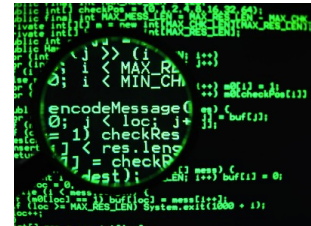
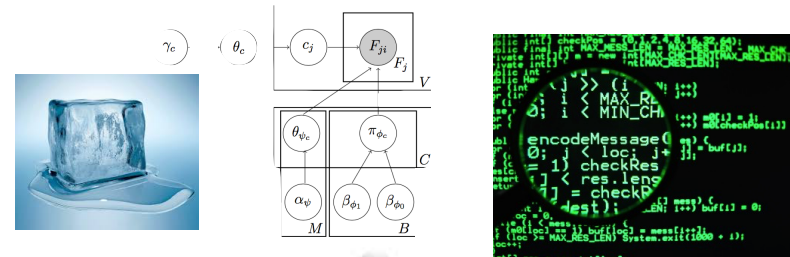
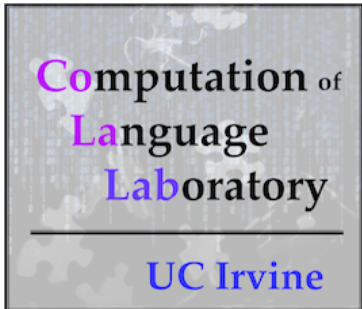


Integrating conceptual and structural cues: Theories for syntactic acquisition

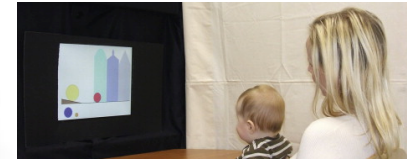
Lisa Pearl

University of California, Irvine



done-to
The ice melted.
The penguin climbed.

doer



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)



Sept 17, 2016: SynLinks 2016

University of Connecticut, Storrs

Syntactic acquisition

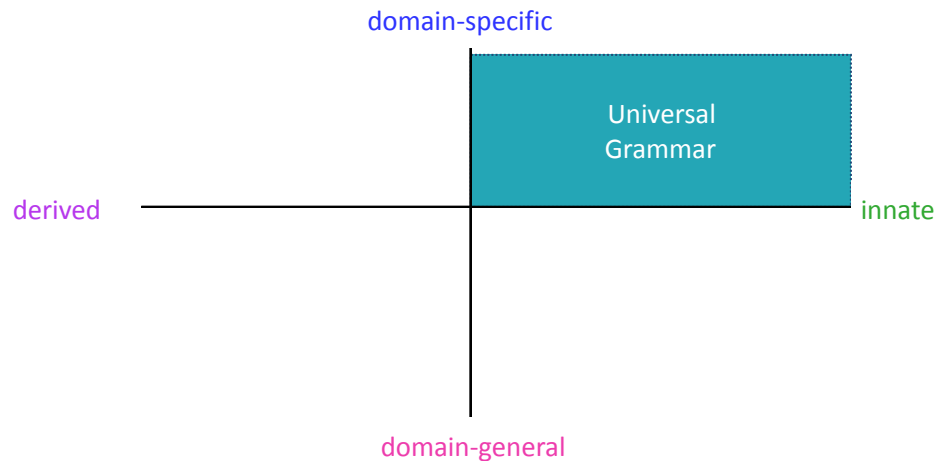
While syntactic acquisition is (by definition) about learning linguistic structure, children use information of different kinds in order to accomplish it.



Syntactic acquisition

While syntactic acquisition is (by definition) about learning linguistic structure, children use information of different kinds in order to accomplish it.

prior knowledge (Universal Grammar or otherwise)

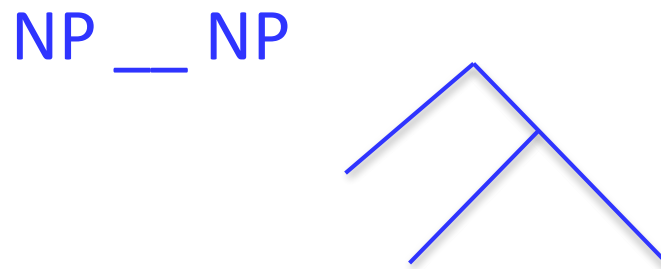


Syntactic acquisition

While syntactic acquisition is (by definition) about learning linguistic structure, children use information of different kinds in order to accomplish it.

prior knowledge (Universal Grammar or otherwise)

syntactic cues



Syntactic acquisition

While syntactic acquisition is (by definition) about learning linguistic structure, children use information of different kinds in order to accomplish it.

prior knowledge (Universal Grammar or otherwise)

syntactic cues

conceptual cues

+animate



-animate



Syntactic acquisition

While syntactic acquisition is (by definition) about learning linguistic structure, children use information of different kinds in order to accomplish it.

prior knowledge (Universal Grammar or otherwise)

syntactic cues

conceptual cues

semantic-syntactic cues

She melted the ice with a blow dryer.

Subject
Agent

Object
Patient

Indirect Object
Instrument



Syntactic acquisition

Given this, it seems useful to consider learning theories that leverage these different information types.

prior knowledge (Universal Grammar or otherwise)

syntactic cues

conceptual cues

semantic-syntactic cues



It's even more useful to be concrete, so let's look at a specific case study:

The Linking Problem (where event participants appear syntactically)

Today's plan

Linking Problem overview & some theories for handling it



done-to

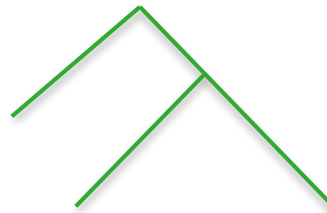
The ice melted.

The penguin climbed.

doer



Theory evaluation with computational modeling: A primer

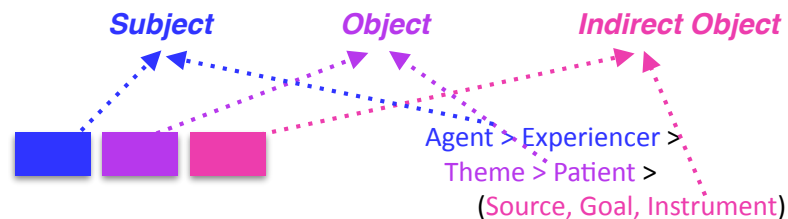


The ice melted.
What happened?
The ground's shaking.



The penguin climbed.
Who laughed?
She's winking.

Theory evaluation: The Linking Problem



Today's plan

Linking Problem overview & some theories for handling it



done-to

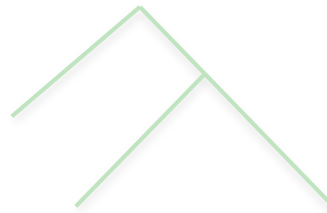
The **ice** melted.

The **penguin** climbed.

doer



Theory evaluation with computational modeling: A primer

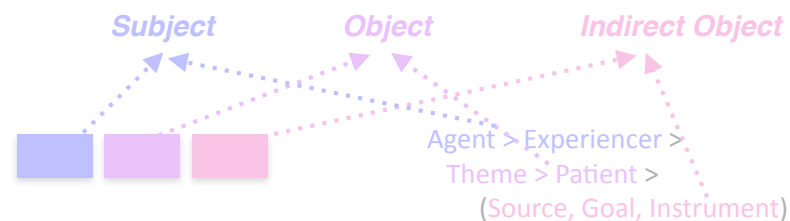


The ice **melted**.
What **happened**?
The ground's **shaking**.



The penguin **climbed**.
Who **laughed**?
She's **winking**.

Theory evaluation: The Linking Problem



The Linking Problem

- **Why?** About how conceptual information maps to syntactic structure, and we have some proposals for how to capture the empirical facts (e.g., (r)UTAH, Case Theory)

The Linking Problem

- **Why?** About how conceptual information maps to syntactic structure, and we have some proposals for how to capture the empirical facts (e.g., (r)UTAH, Case Theory)
- **What?** Predicates such as verbs allow a variety of syntactic options for where and how their arguments appear and **each predicate has certain linguistic patterns of behavior.**

She **melted** the ice.
doer *done-to*

The ice **melted**.
done-to

The ice was **melted**.
done-to



She **tried** to **melt** the ice.
doer *doer* *done-to*

*It **tried** that she **melted** the ice.
doer *done-to*

The penguin **climbed** the hill.
doer *done-to*

The penguin **climbed**.
doer

The hill was **climbed**.
done-to



The penguin **seemed** to **climb** the hill.
doer *done-to*

It **seemed** that the penguin **climbed** the hill.
doer *done-to*

The Linking Problem: Acquisition

One way to figure out how a **new predicate** will behave is to determine **what kind of predicate** it is (i.e., **what predicate category** it belongs to) with the idea that predicates in the same category behave similarly.

unaccusative

She **melted** the ice.
doer *done-to*

The ice **melted**.
done-to

The ice was **melted**.
done-to



control

She **tried** to **melt** the ice.
doer *doer* *done-to*

*It **tried** that she **melted** the ice.
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The penguin **climbed** the hill.
doer *done-to*

The penguin **climbed**.
doer

The hill was **climbed**.
done-to



The penguin **seemed** to **climb** the hill.
doer *done-to*

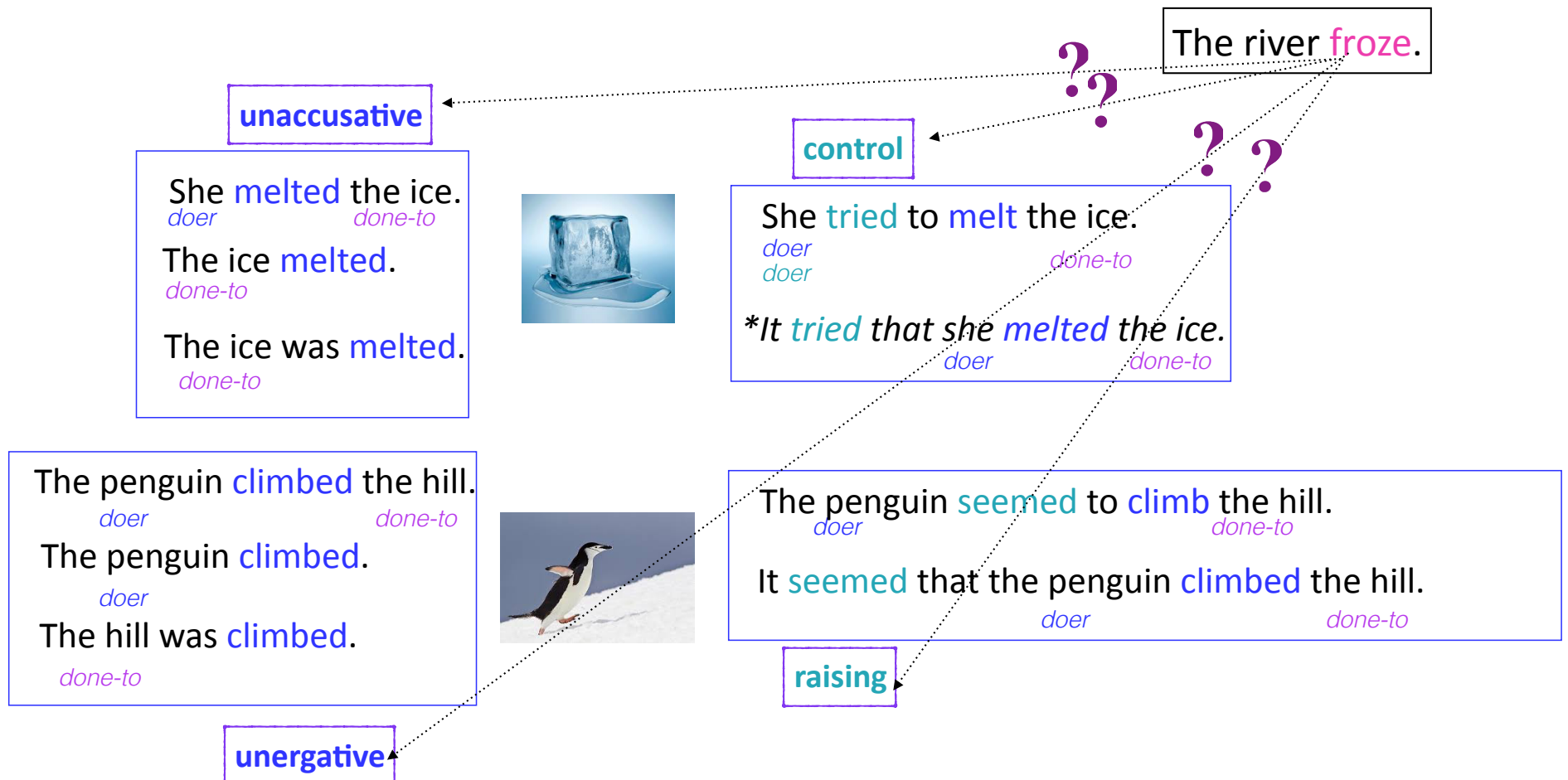
It **seemed** that the penguin **climbed** the hill.
doer *done-to*

raising

unergative

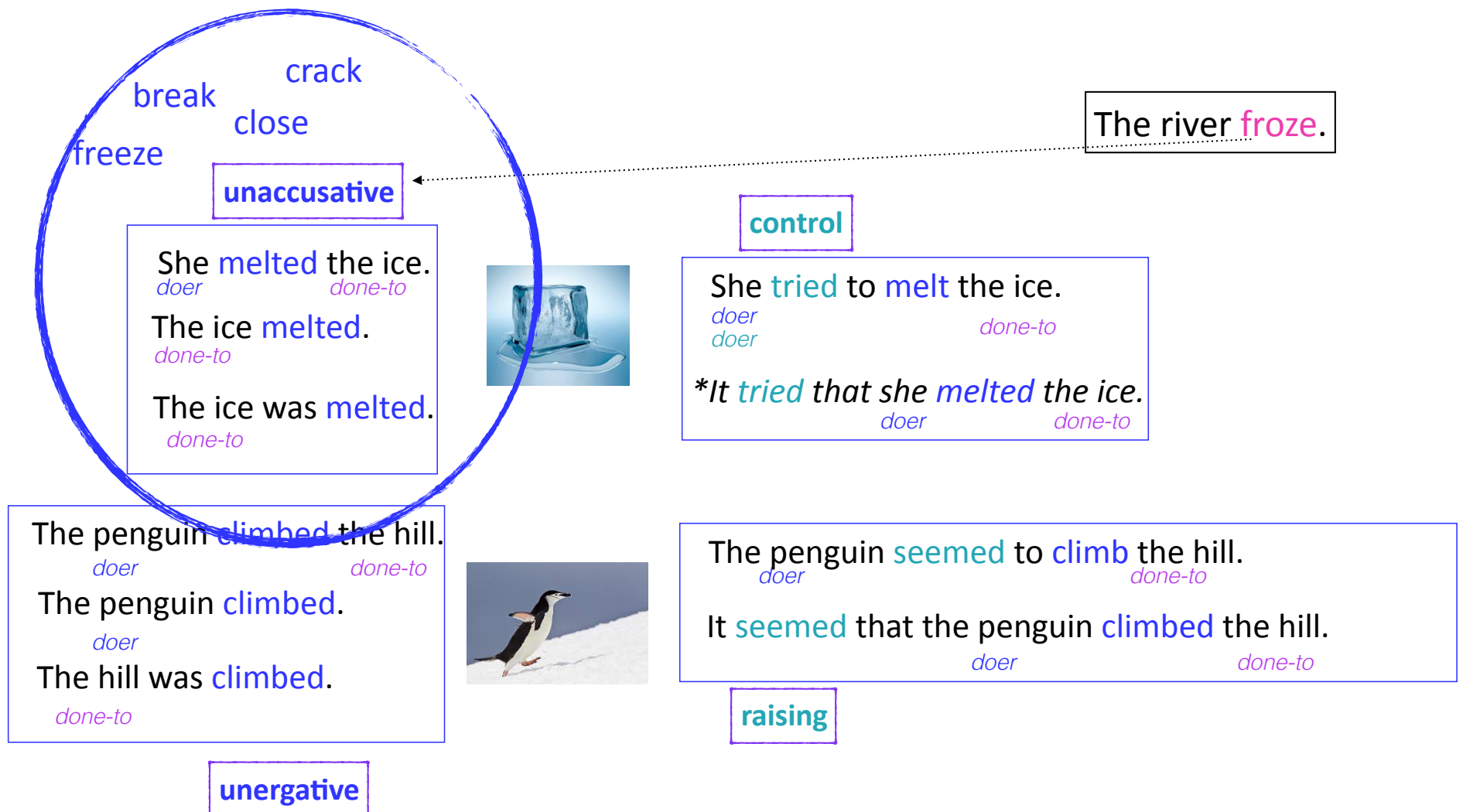
The Linking Problem: Acquisition

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The Linking Problem: Acquisition

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The Linking Problem: Acquisition

Knowledge transfer: Once you figure out how one predicate in the category behaves, you know something about **how all the predicates in the category behave**. This helps you predict how the conceptual arguments will surface syntactically for that new predicate.



The ice froze.
done-to

break crack
freeze close freeze

unaccusative

She melted the ice.
doer done-to

The ice melted.
done-to

The ice was melted.
done-to



control

She tried to melt the ice.
doer doer done-to

*It tried that she melted the ice.
doer done-to

The penguin climbed the hill.
doer done-to

The penguin climbed.
doer

The hill was climbed.
done-to



The penguin seemed to climb the hill.
doer done-to

It seemed that the penguin climbed the hill.
doer done-to

raising

unergative

The Linking Problem: Acquisition

Important foundation: Making **useful predicate categories**. What cues are available to do this?



break crack
 freeze close freeze

unaccusative

The ice froze.
done-to

She melted the ice.
doer *done-to*

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The ice was melted.
done-to



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She tried to melt the ice.
doer *doer* *done-to*

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The penguin climbed.
doer

The hill was climbed.
done-to



The penguin seemed to climb the hill.
doer *done-to*

It seemed that the penguin climbed the hill.
doer *done-to*

raising

unergative

The Linking Problem: Available cues

One type of cue: Syntactic cues

Example: Children are very adept at using **syntactic bootstrapping** to learn useful generalizations about how predicates behave (e.g., Fisher et al. 2010, Gutman et al. 2015, Harrigan et al. 2016).

Relevant cue: **syntactic structure**

unaccusative

She **melted** the ice.
doer *done-to*

The ice **melted**.
done-to

The ice was **melted**.
done-to



May be shallow “syntactic skeleton” (Gutman et al. 2015) that includes tense and aspect information or not.

The penguin **climbed** the hill.
doer *done-to*

The penguin **climbed**.
doer

The hill was **climbed**.
done-to



unergative

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+ some available tense and aspect information

She melted the ice → NP V_{past} NP

The ice melted → NP V_{past}

The ice was melted → NP V_{past_participle}

The ice was melting → NP V_{progressive_participle}

The penguin **climbed** the hill.
doer *done-to*

The penguin **climbed**.
doer

The hill was **climbed**.
done-to



unergative

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The ice was melting → NP V_{progressive_participle}

ignore available tense and aspect information

She melted the ice → NP V NP

The ice melted → NP V

The ice was melted → NP V

The ice was melting → NP V

The penguin **climbed** the hill.
doer *done-to*

The penguin **climbed**.
doer

The hill was **climbed**.
done-to



unergative

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doer *done-to*

The ice **melted**.
done-to

The ice was **melted**.
done-to



Why include tense and aspect information? Tenny's (1987 1994) Aspectual Interface Hypothesis suggests that aspect may be a useful cue to verb class (telic = **unaccusative**, atelic = **unergative**). Tense and aspect affect telicity and are sometimes easily observable in the morphology.

+ some available tense and aspect information

She melted the ice → NP V_{past} NP

The ice melted → NP V_{past}

The ice was melted → NP V_{past_participle}

The ice was melting → NP V_{progressive_participle}

ignore available tense and aspect information

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The penguin **climbed** the hill.
doer *done-to*

The penguin **climbed**.
doer

The hill was **climbed**.
done-to



unergative

The Linking Problem: Available cues

Another type of cue: Conceptual cues (non-linguistic)

Example: **Animacy** is useful for distinguishing predicate classes like **raising** vs. **control** verbs, and **psych-object-experiencer** verbs. Young children have been shown to use this cue in experimental studies (Becker 2009, Kirby 2009, Kirby 2010, Becker 2014, Becker 2015, Hartshorne et al. 2015).

-animate



control

She **tried** to **melt** the ice.

doer
doer

done-to

*It **tried** that she **melted** the ice.

doer

done-to

+animate



The penguin **seemed** to **climb** the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

The Linking Problem: Available cues

Another type of cue: Conceptual cues (non-linguistic)

Example: **Thematic roles** (e.g., *Agent*, *Patient*) that indicate participant roles in an event are salient to very young children [<10 months: Gordon 2003; 6 months: Hamlin, Wynn, & Bloom 2007, Hamlin, Wynn, Bloom, & Mahajan 2011].



-animate



control

She **tried** to **melt** the ice.

doer
doer

done-to

*It **tried** that she **melted** the ice.

doer

done-to

+animate



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done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

Thematic roles & how to use them: semantic-syntactic information

Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object



How do we get from here to here?



thematic-roles

(likely derived from lower-level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Thematic roles & how to use them: semantic-syntactic information



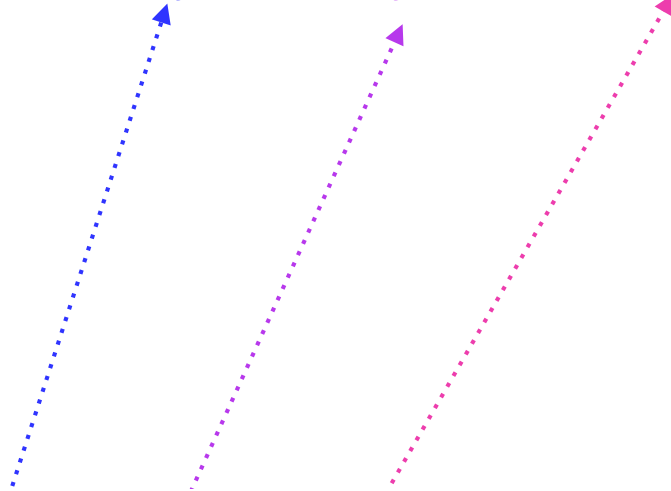
Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object



UG knowledge

Mapping to Syntax

The **U**niformity of **T**heta **A**ssignment **H**ypothesis:

Baker 1988, Baker 1997, Dowty 1991, Fillmore 1968, Grimshaw 1990, Jackendoff 1987, Perlmutter & Postal 1984, Speas 1990

UTAH

Intermediate representations



Thematic roles map to one of three categories.

(likely derived from lower-level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

thematic-roles



Thematic roles & how to use them: semantic-syntactic information



Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object

Mapping to Syntax

UG knowledge

The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis:
Larson 1988, Larson 1990

UTAH

rUTAH

Intermediate representations



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

Thematic roles map to one of three categories.

Thematic roles are ordered with respect to each other.

thematic-roles

(likely derived from lower-level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Thematic roles & how to use them: semantic-syntactic information



Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object

Standard UTAH and rUTAH implementations typically assume this part is included.

Mapping to Syntax

UG knowledge

UTAH

rUTAH

Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

Intermediate representations

Thematic roles map to one of three categories.

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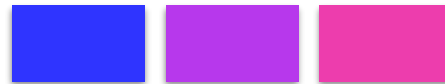
Thematic roles & how to use them: semantic-syntactic information

If children **expect the mapping to hold**, it may be especially salient to them when it doesn't.
Such instances would be accounted for by movement.

