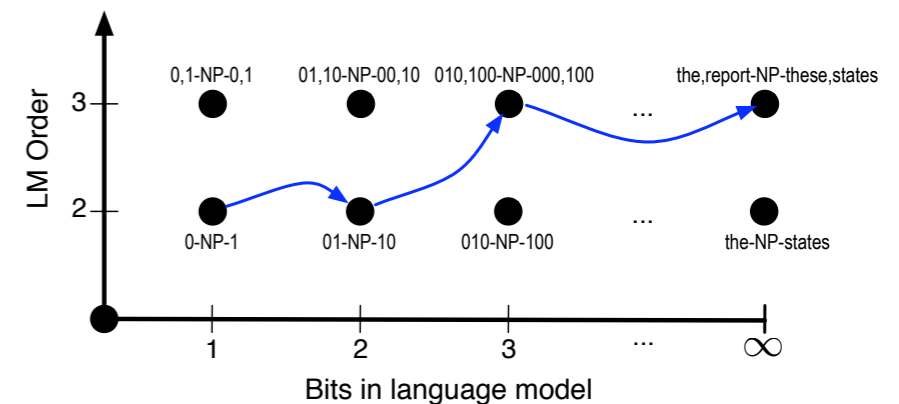
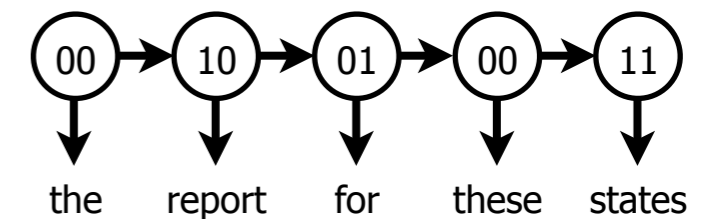
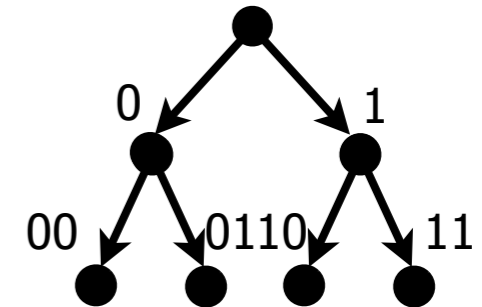
Conclusions

- Hierarchical clusters enable fast coarse-to-fine translation by reducing the vocabulary size
- Better clusters = better pruning
- Multiple, simple passes give the largest speed-ups
- Cluster size and n-gram order projections are complementary

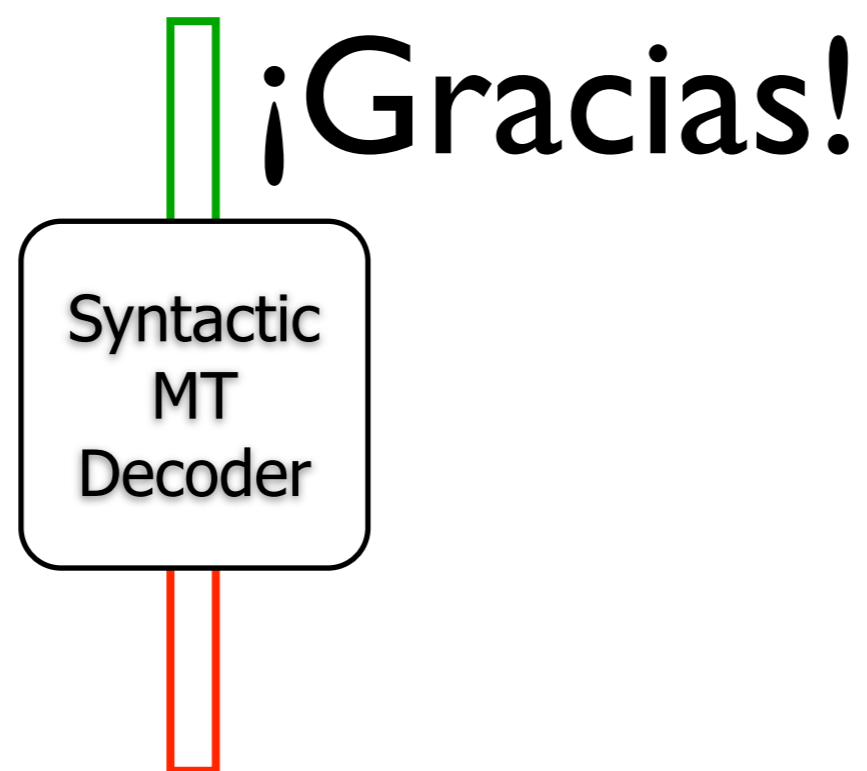


Conclusions

- Hierarchical clusters enable fast coarse-to-fine translation by reducing the vocabulary size
- Better clusters = better pruning
- Multiple, simple passes give the largest speed-ups
- Cluster size and n-gram order projections are complementary
- Minimal memory requirements