UG knowledge

UTAH

rUTAH



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

+exp-mapping



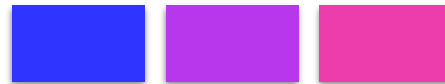
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UTAH

rUTAH



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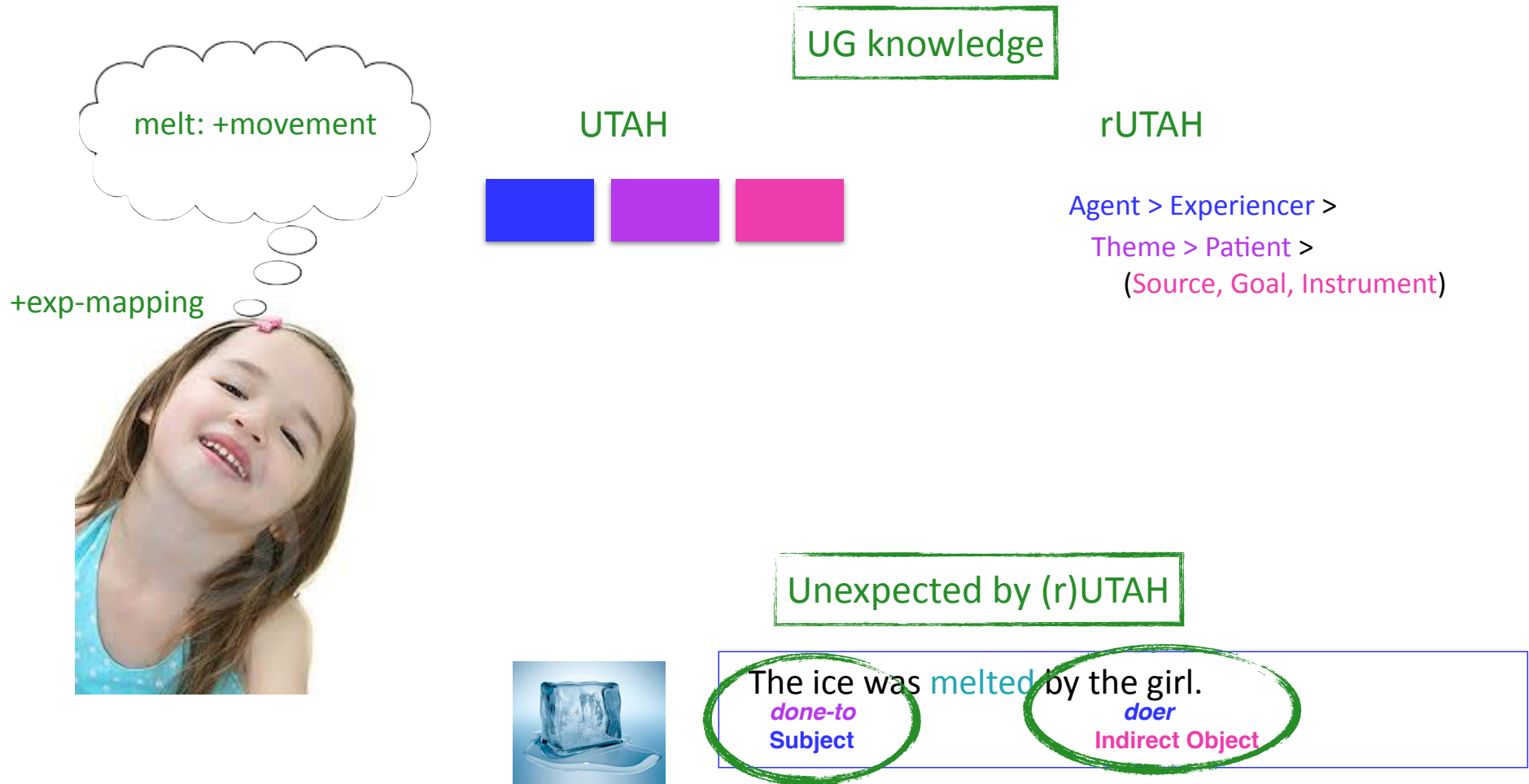
The ice was melted by the girl.

done-to
Subject

doer
Indirect Object

Thematic roles & how to use them: semantic-syntactic information

If children **expect the mapping to hold**, it may be especially salient to them when it doesn't. Such instances would be accounted for by movement.



Thematic roles & how to use them: semantic-syntactic information

Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object



But we could also look at implementations that don't assume this mapping is fixed a priori. This would be a weaker version of standard (r)UTAH implementations.

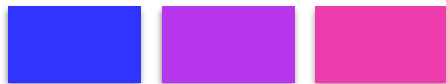
Mapping to Syntax

UG knowledge

UTAH

rUTAH

Intermediate representations



Agent > Experiencer >
Theme > Patient >
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Thematic roles map to one of three categories.

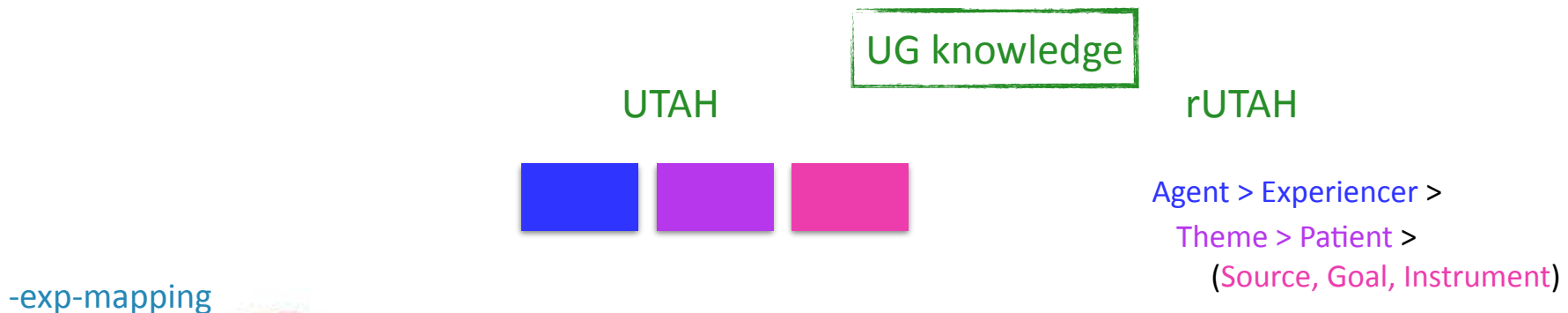
Thematic roles are ordered with respect to each other.

thematic-roles

(likely derived from lower level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Thematic roles & how to use them: semantic-syntactic information

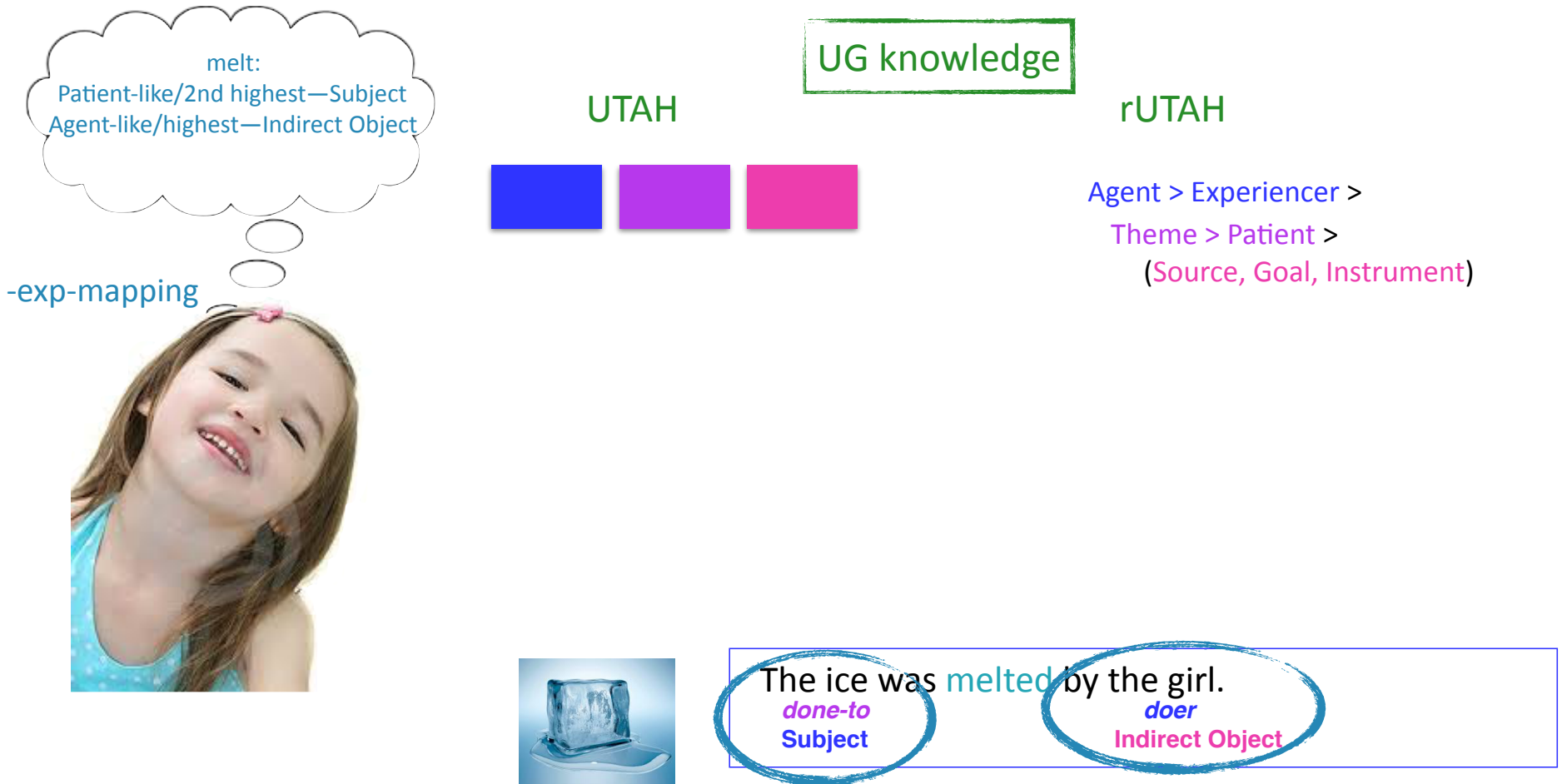
Alternatively, children could simply track the distributions of where intermediate representation roles appear with respect to grammatical positions. (No absolute expectation yet that the mapping will hold. This is something children would have to infer through exposure to the input.)



The ice was melted by the girl.
done-to *doer*
Subject Indirect Object

Thematic roles & how to use them: semantic-syntactic information

Alternatively, children could simply track the distributions of where intermediate representation roles appear with respect to grammatical positions. (No absolute expectation yet that the mapping will hold. This is something children would have to infer through exposure to the input.)



Thematic roles & how to use them: semantic-syntactic information

Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object



+exp-mapping:
movement is salient
because mapping to
syntax is fixed

-exp-mapping:
syntax mapping
distributions are
tracked

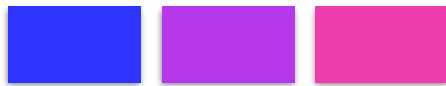
Mapping to Syntax

UG knowledge

UTAH

rUTAH

Intermediate
representations



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

Thematic roles map to one
of three categories.

Thematic roles are ordered
with respect to each other.

thematic-roles

(likely derived from lower level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Thematic roles & how to use them: semantic-syntactic information

Syntax

She melted the ice with a blow dryer.

Subject

Object

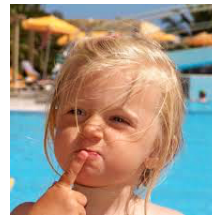
Indirect Object



+exp-mapping:
movement is salient
because mapping to
syntax is fixed

Choice 2

-exp-mapping:
syntax mapping
distributions are
tracked



UG knowledge

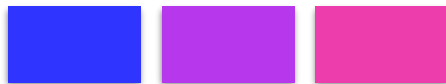
Mapping to Syntax

UTAH

Choice 1

rUTAH

Intermediate
representations



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

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of three categories.

Thematic roles are ordered
with respect to each other.

thematic-roles

(likely derived from lower-level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Potential learning strategies

UG knowledge options

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping



Potential learning strategies



UG knowledge options

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

Additional learner information: Syntactic options (+/- tense & aspect in the surface morphology)

+ some available tense and aspect information

She melted the ice → NP V_{past} NP

The ice melted → NP V_{past}

The ice was melted → NP V_{past_participle}

The ice was melting → NP V_{progressive_participle}

ignore available tense and aspect information

She melted the ice → NP V NP

The ice melted → NP V

The ice was melted → NP V

The ice was melting → NP V

Potential learning strategies



UG knowledge options

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

+ some available tense and aspect information

The ice was melted → NP V_{past_participle}

8 different learning strategy variants

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

ignore available tense and aspect information

The ice was melted → NP V

Potential learning strategies



UG knowledge options

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

+ some available tense and aspect information

The ice was melted → NP V_{past_participle}

8 different learning strategy variants

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

ignore available tense and aspect information

The ice was melted → NP V

All learners are sensitive to the animacy of NPs.

-animate



+animate



Learning strategy options

Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object

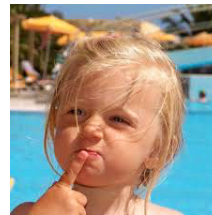


-tense/aspect info ← Choice 3 → +tense/aspect info

+exp-mapping:
movement is salient
because mapping to
syntax is fixed

Choice 2

-exp-mapping:
syntax mapping
distributions are
tracked

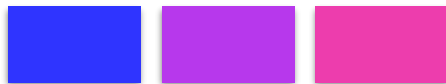


UG knowledge

Mapping to Syntax

UTAH ← Choice 1 → rUTAH

Intermediate representations



Agent > Experiencer >
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(Source, Goal, Instrument)

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(likely derived from lower-level conceptual info) =
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Today's plan

Linking Problem overview & some theories for handling it



done-to

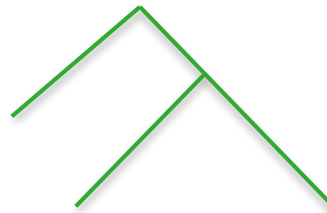
The ice melted.

The penguin climbed.

doer



Theory evaluation with computational modeling: A primer

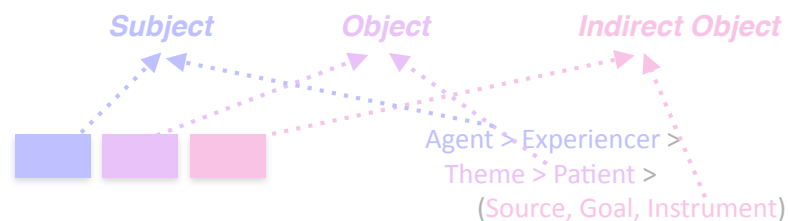


The ice melted.
What happened?
The ground's shaking.



The penguin climbed.
Who laughed?
She's winking.

Theory evaluation: The Linking Problem



Learning theory proposals: Generation & evaluation

How to **generate** a learning theory proposal:

Characterize the learning problem precisely and identify a potential solution.

8 different learning strategy variants



Learning theory proposals: Generation & evaluation

How to **generate** a learning theory proposal:

Characterize the learning problem precisely and identify a potential solution.

Benefit of **computational modeling**:

We can make sure the learning problem is **characterized precisely enough** to implement. It's not always obvious what pieces are missing until you try to build a model of the learning process.

(Pearl 2014, Pearl & Sprouse 2015)



Learning theory proposals: Generation & evaluation

How to **generate** a learning theory proposal:

Characterize the learning problem precisely and identify a potential solution.

How to **evaluate** a learning theory proposal:

See if it's **successful when embedded in a model of the acquisition process** for that learning problem.

Learning theory proposals: Generation & evaluation

How to **generate** a learning theory proposal:

Characterize the learning problem precisely and identify a potential solution.

How to **evaluate** a learning theory proposal:

See if it's **successful when embedded in a model of the acquisition process** for that learning problem.



Recently, in computational modeling, we've seen the integration of **rich hypothesis spaces with probabilistic/statistical learning mechanisms** (Sakas & Fodor 2001, Yang 2004, Pearl 2011, Dillon et al. 2013, Pearl & Sprouse 2013, Pearl et al. 2014, Pearl & Mis 2016, among many others).

Learning theory proposals: Generation & evaluation

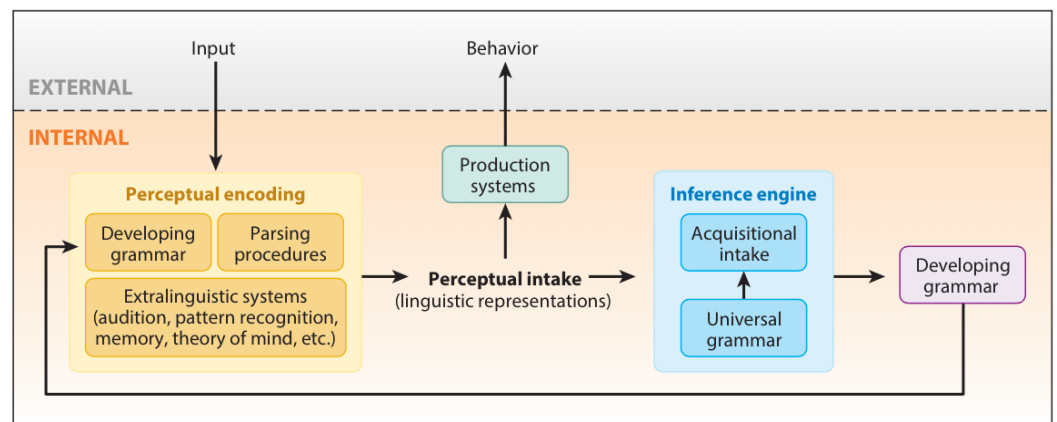
How to **generate** a learning theory proposal:

Characterize the learning problem precisely and identify a potential solution.

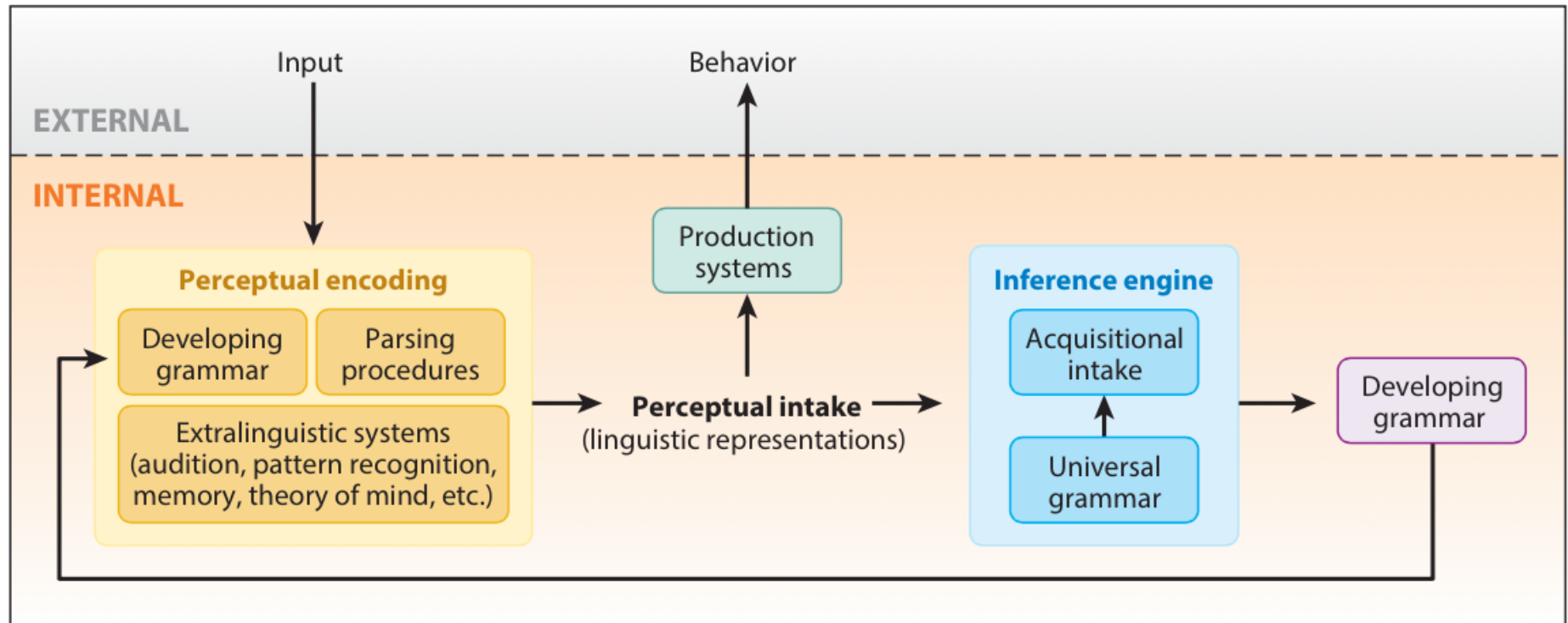
How to **evaluate** a learning theory proposal:

See if it's **successful when embedded in a model of the acquisition process** for that learning problem.

We've also seen the development of **more sophisticated acquisition frameworks** that highlight the precise role of different components (Lidz & Gagliardi 2015, Omaki & Lidz 2015).



The Lidz & Gagliardi (2015) acquisition framework



Lidz & Gagliardi 2015

Learning theory proposals: Generation & evaluation

How to **generate** a learning theory proposal:

Characterize the learning problem precisely and identify a potential solution.

How to **evaluate** a learning theory proposal:

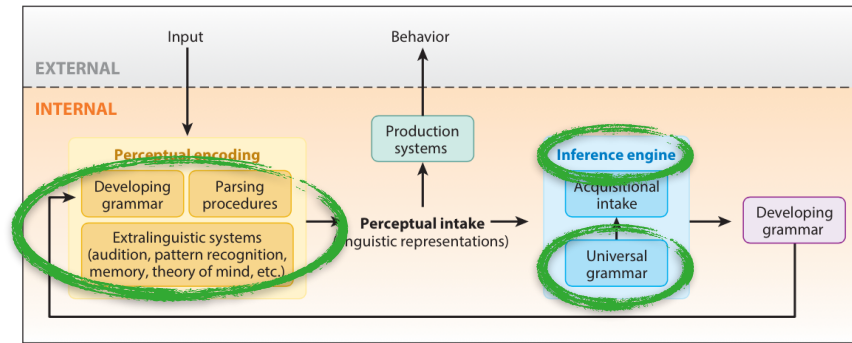
See if it's **successful when embedded in a model of the acquisition process** for that learning problem.

This computational modeling approach helps us refine our theories about both **the knowledge representation** the learning theory relies on and **the acquisition process that uses that representation**.



Characterizing learning problems

Initial state:



Lidz & Gagliardi 2015

Characterizing learning problems

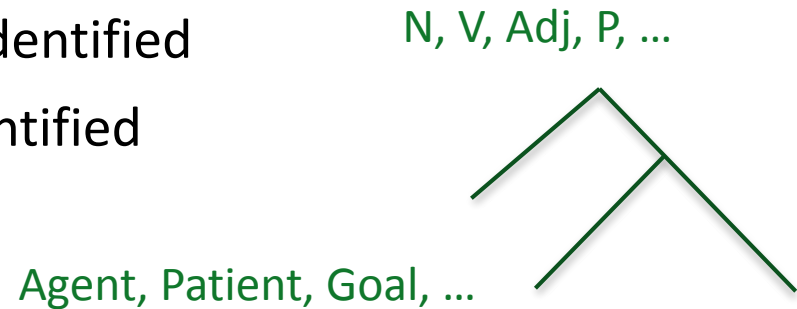
Initial state:

- initial knowledge state

ex: syntactic categories exist and can be identified

ex: phrase structure exists and can be identified

ex: participant roles can be identified



Characterizing learning problems

Initial state:

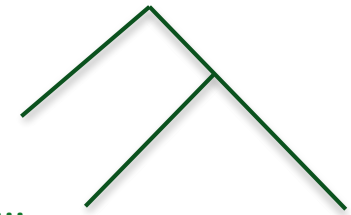
- initial knowledge state

ex: syntactic categories exist and can be identified

ex: phrase structure exists and can be identified

ex: participant roles can be identified

N, V, Adj, P, ...



Agent, Patient, Goal, ...

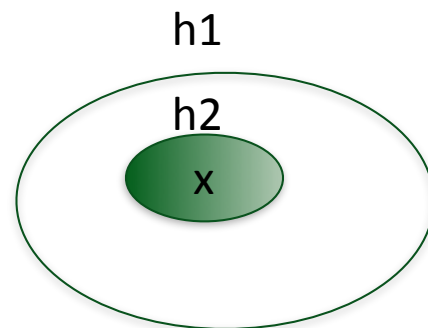
- learning biases & capabilities

ex: frequency information can be tracked

ex: distributional information can be leveraged

The ice melted. 😊
What happened?
The ground's shaking.

The penguin climbed. 😞
Who laughed?
She's winking.

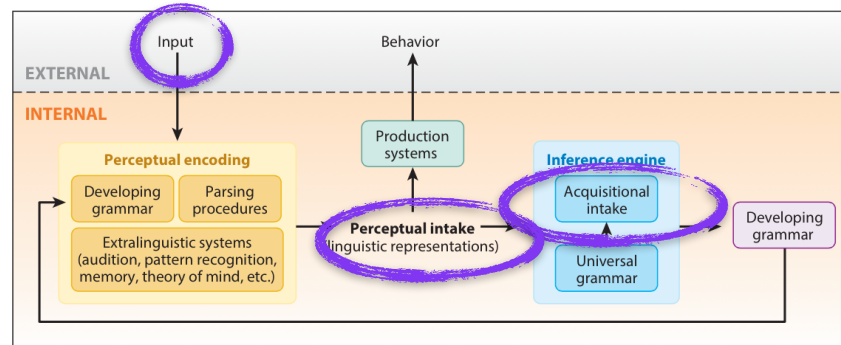


h2 more likely

Characterizing learning problems

Initial state: initial knowledge state + learning biases & capabilities

Data intake:



Lidz & Gagliardi 2015

Characterizing learning problems

Initial state: initial knowledge state + learning biases & capabilities

Data intake:

- input + encoding + acquisitional intake = data perceived as relevant for learning

(Fodor 1998, Lidz & Gagliardi 2015)

ex: syntactic and conceptual data for learning syntactic knowledge that links with conceptual knowledge

[defined by knowledge & biases/capabilities in the initial state]

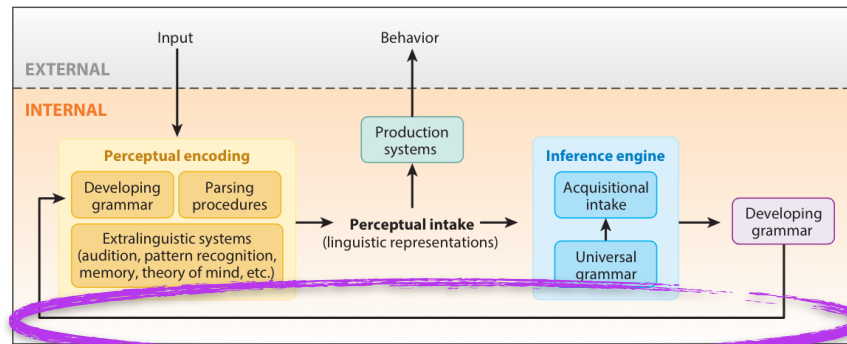


Characterizing learning problems

Initial state: initial knowledge state + learning biases & capabilities

Data intake: data perceived as relevant for learning

Learning period:



Lidz & Gagliardi 2015

Characterizing learning problems

Initial state: initial knowledge state + learning biases & capabilities

Data intake: data perceived as relevant for learning

Learning period:

- how long children have to reach the target knowledge state
(when inference & iteration happen)

ex: 3 years, ~1,000,000 data points

ex: 4 months, ~36,500 data points



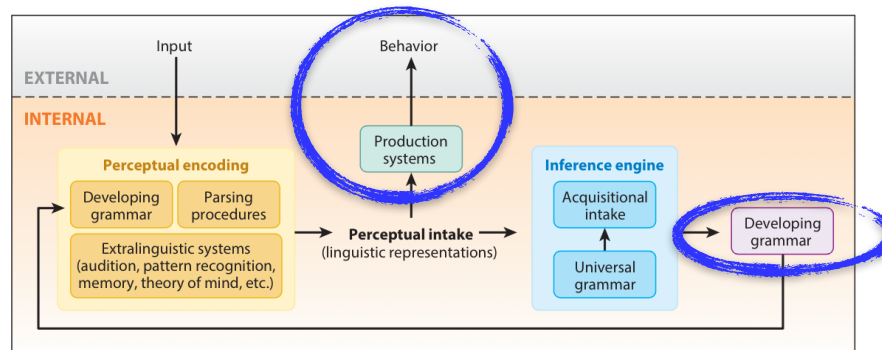
Characterizing learning problems

Initial state: initial knowledge state + learning biases & capabilities

Data intake: data perceived as relevant for learning

Learning period: how long children have to learn

Target state:



Lidz & Gagliardi 2015

Characterizing learning problems

Initial state: initial knowledge state + learning biases & capabilities

Data intake: data perceived as relevant for learning

Learning period: how long children have to learn

Target state:

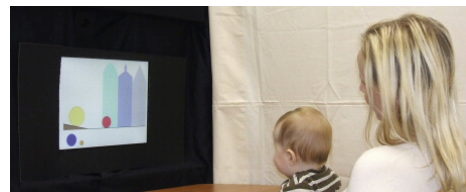
- the knowledge children are trying to attain (as indicated by their behavior)

ex: *done-to*

The ice melted.

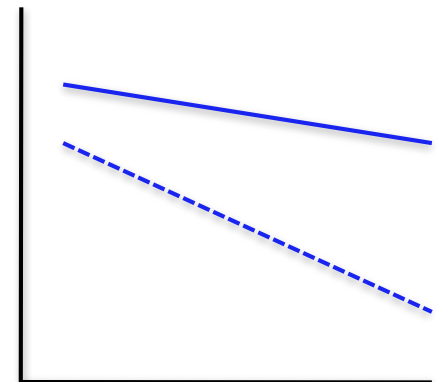
The penguin climbed.

doer



looking time preferences

z-score rating



Characterizing learning problems

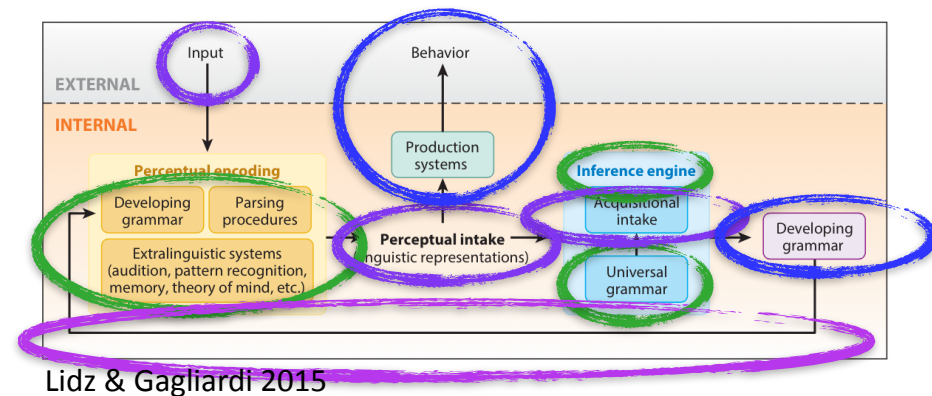
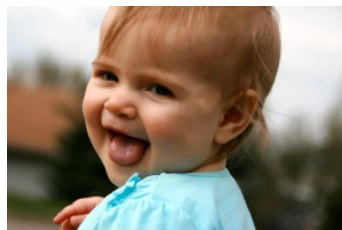
Initial state: initial knowledge state + learning biases & capabilities

Data intake: data perceived as relevant for learning

Learning period: how long children have to learn

Target state: the knowledge children must attain

Once we have all these pieces specified, we should be able to implement an informative model of the learning process.

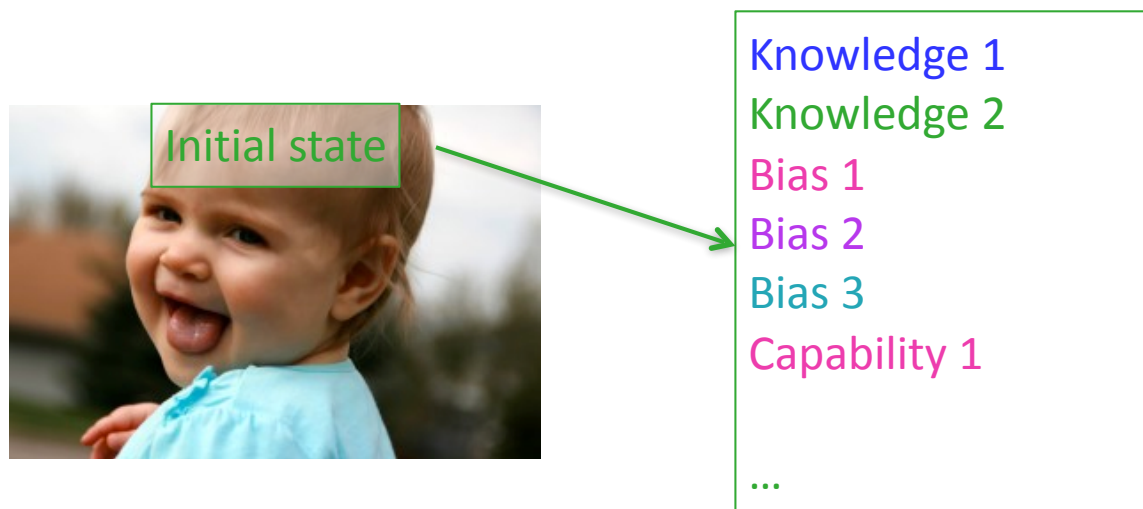


Pearl & Sprouse 2015, Pearl & Mis 2016

Informing theories of knowledge & learning

When we identify a successful learning strategy via modeling, this is an existence proof that children could solve that learning problem using **the knowledge, learning biases, and capabilities comprising that strategy**.

This identifies useful learning strategy components, which include both the knowledge components (= **theories of representation**) and the biases & capabilities that must exist for that knowledge to be successfully deployed during acquisition (= **theories of the learning process**).



Today's plan

Linking Problem overview & some theories for handling it



done-to

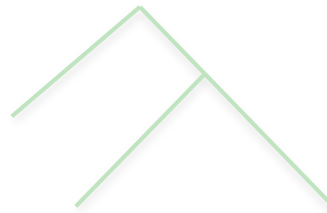
The ice melted.

The penguin climbed.

doer



Theory evaluation with computational modeling: A primer

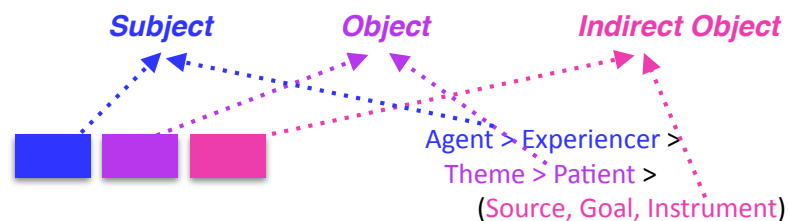


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She's winking.

Theory evaluation: The Linking Problem



Potential learning strategies revisited



UG knowledge options

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

+ some available tense and aspect information

The ice was melted → NP V_{past_participle}

8 different learning strategy variants

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

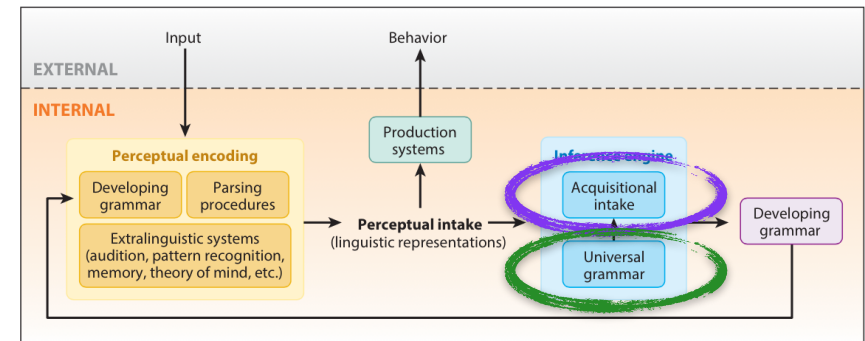
ignore available tense and aspect information

The ice was melted → NP V

-animate



+animate



Lidz & Gagliardi 2015

Initial state

The ability to identify and extract all relevant information reliably (syntactic + conceptual + semantic-syntactic cues) + sufficient statistical learning abilities to track and use this information.



UG knowledge options

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

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8 different learning strategy variants

UTAH, -exp-mapping

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rUTAH, +exp-mapping

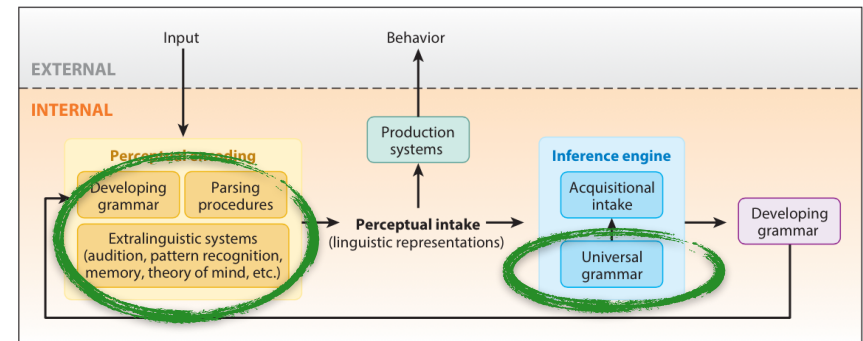
ignore available tense and aspect information

The ice was melted → NP V

-animate



+animate



Lidz & Gagliardi 2015

Initial state

The ability to identify and extract all relevant information reliably (syntactic + conceptual + semantic-syntactic cues) + sufficient statistical learning abilities to track and use this information.



UG knowledge options

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

+ some available tense and aspect information

The ice was melted → NP V_{past_participle}

8 different variants, which all cause different **acquisitional intakes**

UTAH, -exp-mapping

UTAH, +exp-mapping

rUTAH, -exp-mapping

rUTAH, +exp-mapping

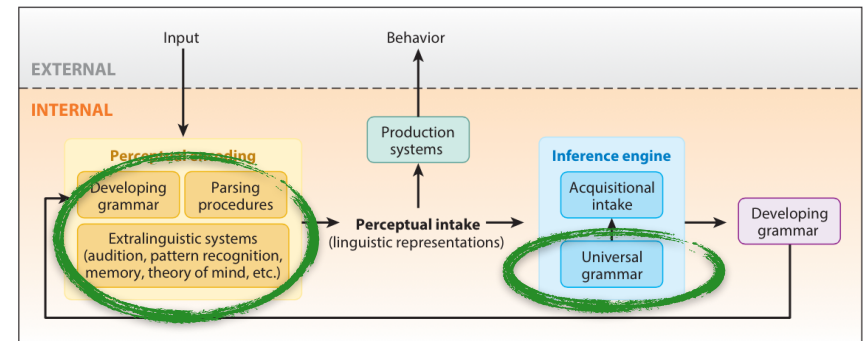
ignore available tense and aspect information

The ice was melted → NP V

-animate



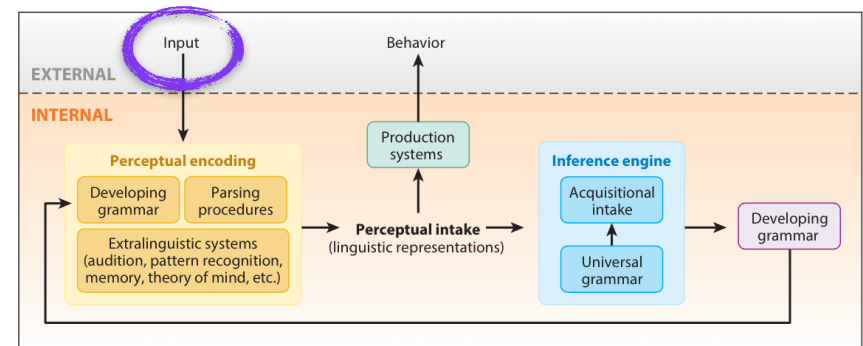
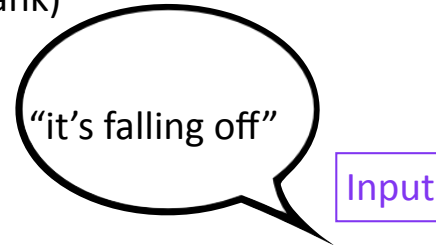
+animate



Lidz & Gagliardi 2015

Acquisitional intake options

(from Brown-Eve corpus from CHILDES Treebank)

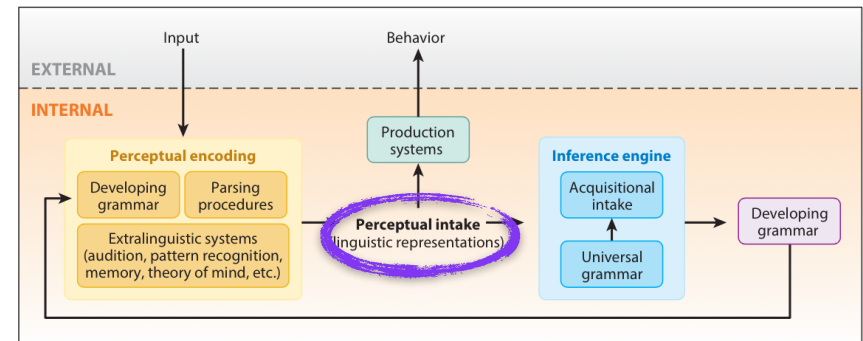
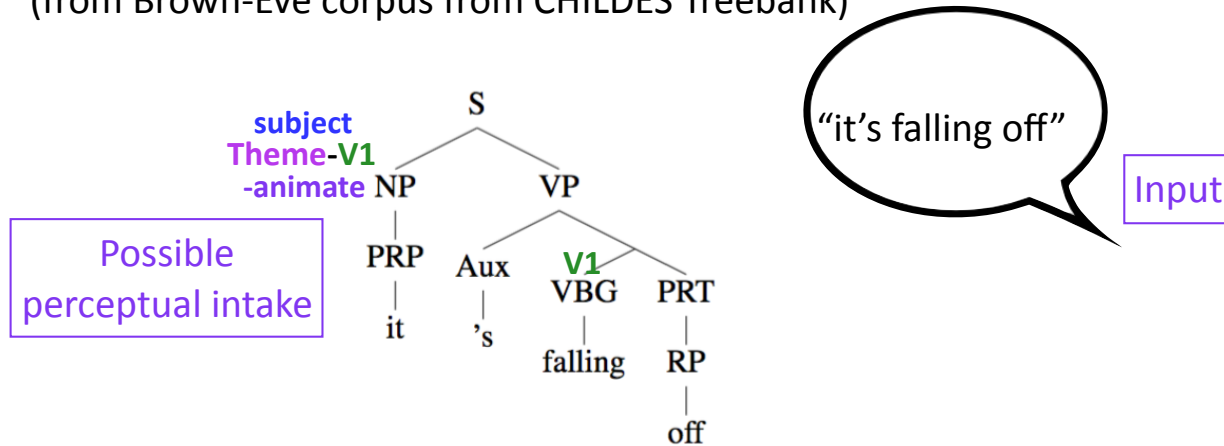


Lidz & Gagliardi 2015

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)

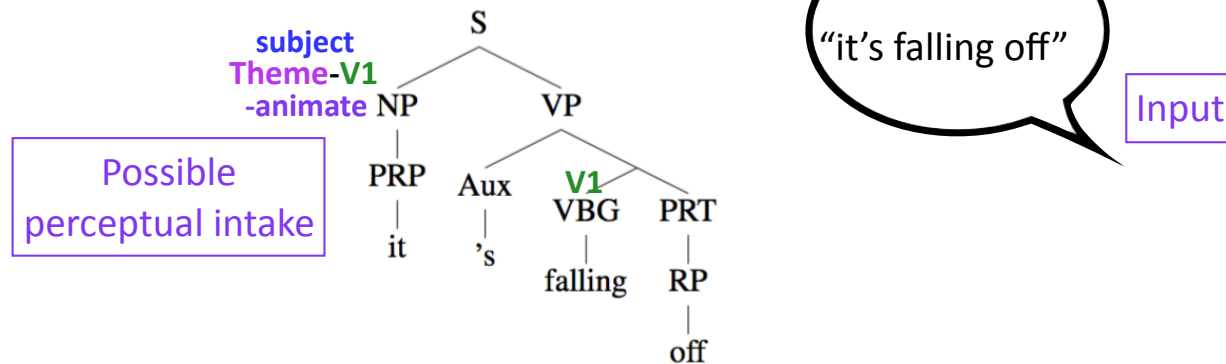


Lidz & Gagliardi 2015

Acquisitional intake options

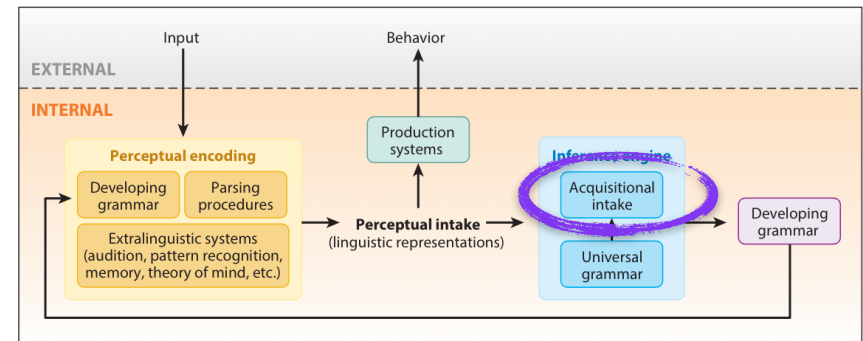


(from Brown-Eve corpus from CHILDES Treebank)



Acquisitional intake

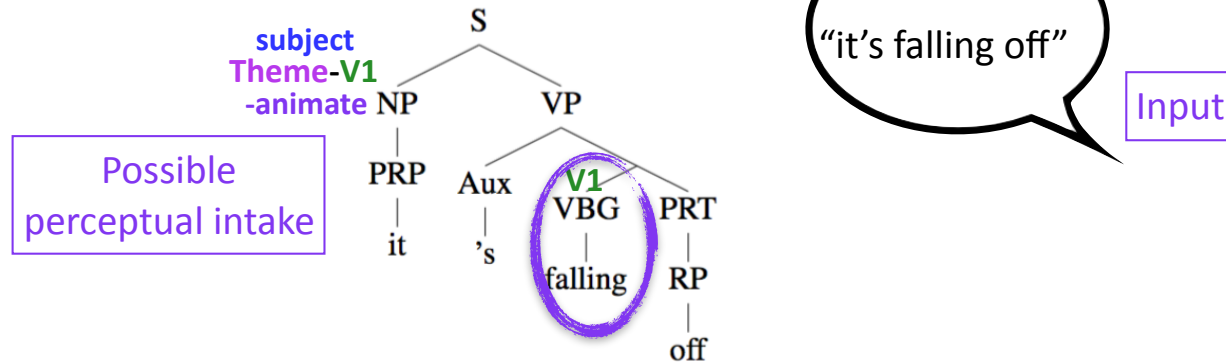
(1) UTAH, -exp-mapping, +some available tense and aspect information