The End



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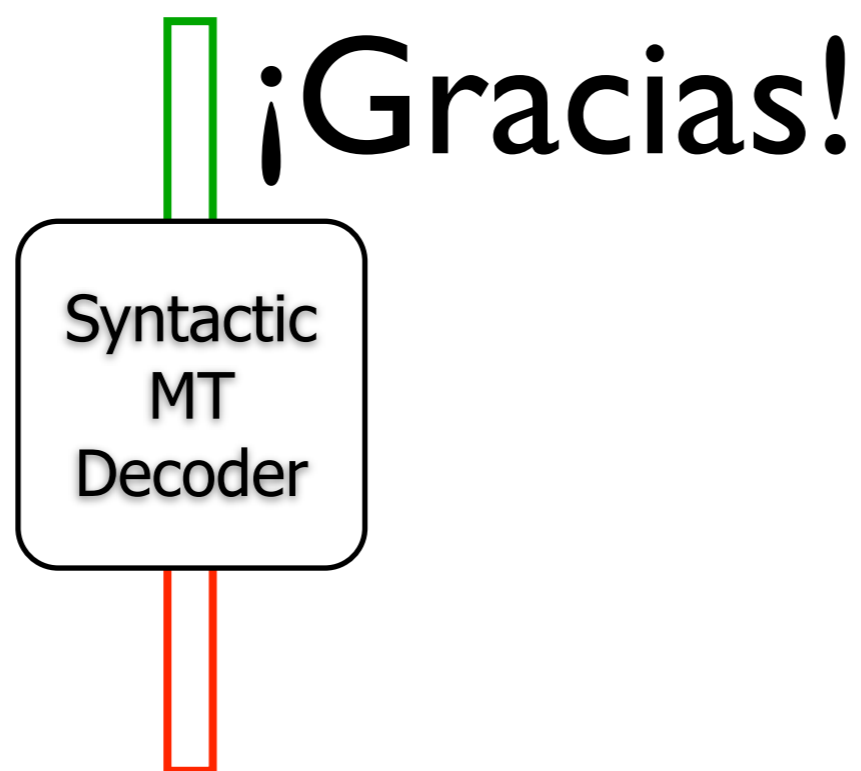
The End

¡Gracias!

Thanks!

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The End



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The End

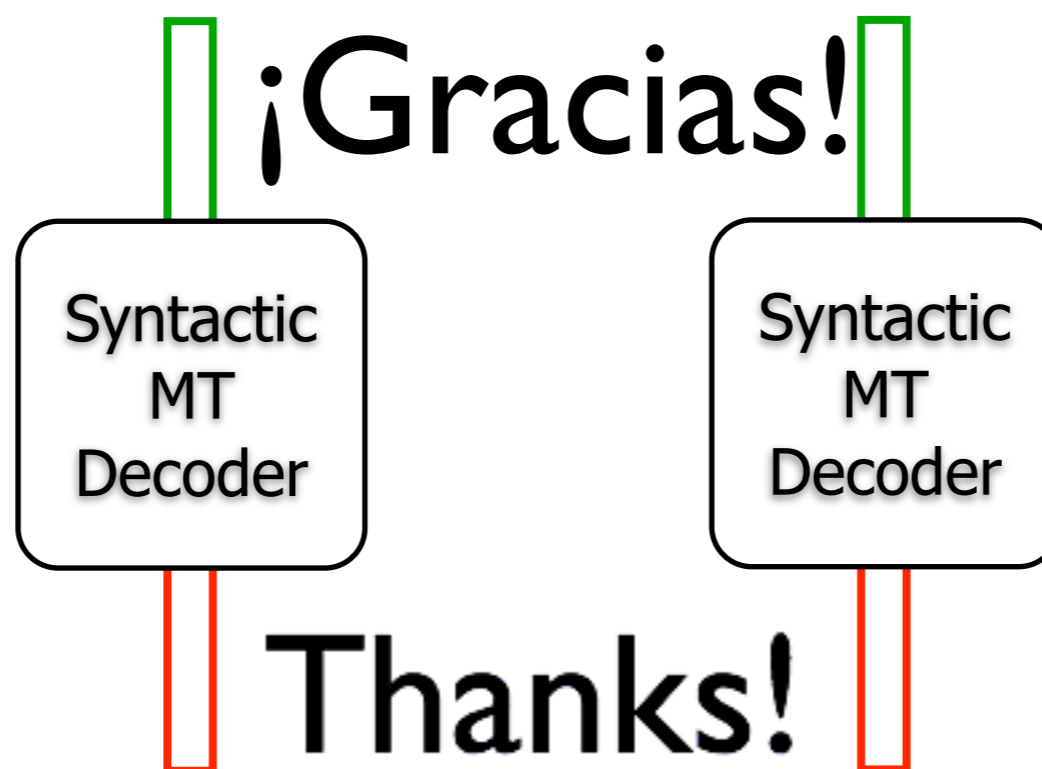
¡Gracias!

Syntactic
MT
Decoder

Thanks!

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The End



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