Lidz & Gagliardi 2015

Acquisitional intake options

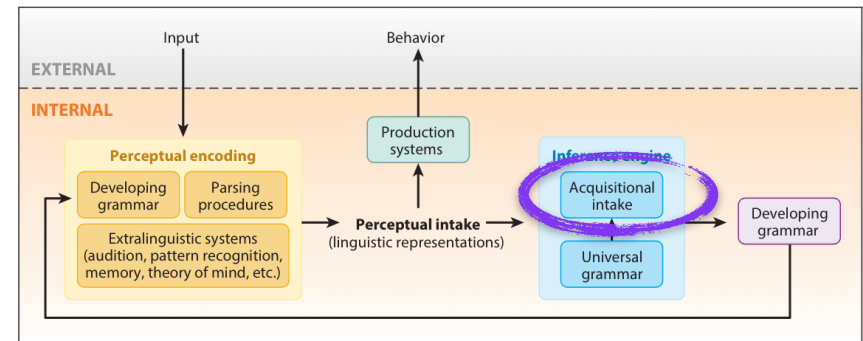
(from Brown-Eve corpus from CHILDES Treebank)



Acquisitional intake

(1) UTAH, -exp-mapping, +some available tense and aspect information

FALL



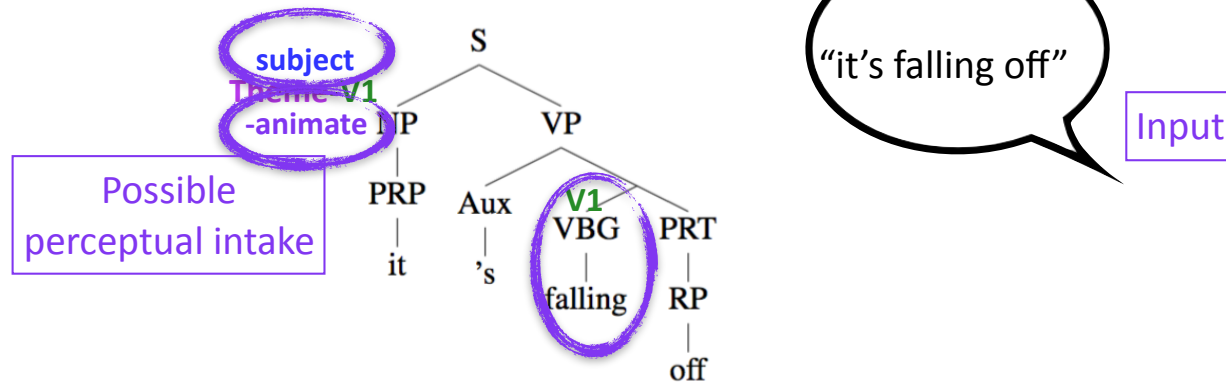
Lidz & Gagliardi 2015

Pearl & Sprouse in progress

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)

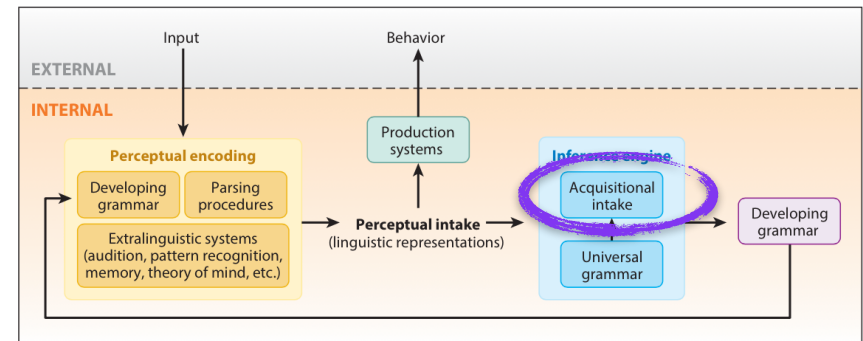


Acquisitional intake

(1) UTAH, -exp-mapping, +some available tense and aspect information

FALL

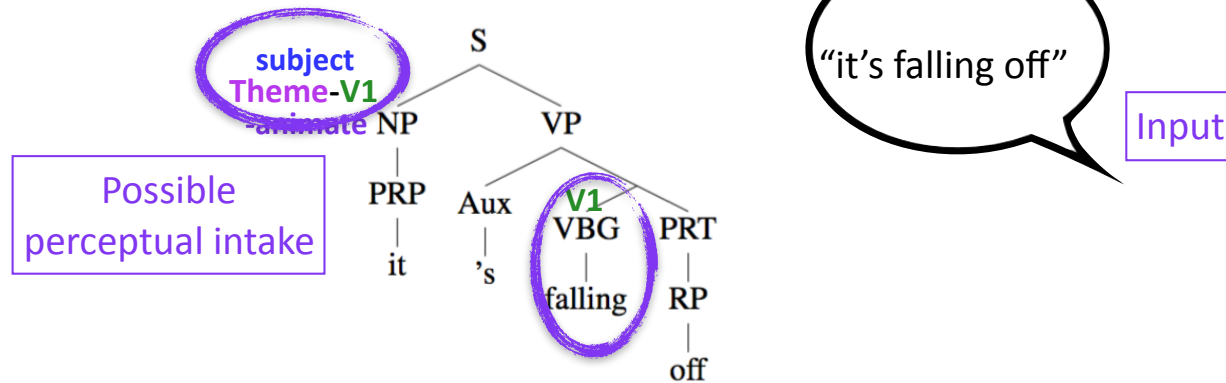
-animate subject: 1



Lidz & Gagliardi 2015

Acquisitional intake options

(from Brown-Eve corpus from CHILDES Treebank)



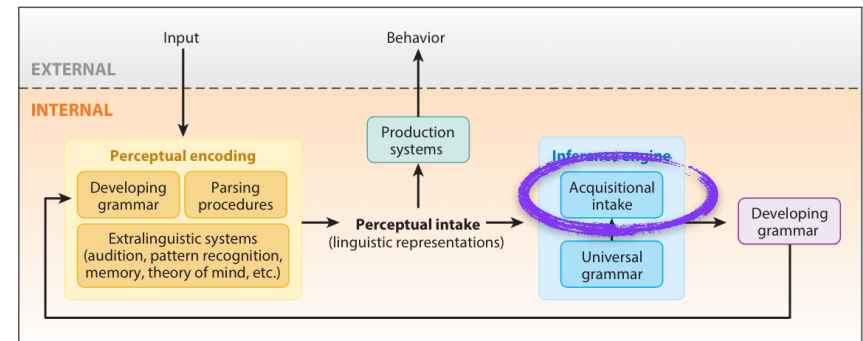
Acquisitional intake

(1) UTAH, -exp-mapping, +some available tense and aspect information

FALL

-animate subject: 1

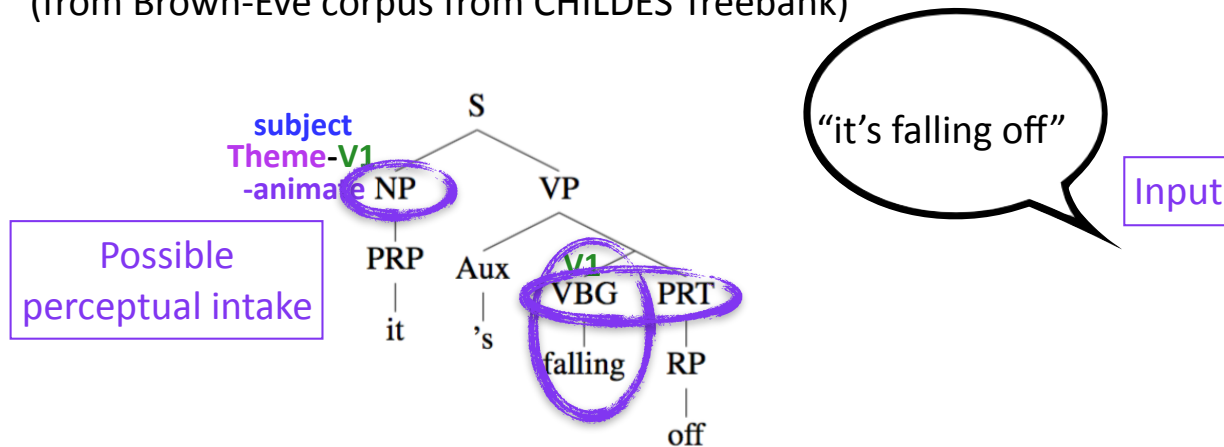
Patient-like as subject: 1



Lidz & Gagliardi 2015

Acquisitional intake options

(from Brown-Eve corpus from CHILDES Treebank)



Acquisitional intake

(1) UTAH, -exp-mapping, +some available tense and aspect information

FALL

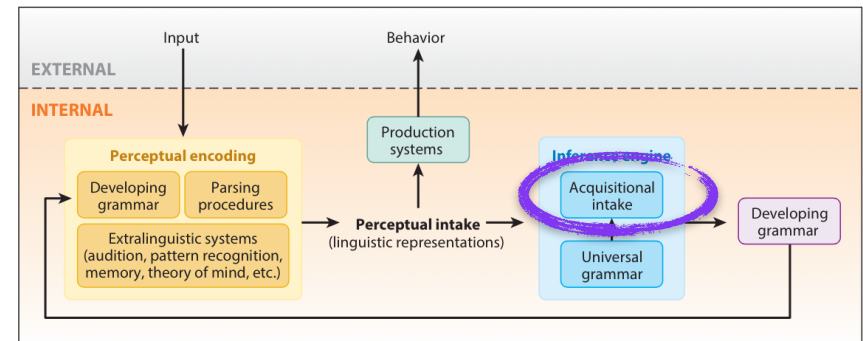
-animate subject: 1

Patient-like as subject: 1

NP V_{present_participle} PRT

Note: CHILDES Treebank syntactic encoding captures these distinctions:

- (i) present (VBP) vs. past tense (VBD)
- (ii) present participle (VBG) vs. past participle (VBN)
- (iii) non-finite usage (VB)

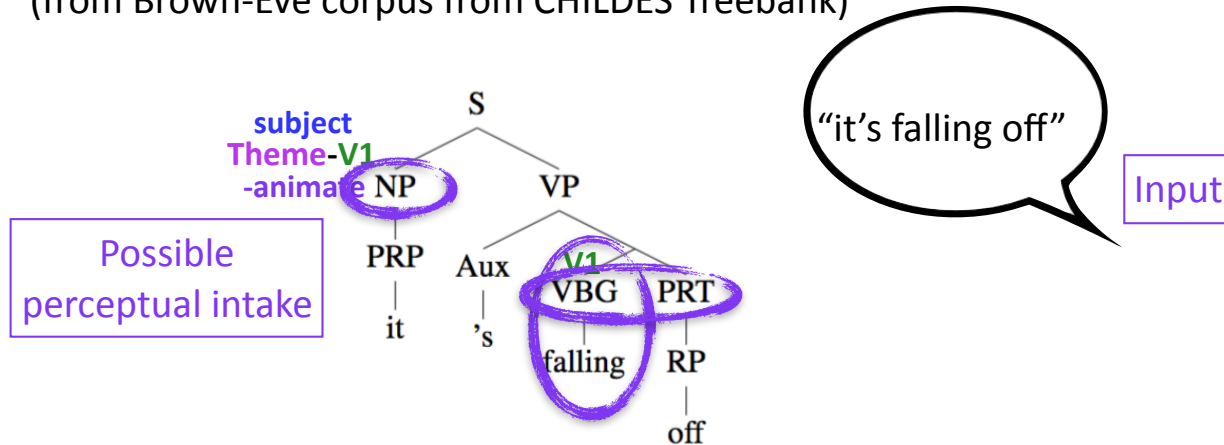


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Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)



Acquisitional intake

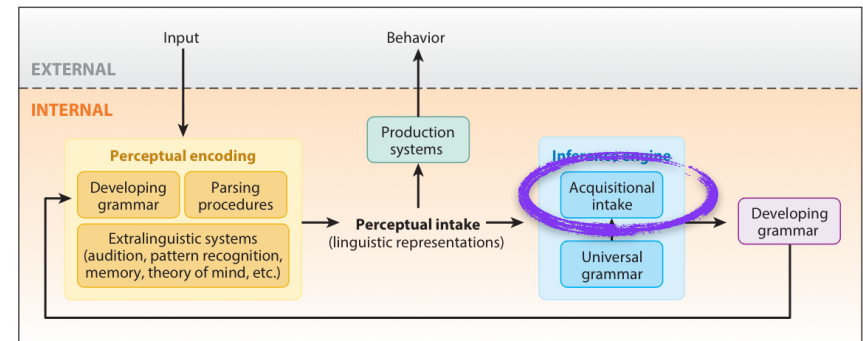
(2) UTAH, -exp-mapping, -some available tense and aspect information

FALL

-animate subject: 1

Patient-like as subject: 1

NP V PRT



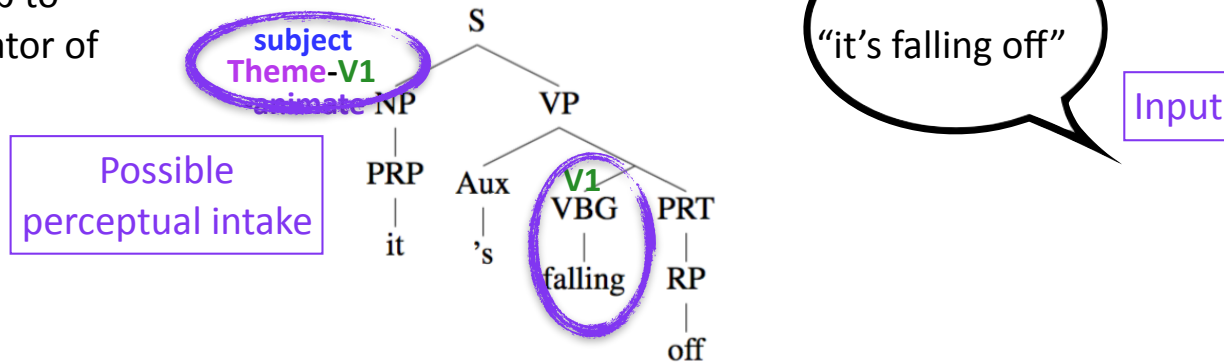
Lidz & Gagliardi 2015

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)

Theme is expected to map to object, not subject. Indicator of movement.



Acquisitional intake

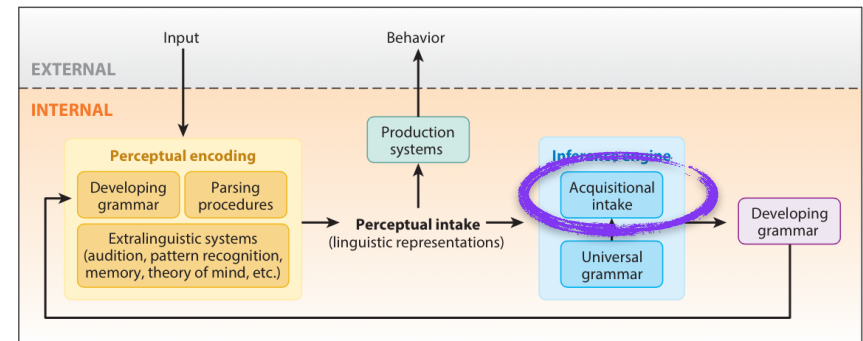
(3) UTAH, +exp-mapping, -some available tense and aspect information

FALL

-animate subject: 1

+movement: 1

NP V PRT

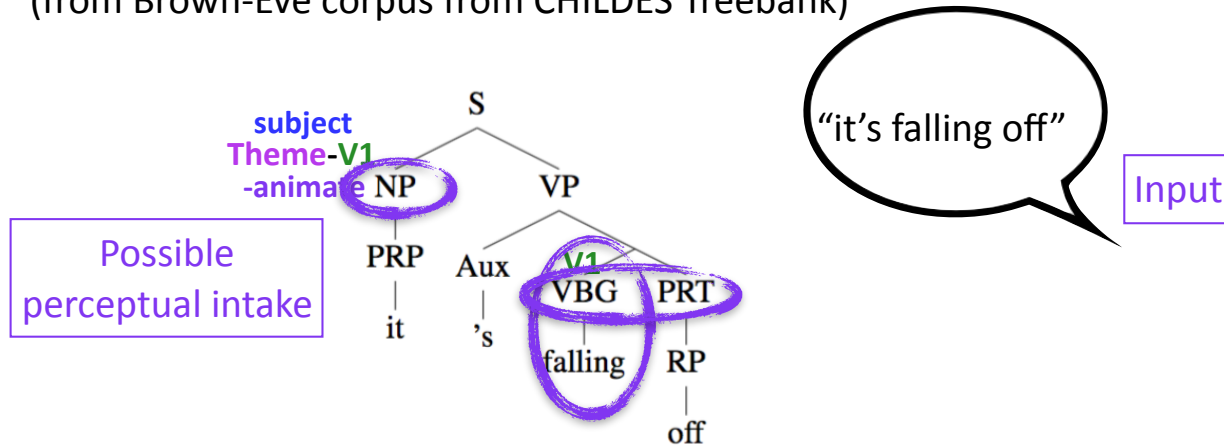


Lidz & Gagliardi 2015

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)



Acquisitional intake

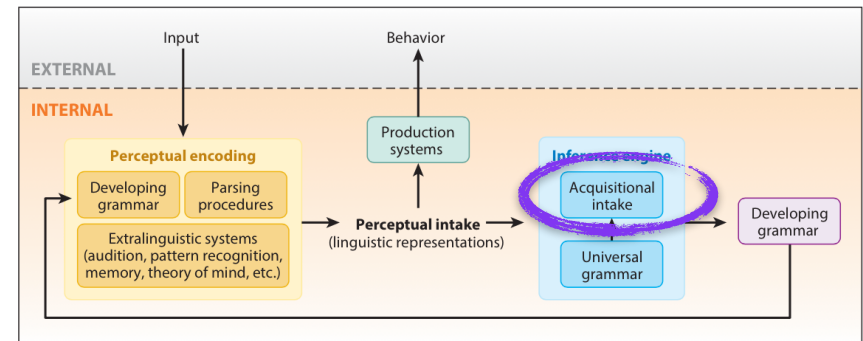
(4) UTAH, +exp-mapping, +some available tense and aspect information

FALL

-animate subject: 1

+movement: 1

NP V_{present_participle} PRT



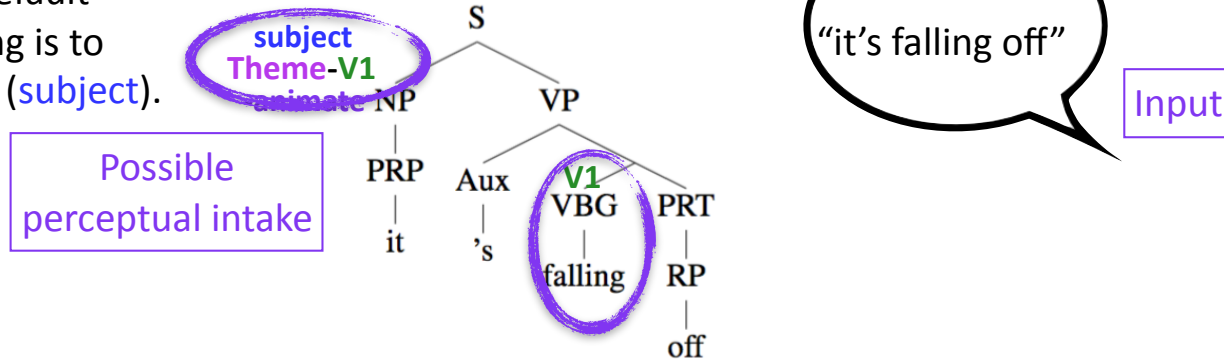
Lidz & Gagliardi 2015

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)

Theme is only role so is default highest. Expected mapping is to highest syntactic position (subject).



Acquisitional intake

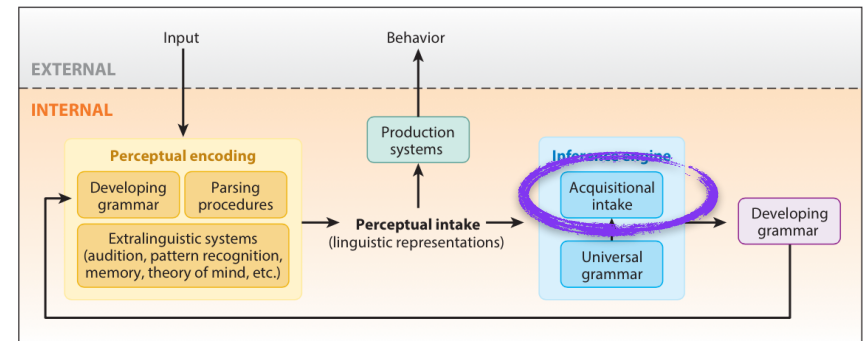
(5) rUTAH, +exp-mapping, +some available tense and aspect information

FALL

-animate subject: 1

+movement: 0

NP V_{present_participle} PRT

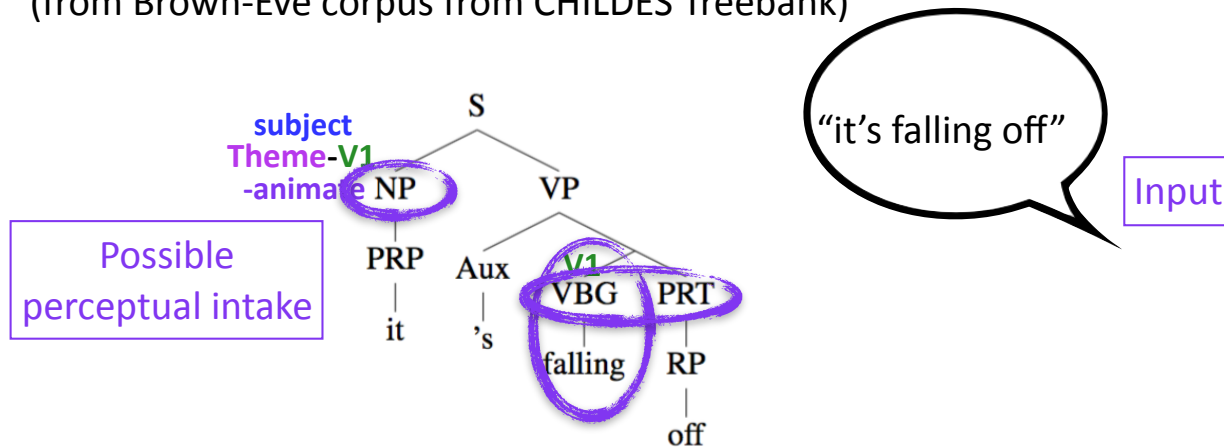


Lidz & Gagliardi 2015

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)



Acquisitional intake

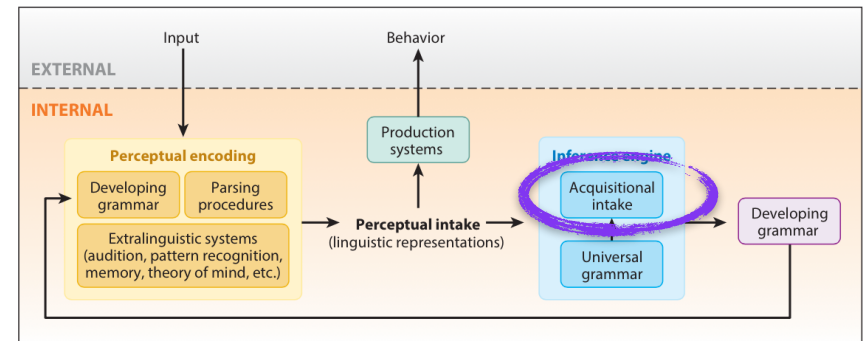
(6) rUTAH, +exp-mapping, -some available tense and aspect information

FALL

-animate subject: 1

+movement: 0

NP V PRT



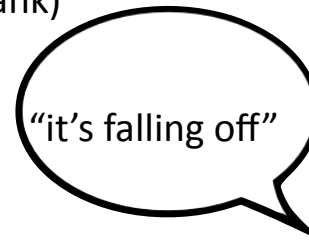
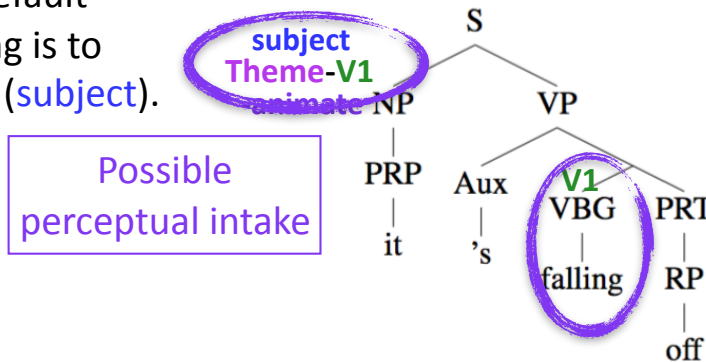
Lidz & Gagliardi 2015

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)

Theme is only role so is default highest. Expected mapping is to highest syntactic position (subject).



Input

Acquisitional intake

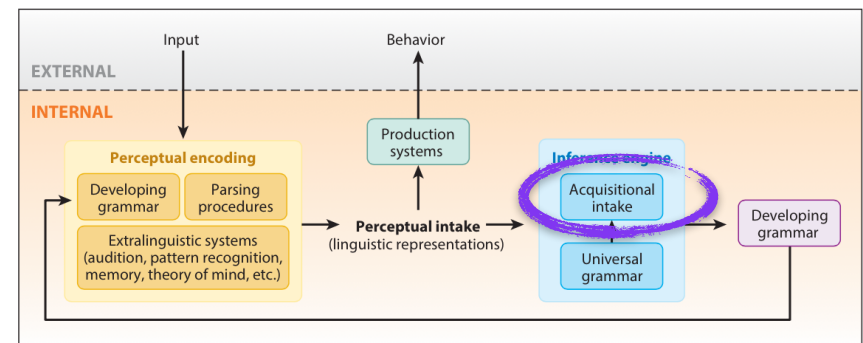
(7) rUTAH, -exp-mapping, -some available tense and aspect information

FALL

-animate subject: 1

Highest role as subject: 1

NP V PRT

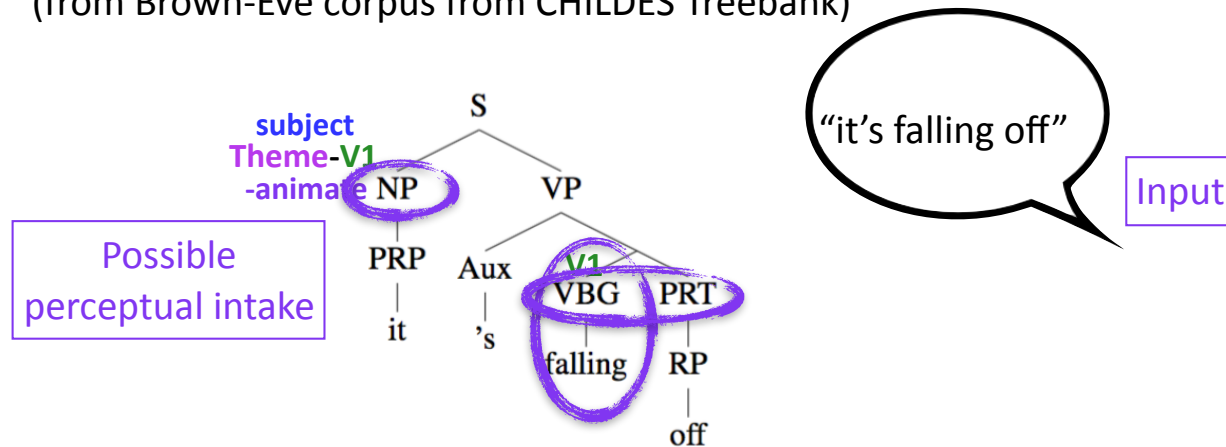


Lidz & Gagliardi 2015

Acquisitional intake options



(from Brown-Eve corpus from CHILDES Treebank)



Acquisitional intake

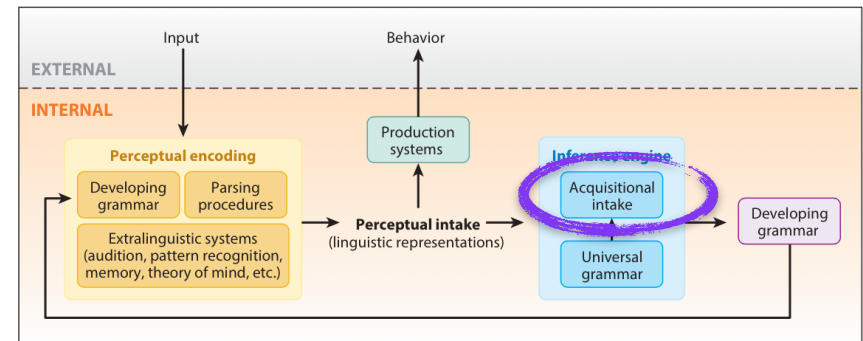
(8) rUTAH, -exp-mapping, +some available tense and aspect information

FALL

-animate subject: 1

Highest role as subject: 1

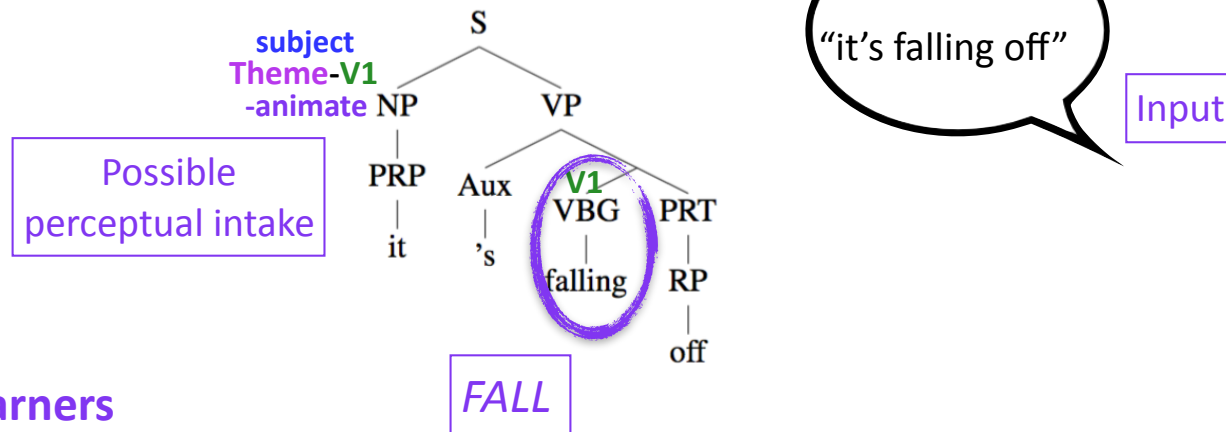
NP V_{present_participle} PRT



Lidz & Gagliardi 2015

Acquisitional intake options

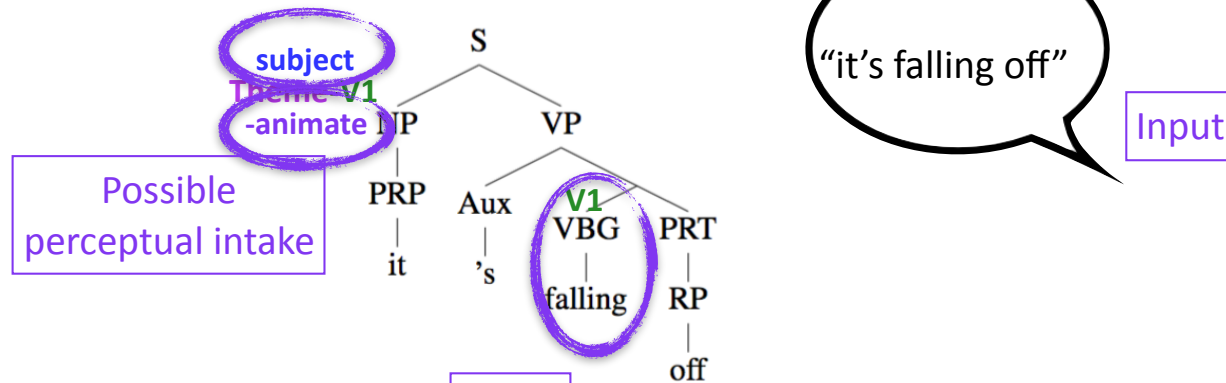
(from Brown-Eve corpus from CHILDES Treebank)



Comparison: 8 learners

Acquisitional intake options

(from Brown-Eve corpus from CHILDES Treebank)



Comparison: 8 learners

FALL

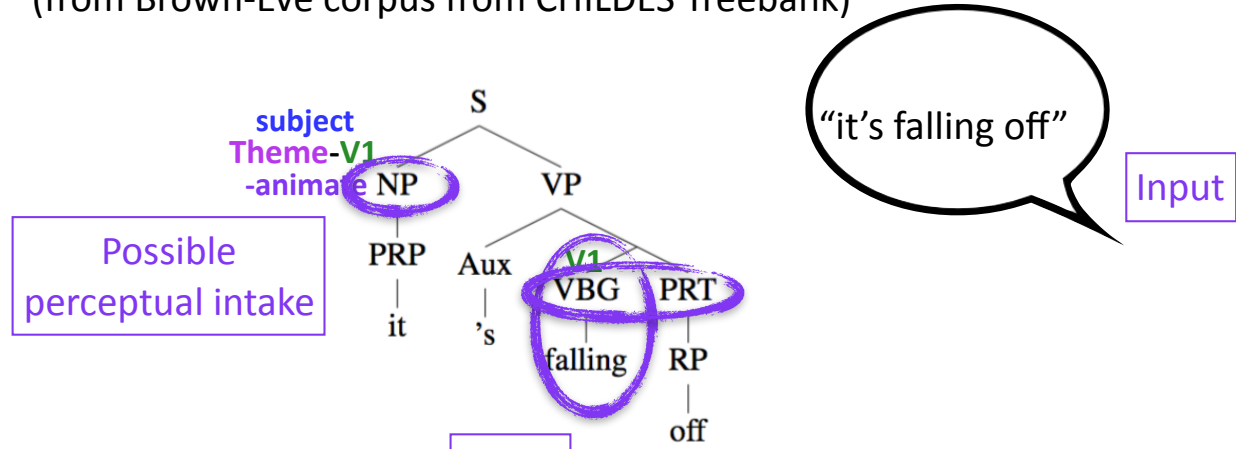
animacy

-animate subject: 1

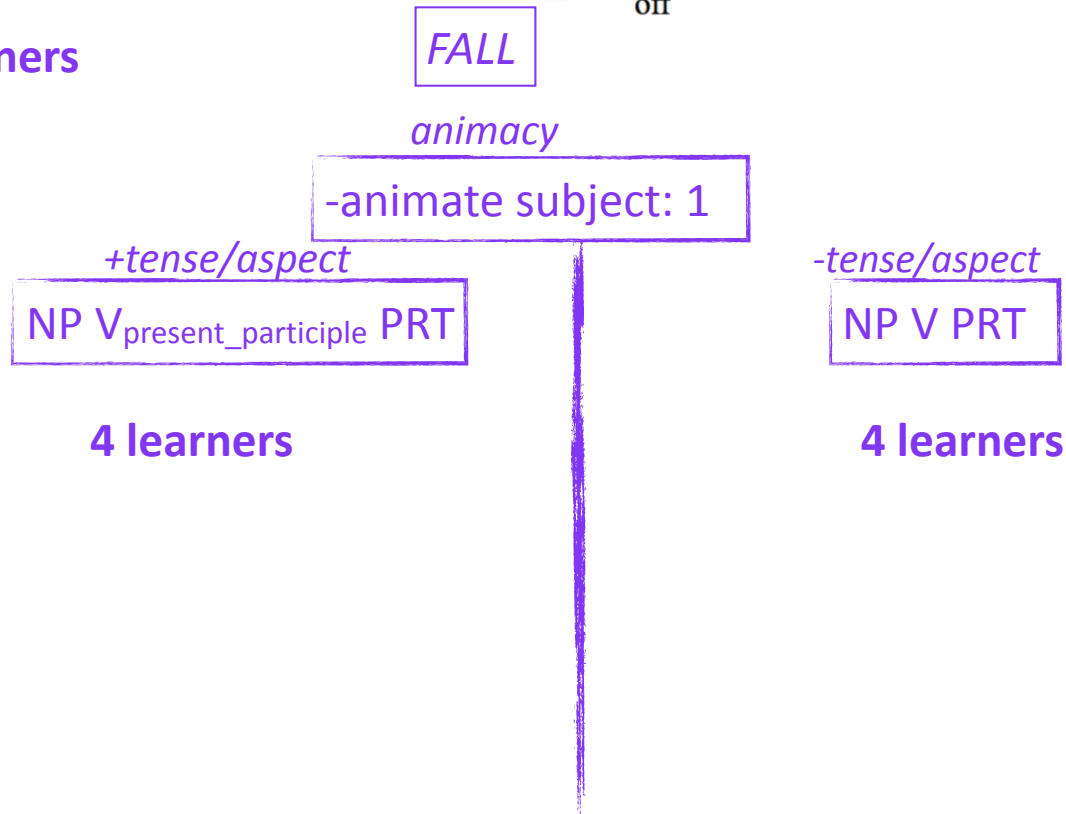
All 8 learners

Acquisitional intake options

(from Brown-Eve corpus from CHILDES Treebank)

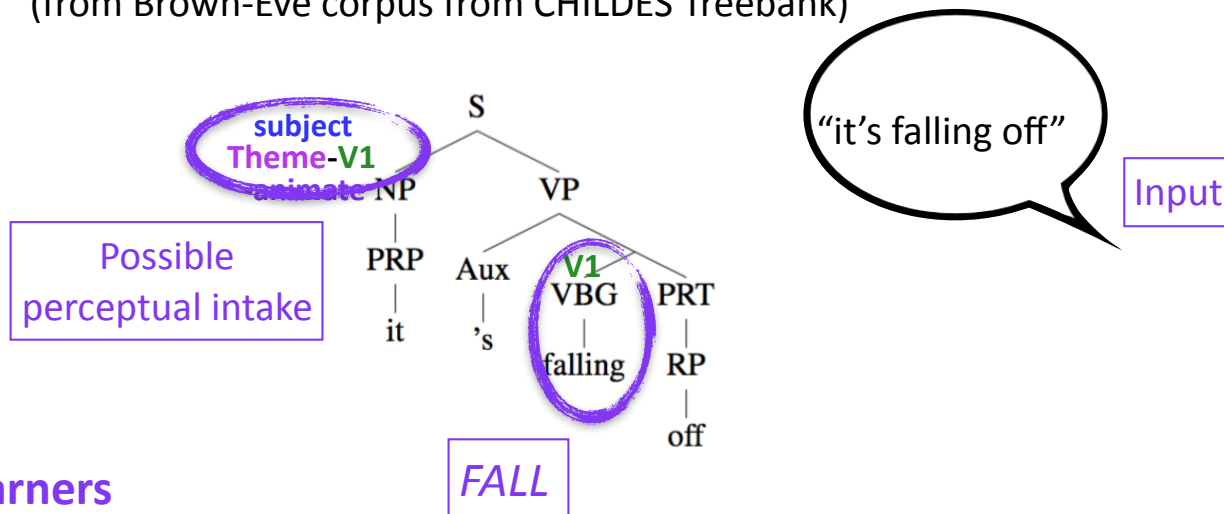


Comparison: 8 learners

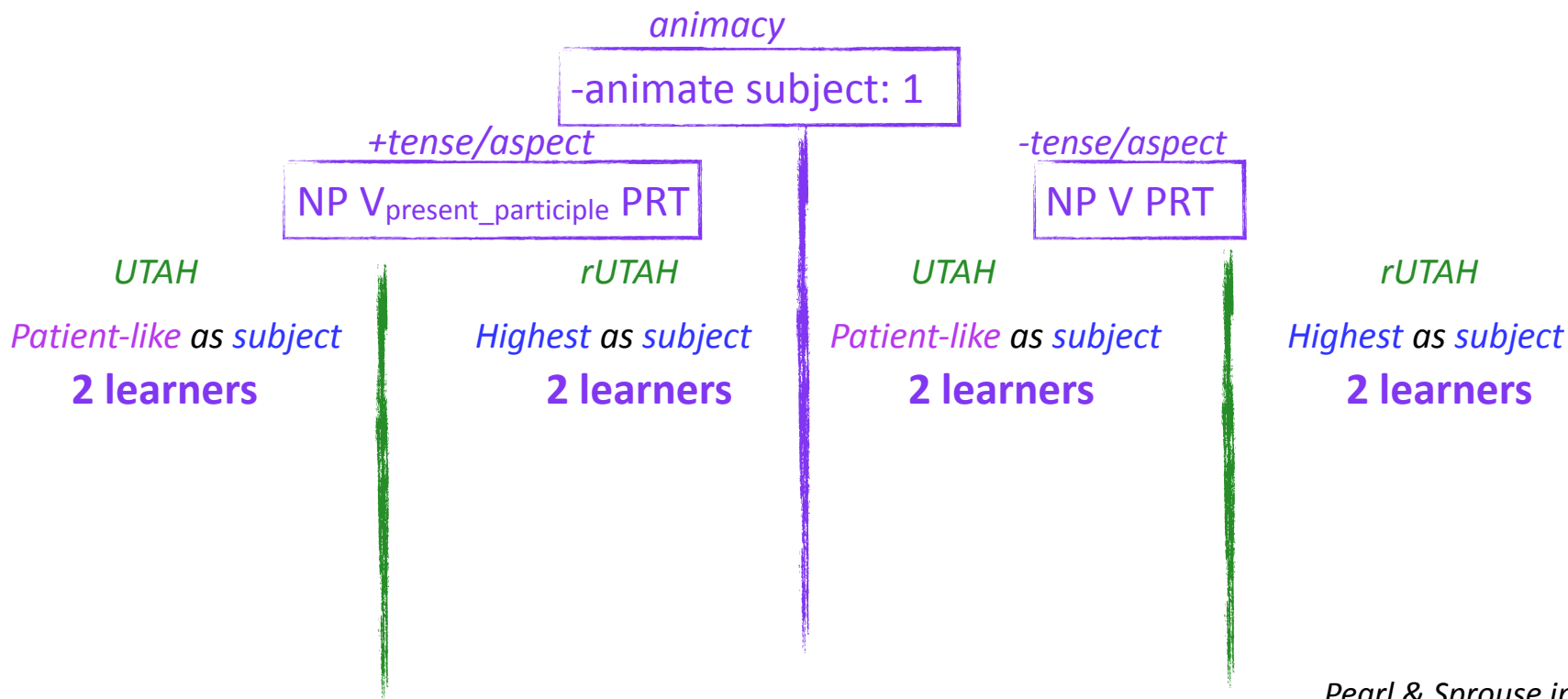


Acquisitional intake options

(from Brown-Eve corpus from CHILDES Treebank)



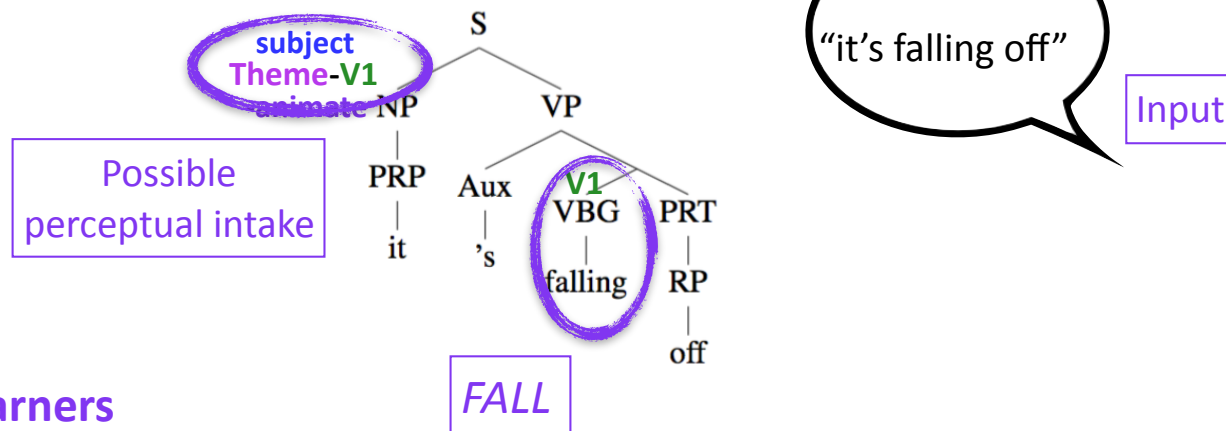
Comparison: 8 learners



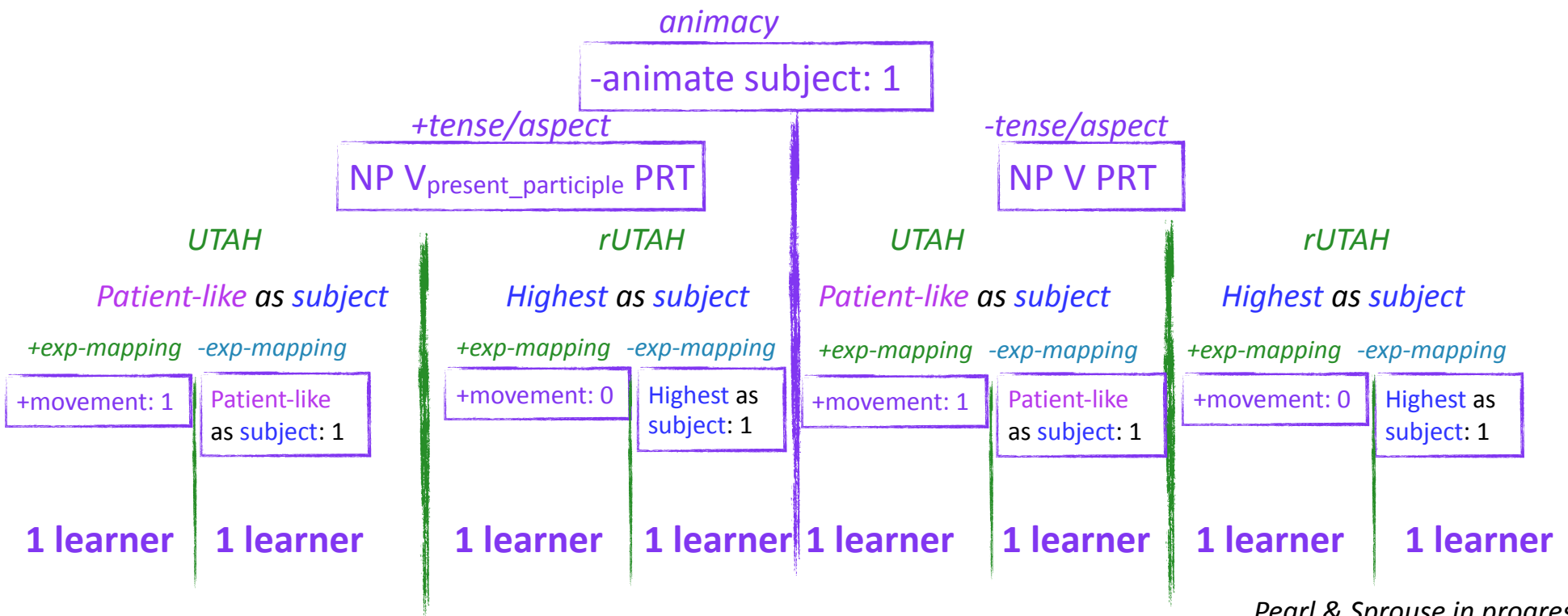
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Comparison: 8 learners



Acquisitional intake: Input data

Brown-Eve+Valian

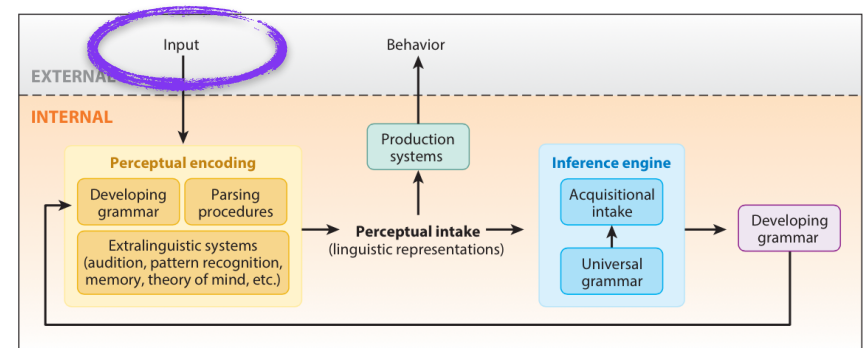


Data come from the Brown-Eve corpus (Brown 1973) and the Valian corpus (Valian 1991), with syntactic & thematic annotations provided by the CHILDES Treebank (Pearl & Sprouse 2013).



This corpus (Brown-Eve+Valian) contains speech directed at 22 children between the ages of 18 and 32 months.

There are ~40,000 utterances total, comprised of ~193,000 word tokens. Of the 553 verb lexical items that appear, 239 occur 5 or more times.



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Acquisitional intake: Input data

Brown-Eve+Valian



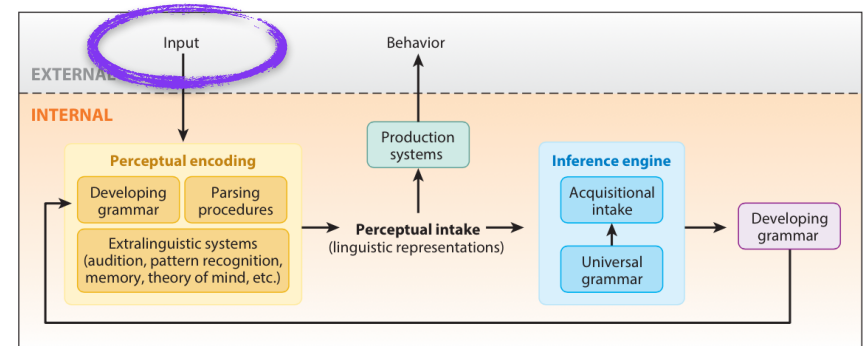
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There are ~40,000 utterances total, comprised of ~193,000 word tokens. Of the 553 verb lexical items that appear, 239 occur 5 or more times.

Focus on learning the predicate categories for these for now.
Intuition: Frequent enough to be useful to distributionally learn from.

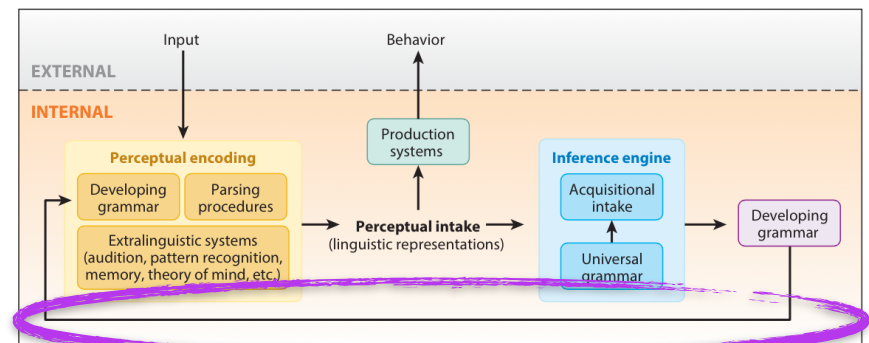


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Learning period



Basic question: Is it possible for the child to use the **acquisitional intake** to achieve the **target knowledge/behavior** in the **amount of time** children typically get to do it, given the incremental nature of learning and children's cognitive constraints?



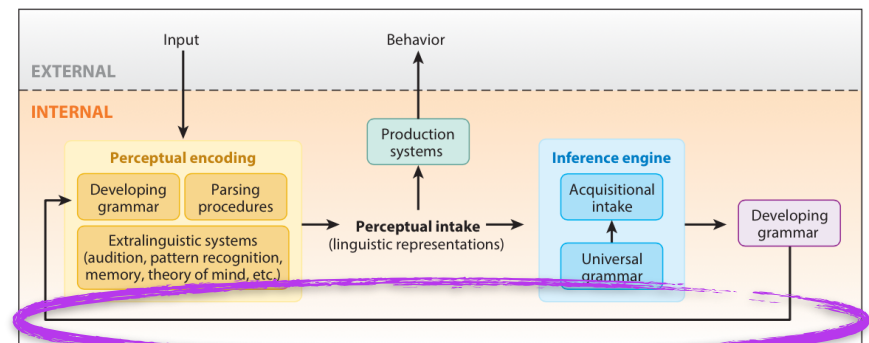
Lidz & Gagliardi 2015

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However, before we try to answer this, there's an **even more basic question** that's often worth asking.



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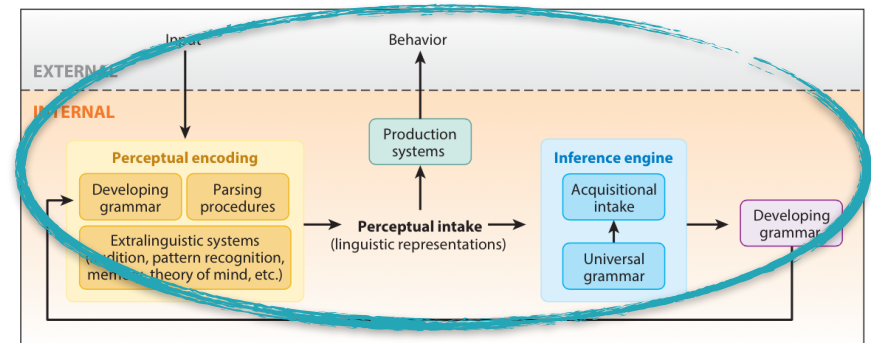
Learning period



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Pearl & Sprouse in progress

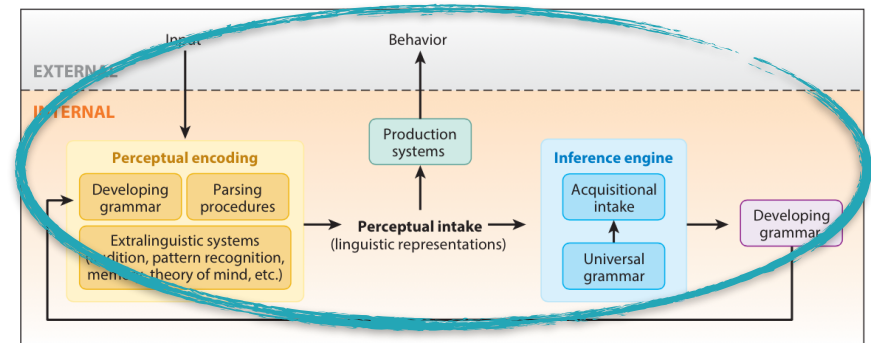
Learning period



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This is the goal of learnability approaches (often posed at the computational-level of analysis [Marr 1982]): Frank et al. 2009, Goldwater et al. 2009, Pearl et al. 2010, Pearl 2011, Legate & Yang 2012, Dillon et al. 2013, Doyle & Levy 2013, Feldman et al. 2013, Orita et al. 2013



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Pearl & Sprouse in progress

Learning period

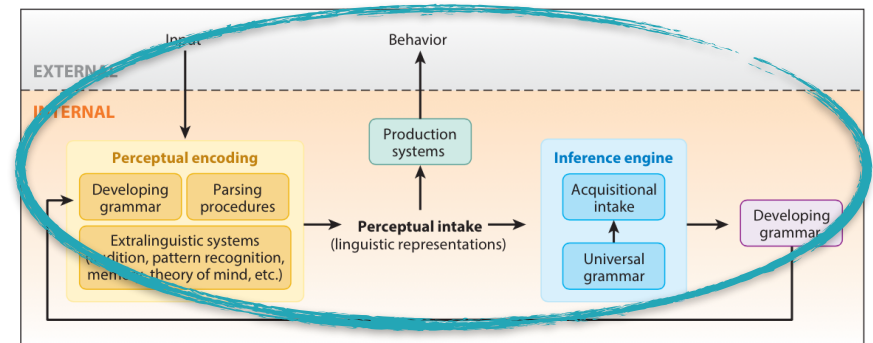


Even more

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This kind of analysis is very helpful for determining if this implementation of the acquisition task is the right one. In particular, **if children are sensitive to this information in the perceptual intake**, is that enough to yield the **target knowledge/behavior**? Are these **useful** learning assumptions for children to have to create the **acquisitional intake**? Are these **useful** representations?



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Pearl & Sprouse in progress

Learning period



Even more

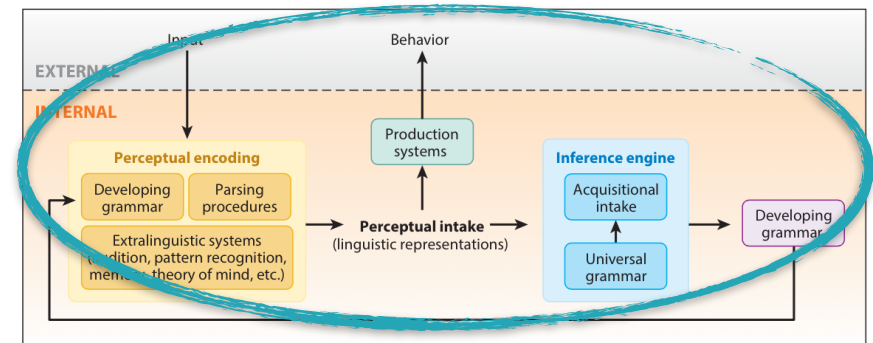
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This is typically implemented as an **ideal learner model**, which isn't concerned with the cognitive limitations and incremental learning restrictions children have.



(That is, **useful** for children is different from **useable** by children in real life.)



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Learning period

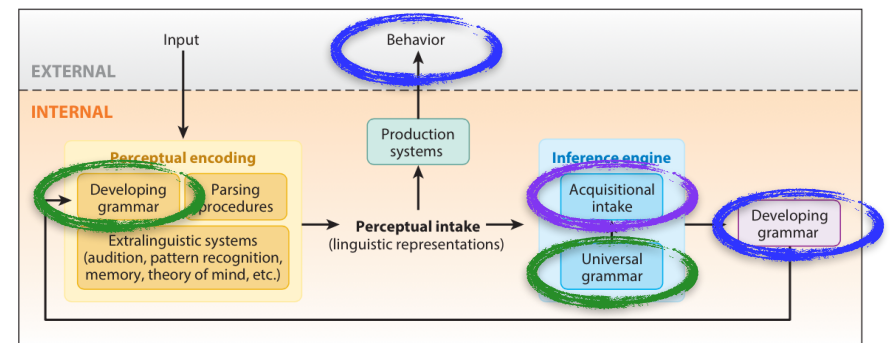


Even more

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So, for an **ideal learner**, learning period considerations aren't as important as considerations about the **initial state**, **data intake**, and **target knowledge/behavior**.



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Learning period

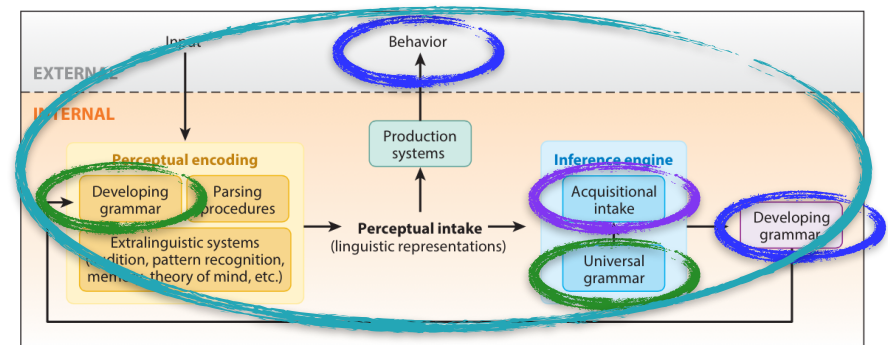


Even more

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Practical note: Doing a computational analysis is often a really good idea to make sure we've got the **right conceptualization of the acquisition task** (see Pearl 2011 for the trouble you can get into when you don't do this first).



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Learning period

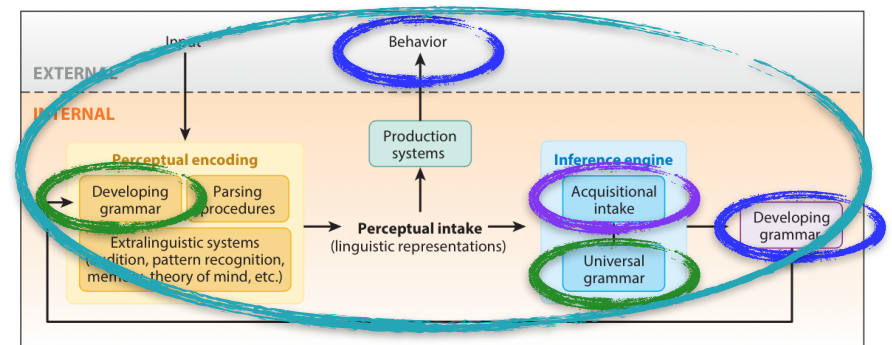


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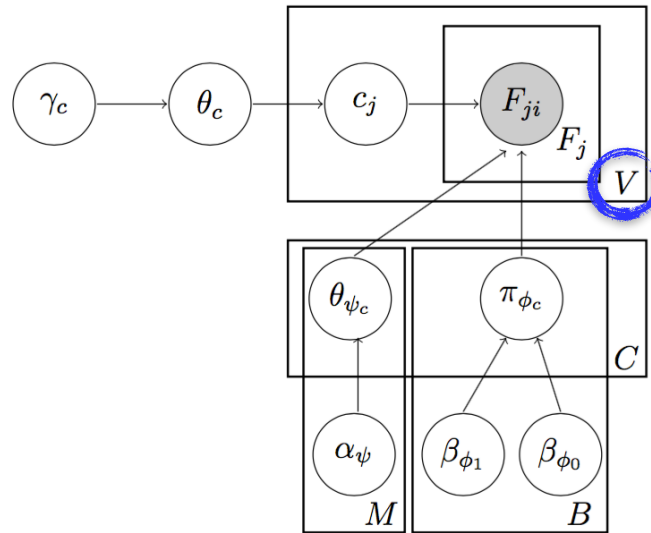
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So, that's why we're going to start with a **computational-level** model of the acquisition process.



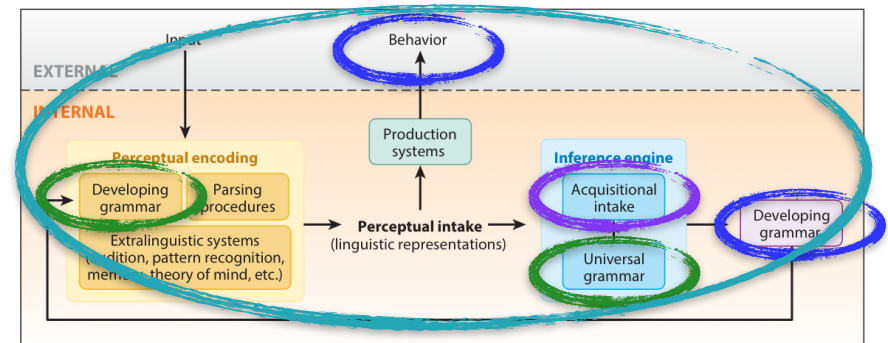
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Learning process: Computational-level



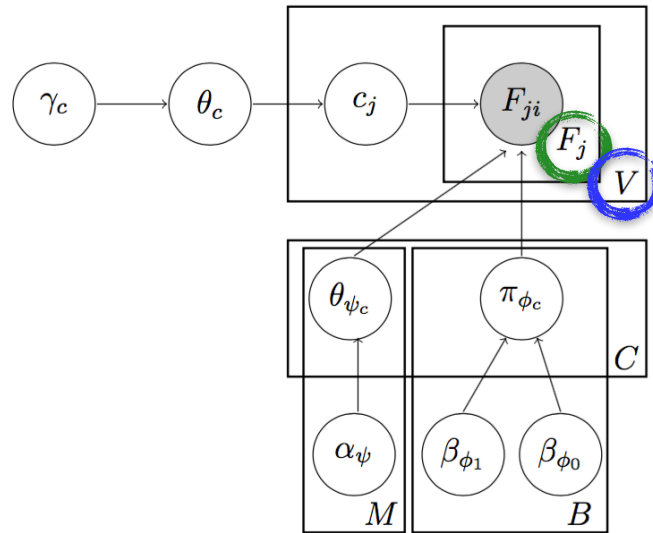
Generative model of how the observable data for each verb are created.

FALL



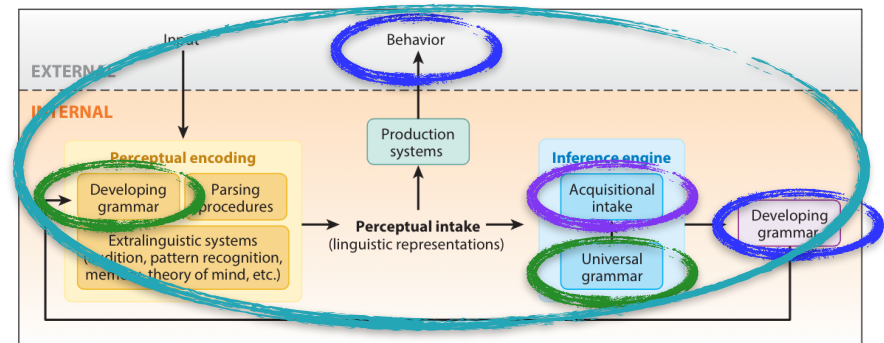
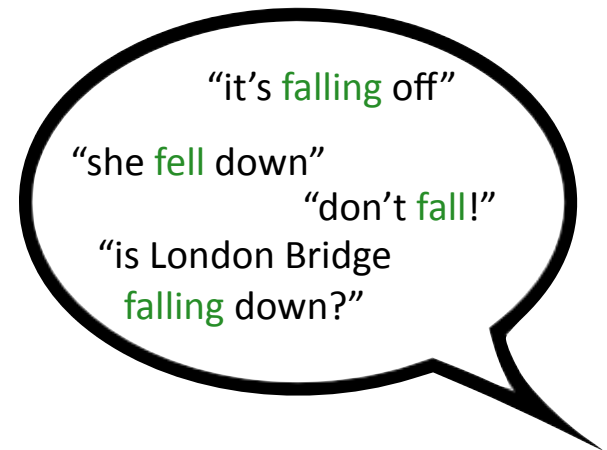
Lidz & Gagliardi 2015

Learning process: Computational-level



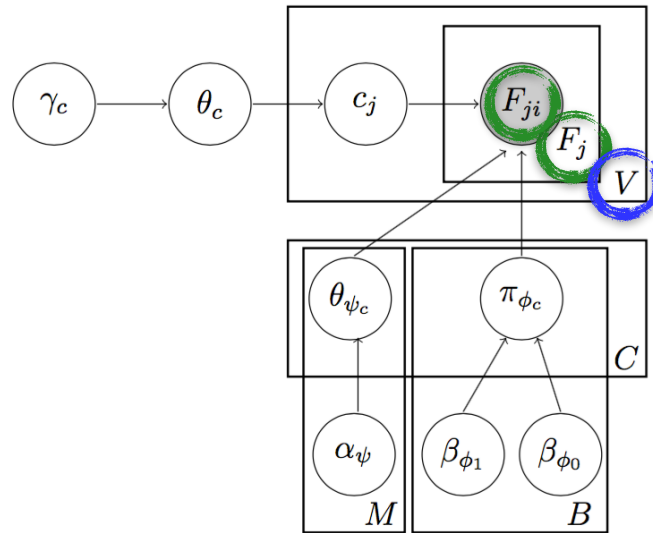
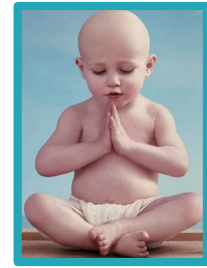
Each verb is observed in a certain number of instances in the **input**.

FALL



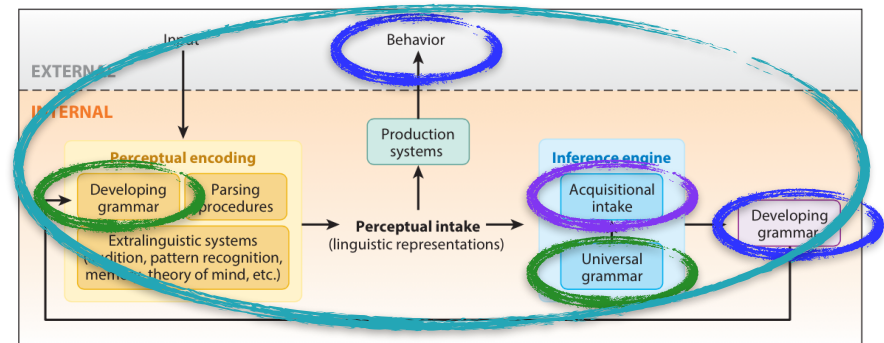
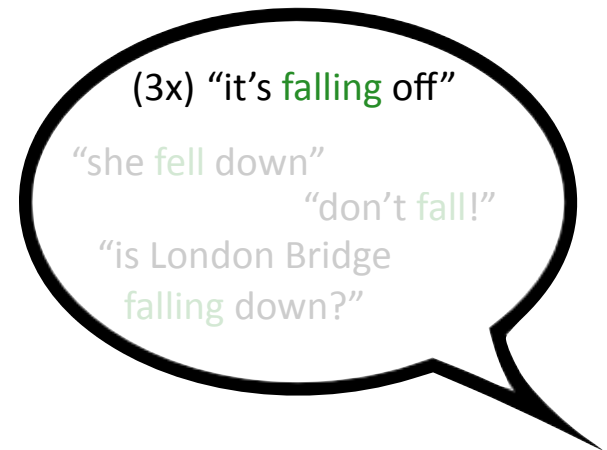
Lidz & Gagliardi 2015

Learning process: Computational-level



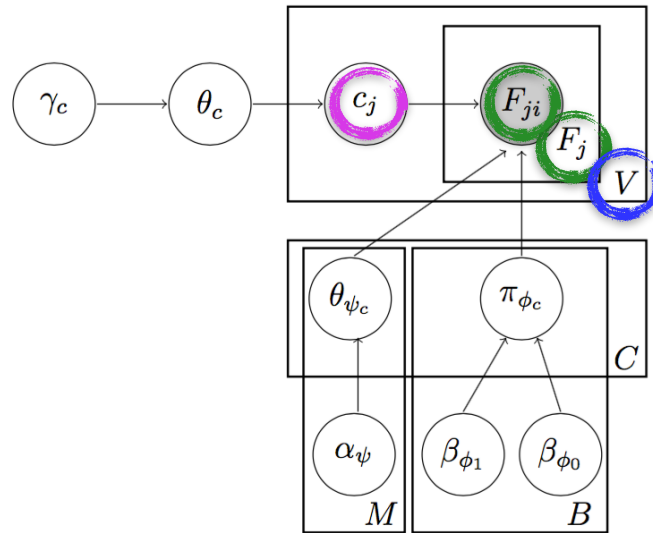
Each instance is observed some number of times.

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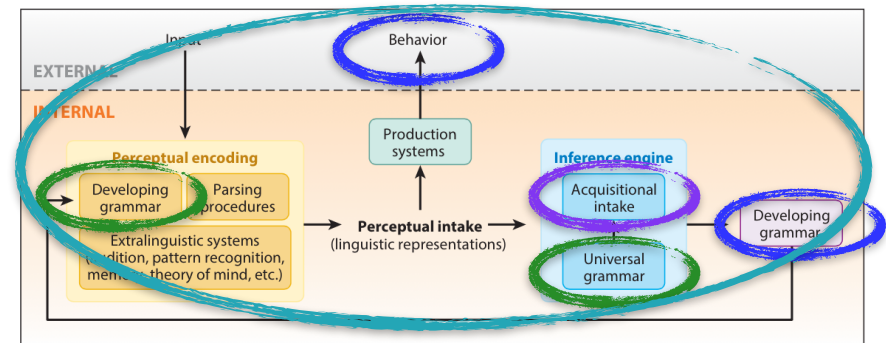
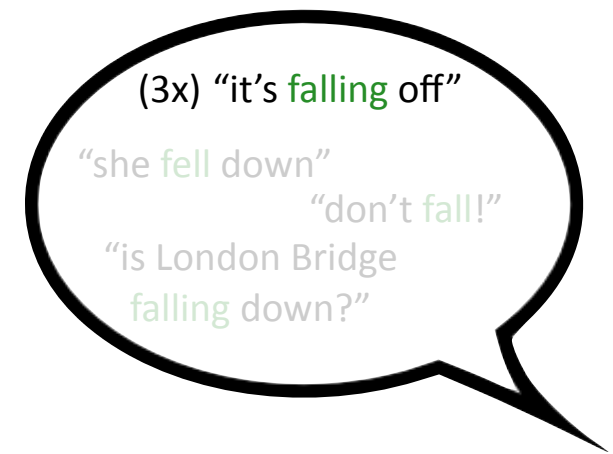
Learning process: Computational-level



Each **verb** belongs to some **class** which determines its linguistic behavior.

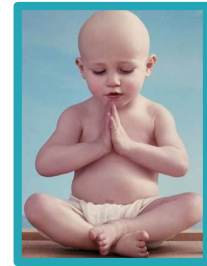
FALL

unaccusatives

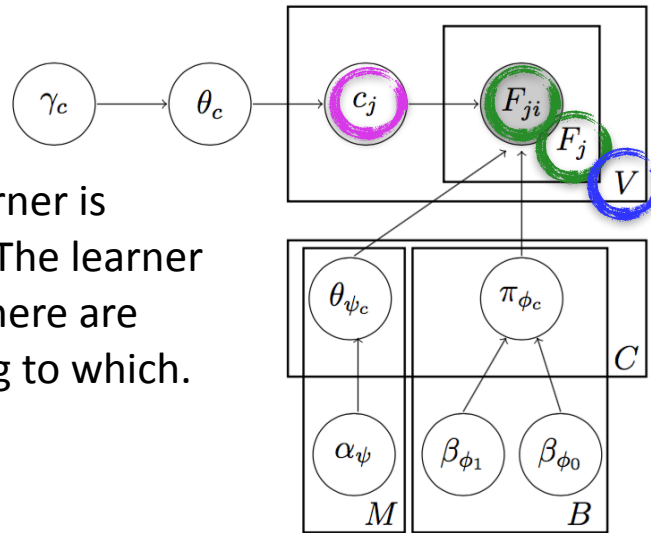


Lidz & Gagliardi 2015

Learning process: Computational-level



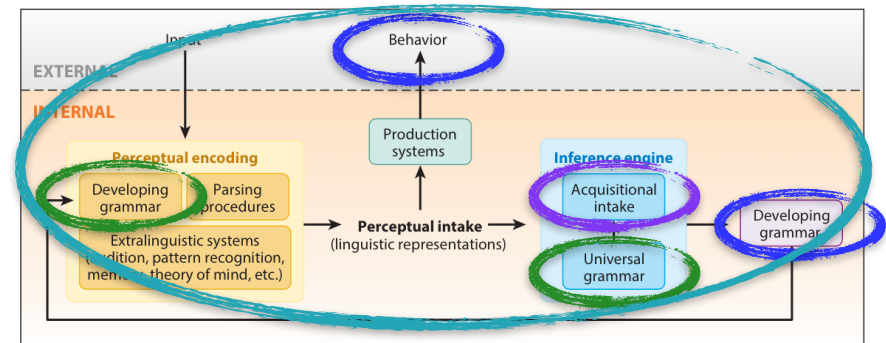
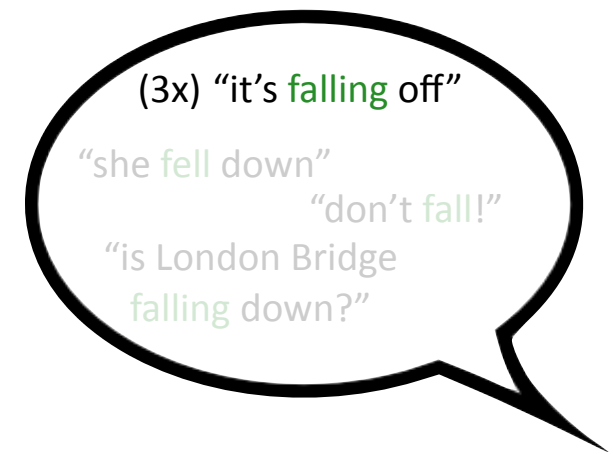
The **class** is the main thing the learner is trying to figure out for each verb. The learner doesn't know how many classes there are beforehand, or which verbs belong to which.



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives

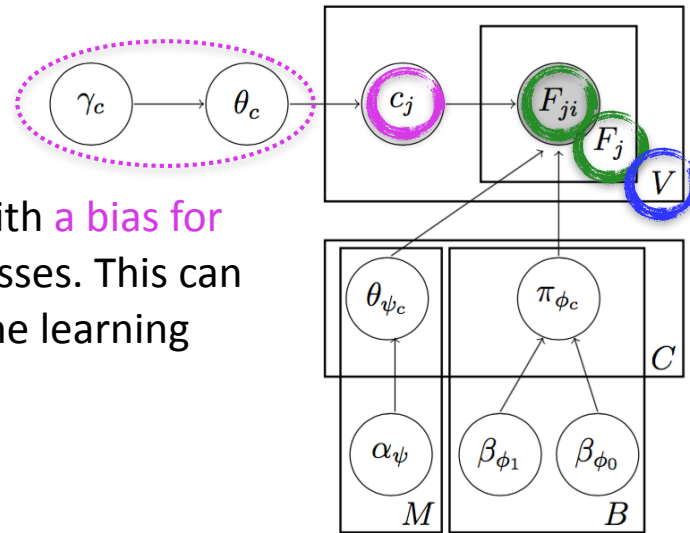


Lidz & Gagliardi 2015

Learning process: Computational-level



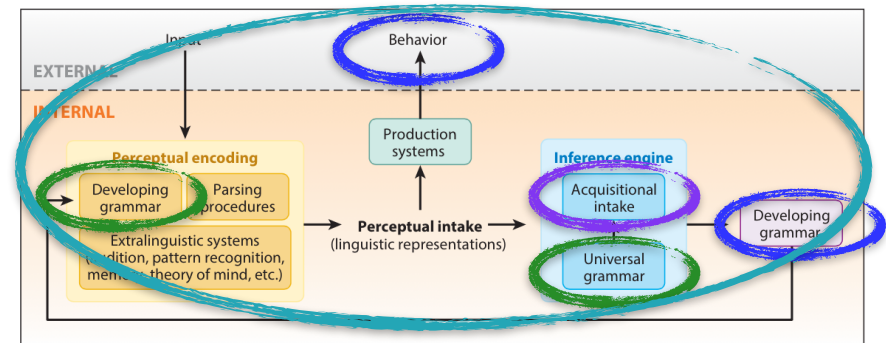
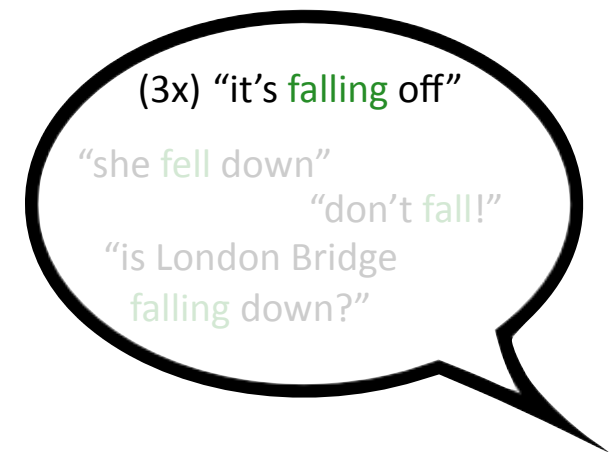
However, the learner does begin with a bias for fewer classes, rather than more classes. This can be adjusted automatically during the learning process.



Each verb belongs to some class which determines its linguistic behavior.

FALL

unaccusatives

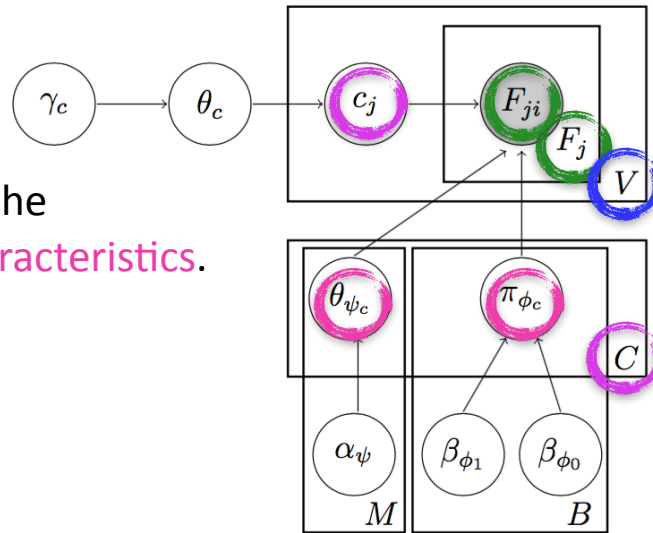


Lidz & Gagliardi 2015

Learning process: Computational-level



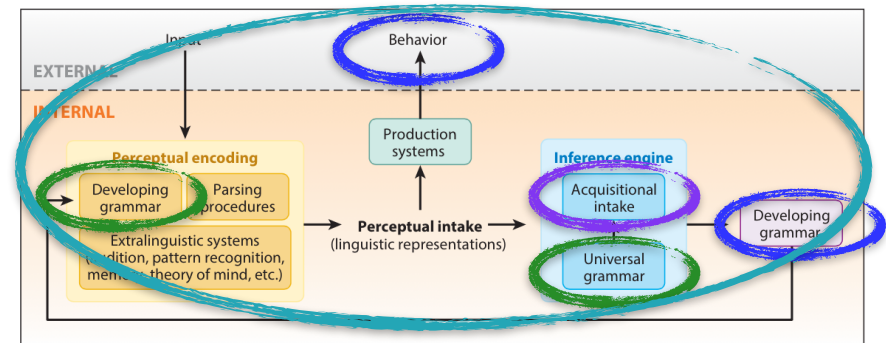
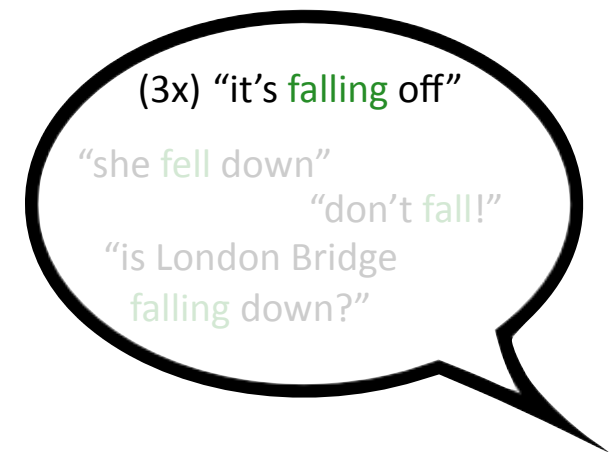
Depending on the **class** of the verb, the **observed usage** will have **certain characteristics**.



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives

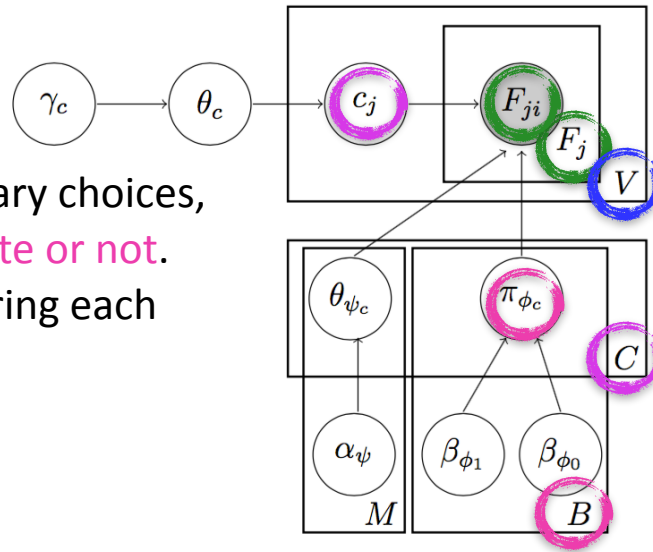


Lidz & Gagliardi 2015

Learning process: Computational-level



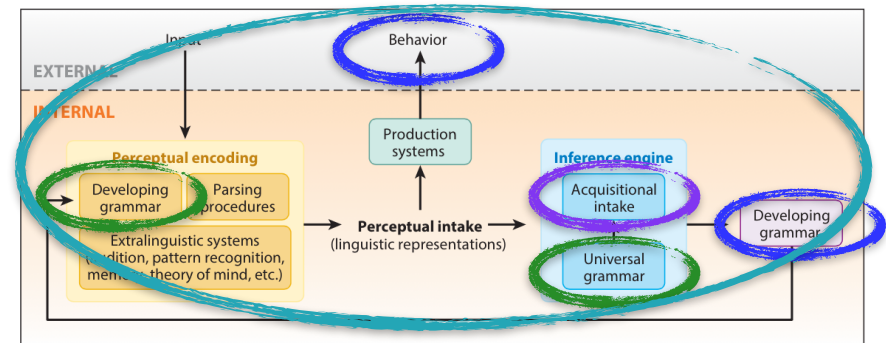
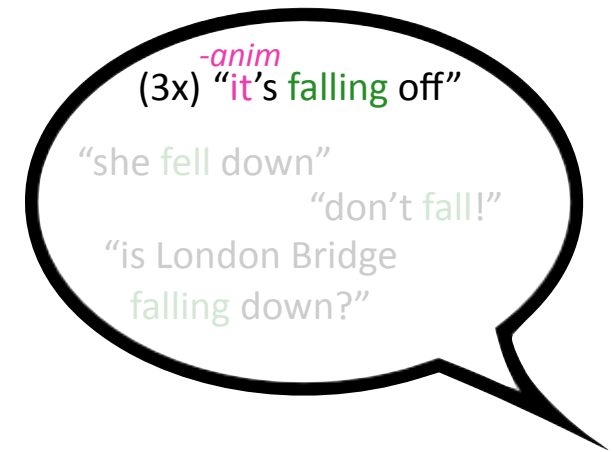
These characteristics can include binary choices, such as whether the subject is **animate or not**. Each class has a **probability** of preferring each option.



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives



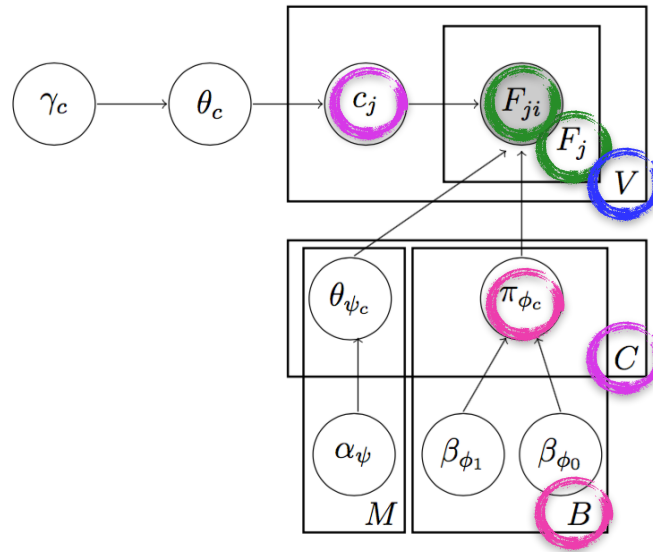
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Learning process: Computational-level



Binary properties include:

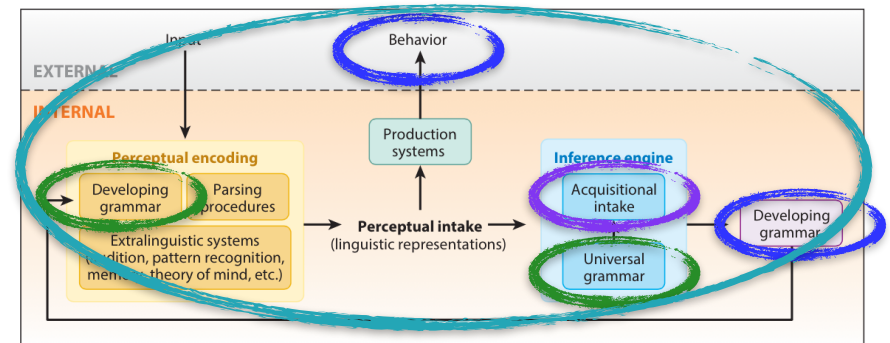
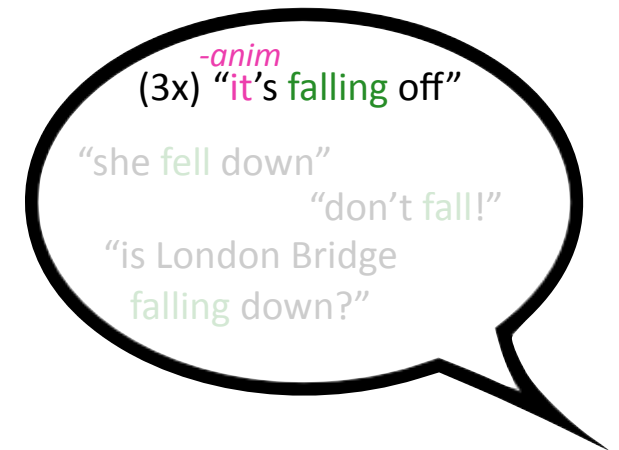
- +/-animate subject
- +/-animate object
- +/-animate indirect object
- +/-movement (when +exp-mapping)



Each verb belongs to some class which determines its linguistic behavior.

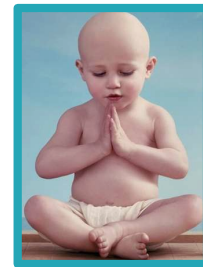
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unaccusatives

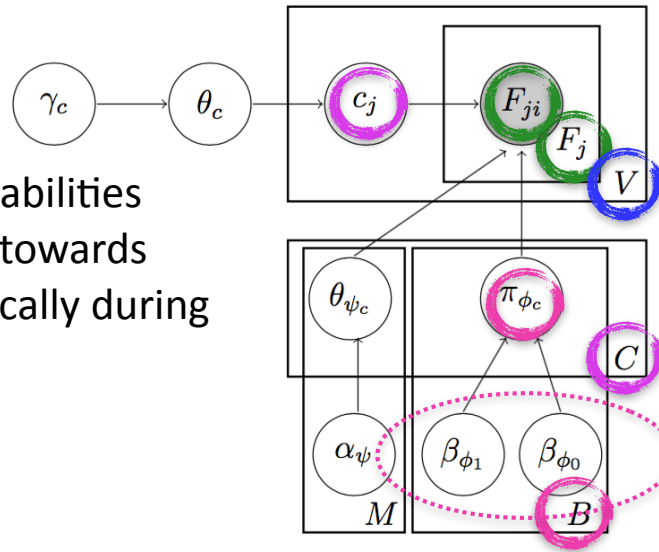


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Learning process: Computational-level



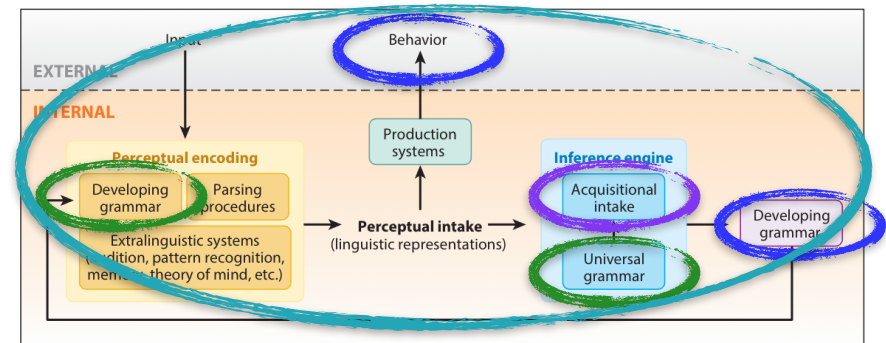
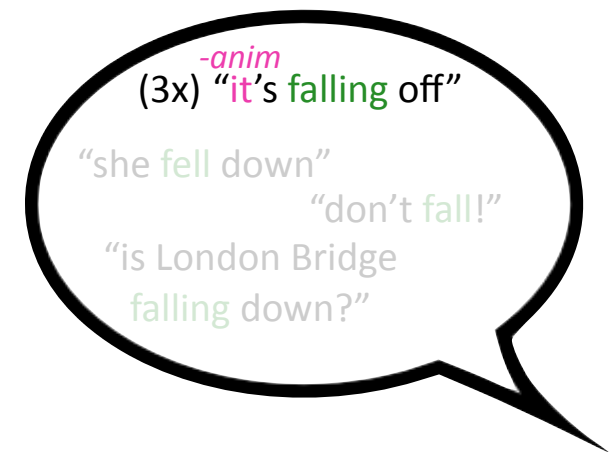
The learner doesn't know these probabilities beforehand, and begins with no bias towards either. This can be adjusted automatically during the learning process.



Each verb belongs to some class which determines its linguistic behavior.

FALL

unaccusatives

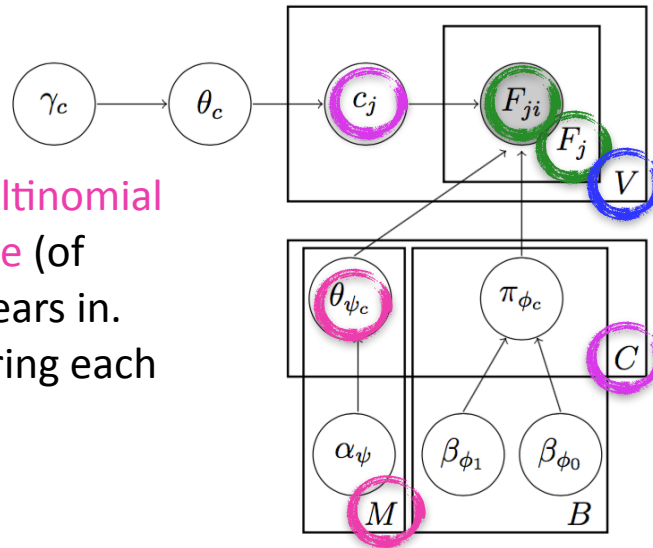


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Learning process: Computational-level



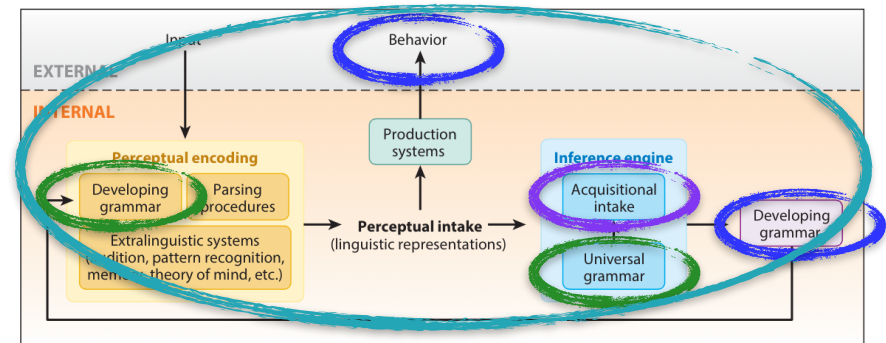
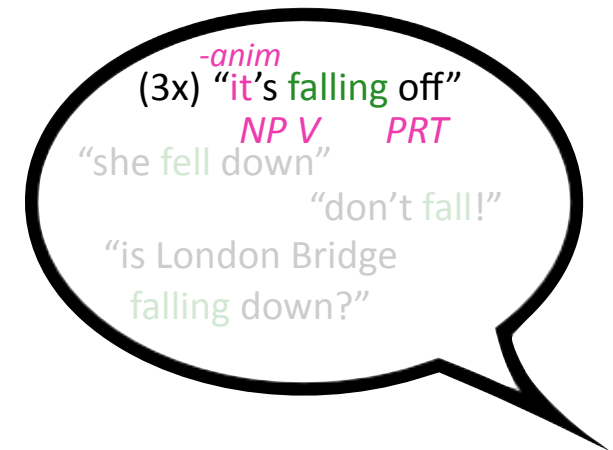
These characteristics also include **multinomial** choices, such as **which syntactic frame** (of however many there are) a verb appears in. Each class has a **probability** of preferring each option.



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives



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Learning process: Computational-level



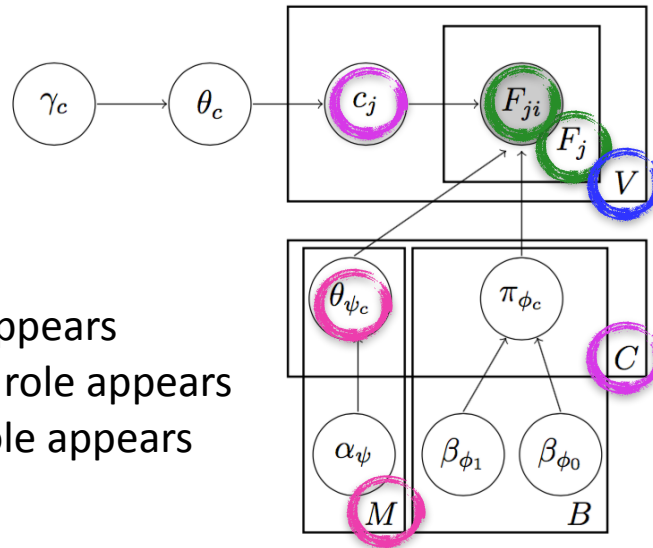
Multinomial properties include:

which syntactic frame is used
(if -exp-mapping)

where the Agent-like/Highest role appears

where the Patient-like/next-Highest role appears

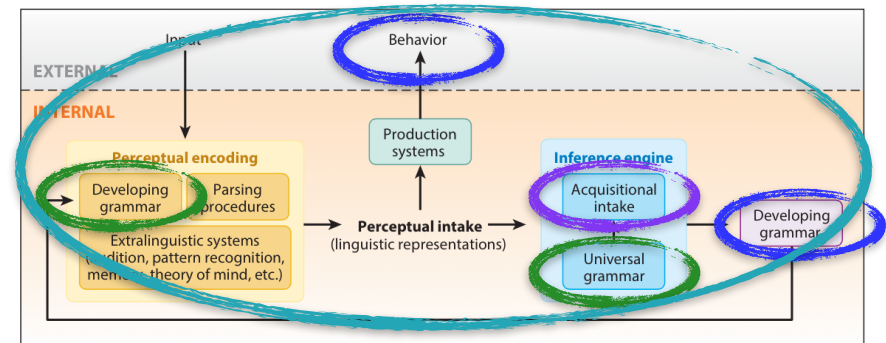
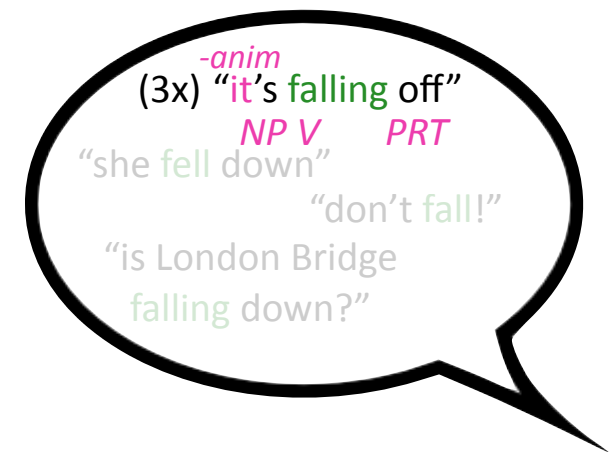
where the Goal-like/third-highest role appears



Each verb belongs to some class which determines its linguistic behavior.

FALL

unaccusatives

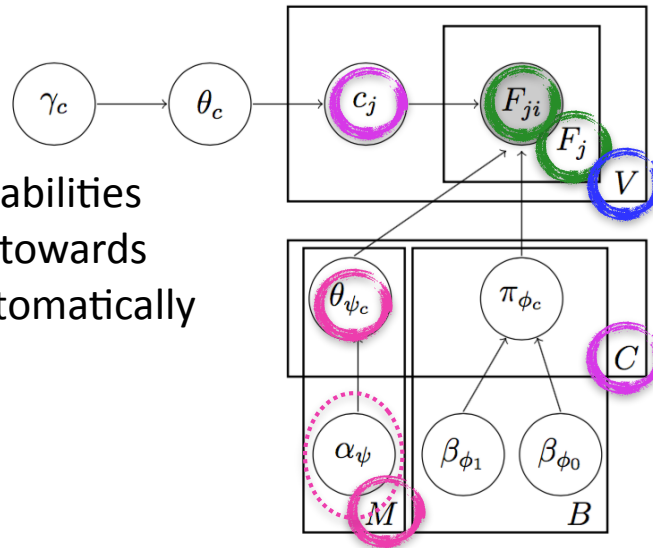


Lidz & Gagliardi 2015

Learning process: Computational-level



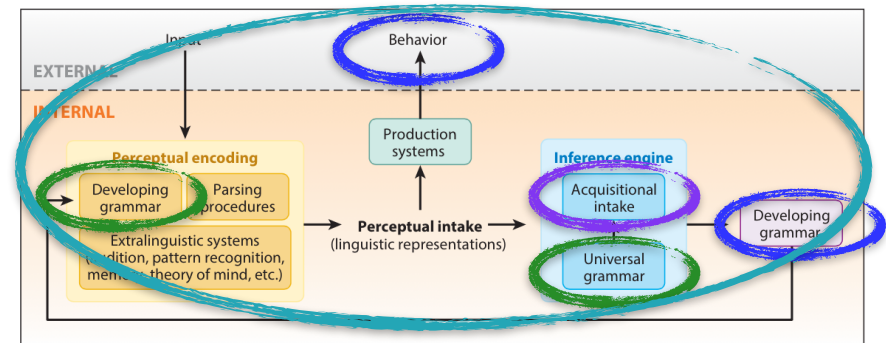
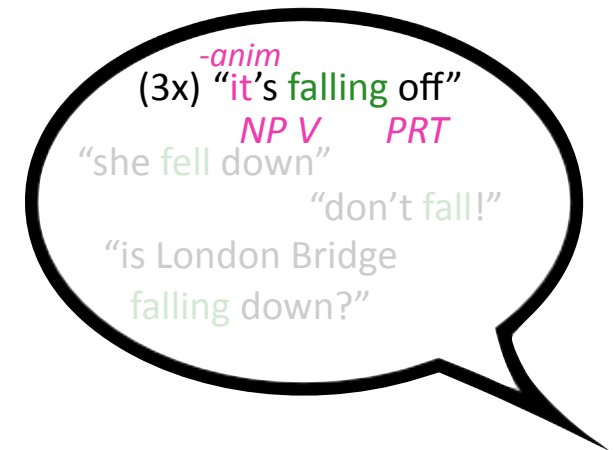
The learner doesn't know these probabilities beforehand, and begins with no bias towards any of them. This can be adjusted automatically during the learning process.



Each verb belongs to some class which determines its linguistic behavior.

FALL

unaccusatives



Lidz & Gagliardi 2015

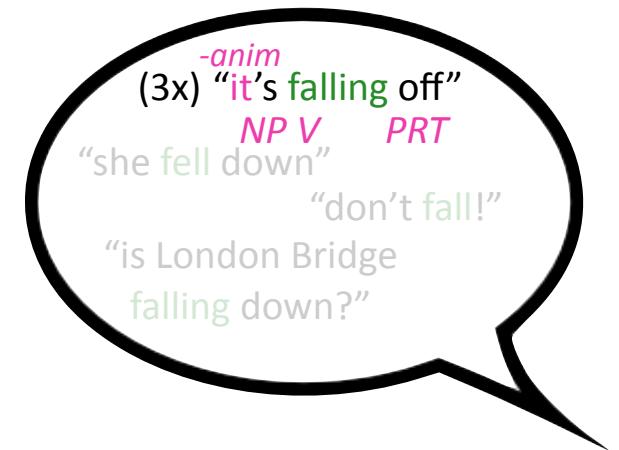
Learning process: Computational-level



Each **verb** belongs to some **class** which determines its linguistic behavior.

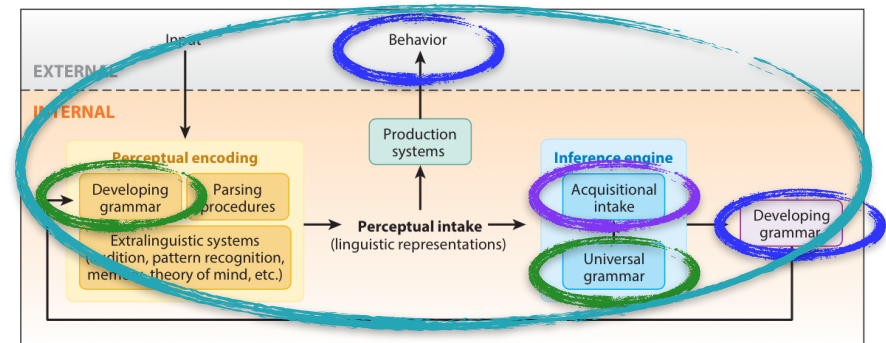
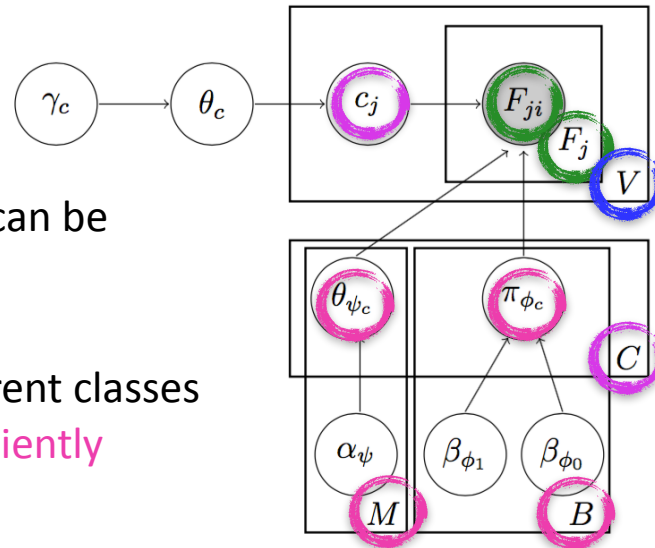
FALL

unaccusatives



All the characteristics for each class can be inferred during the learning process.

Expectation: The learner forms different classes because the characteristics are sufficiently different for each class.

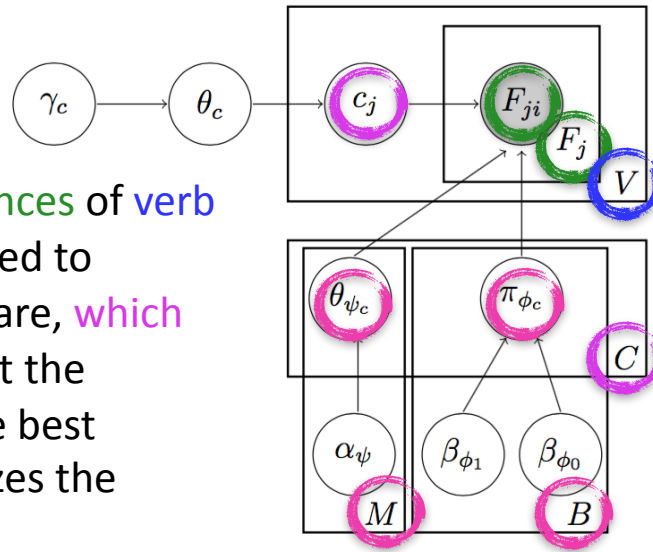


Lidz & Gagliardi 2015

Learning process: Computational-level

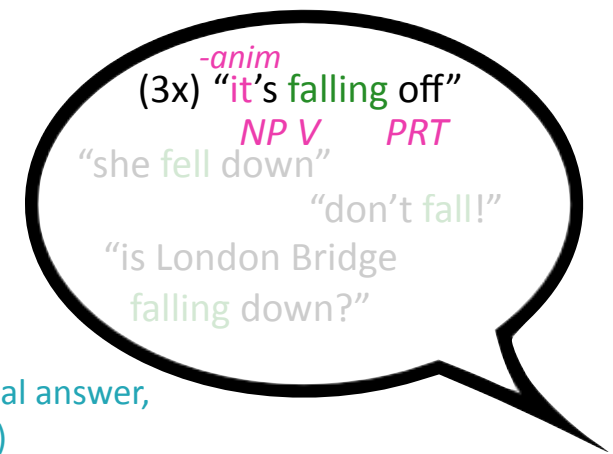


Summary: Using the **observed instances** of verb usage, **Bayesian inference** can be used to determine **how many classes** there are, **which class** each verb belongs to, and what the **characteristics are of each class**. The best answer will be the one that maximizes the probability of the **observed data**.



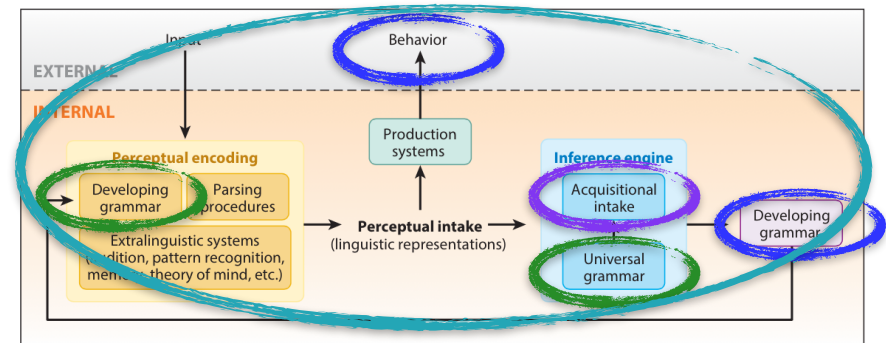
Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL *unaccusatives*



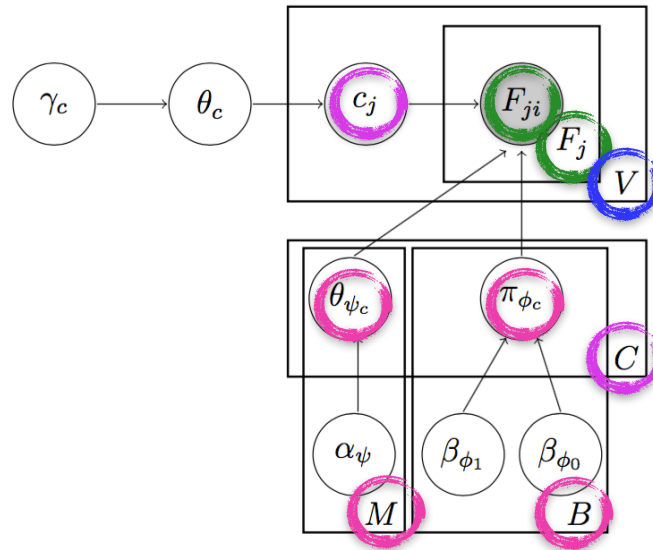
$$p_{c_j} = P(c_j | c_{-j}, \gamma_c, F_{-j}, \lambda) = p_{cat_j} * p_{binary_{c_j}} * p_{multinomial_{c_j}}$$

+ Gibbs sampling (method guaranteed to find optimal answer, given sufficient time to search the hypothesis space)



Lidz & Gagliardi 2015

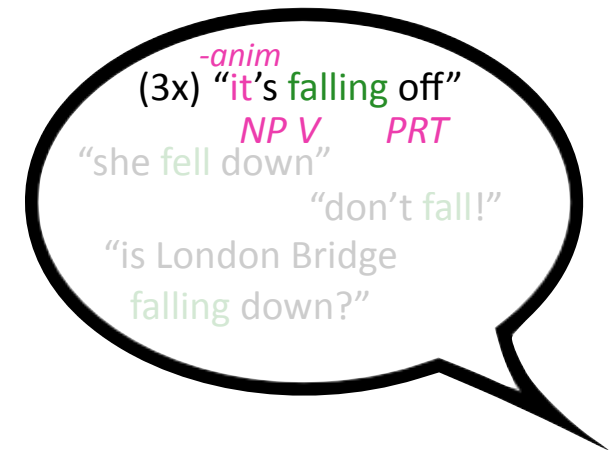
Learning process: Computational-level



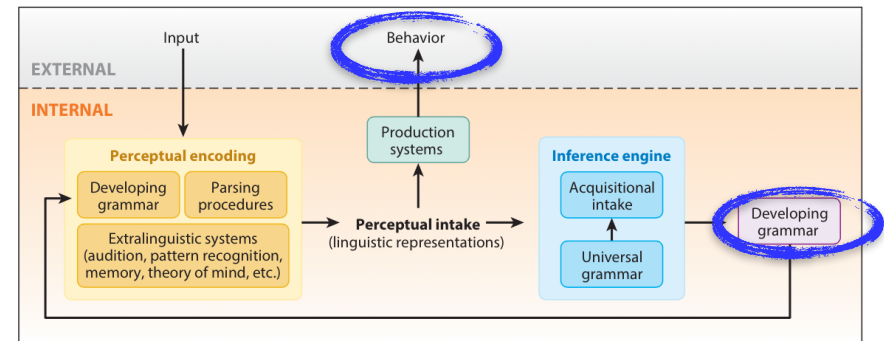
Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives



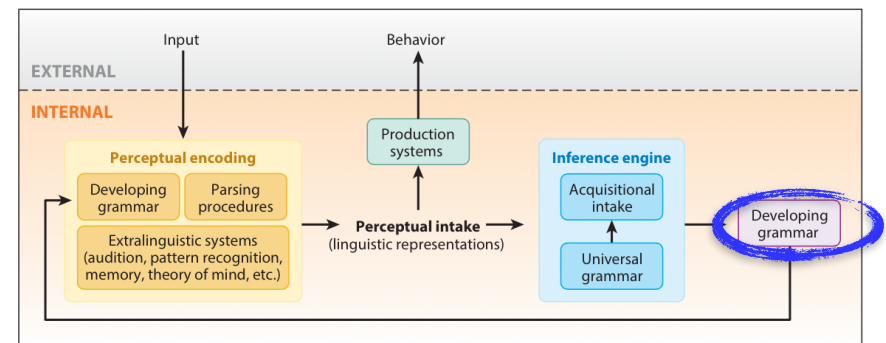
Goal: Determine if the information provided (syntactic, conceptual, and semantic-syntactic cues) is sufficient to identify useful verb classes this way.



Lidz & Gagliardi 2015

Target state: Useful verb classes

Adult knowledge is the eventual target state for acquisition, and there are a variety of verb distinctions that have different syntactic and/or thematic role implications. Do some of these distinctions fall out directly by using the syntactic, conceptual, and semantic-syntactic cues we're using?



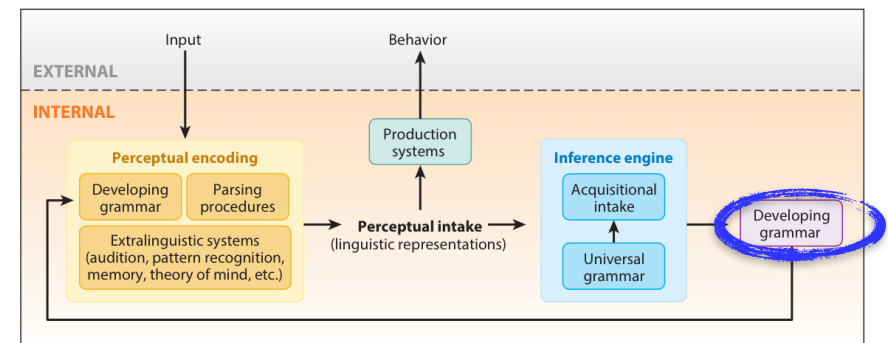
Lidz & Gagliardi 2015

Pearl & Sprouse in progress

Target state: Useful verb classes

Adult knowledge is the eventual target state for acquisition, and there are a variety of verb distinctions that have different syntactic and/or thematic role implications. Do some of these distinctions fall out directly by using the syntactic, conceptual, and semantic-syntactic cues we're using?

Given the input data we have from the Brown-Eve+Valian corpus (which is directed at children age 2;8 and younger), we should probably focus on distinctions children seem to have made by age three.



Lidz & Gagliardi 2015

Pearl & Sprouse in progress

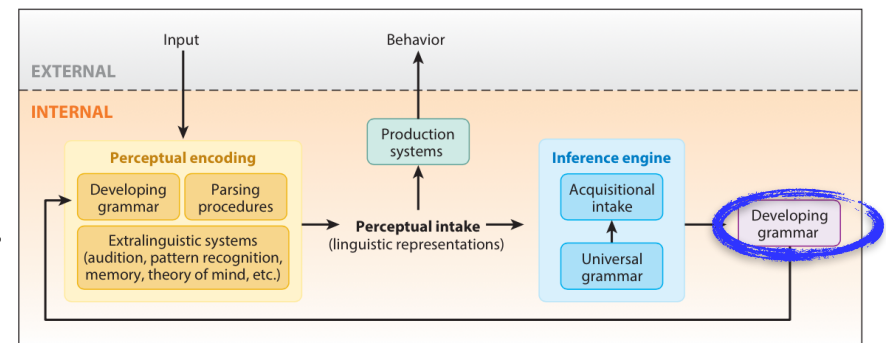
Target state: Useful verb classes

Cues to **transitives** (allowing a single object) seem to be recognized as early as two years old in English: Naigles 1990, Naigles & Kako 1993, Yuan & Fisher 2009.

Transitive, single object “Jack ___ it.”



+ = bite, eat, forget, kick, understand, ...
- = cough, laugh, sleep, sneeze, ...



Lidz & Gagliardi 2015

Target state: Useful verb classes

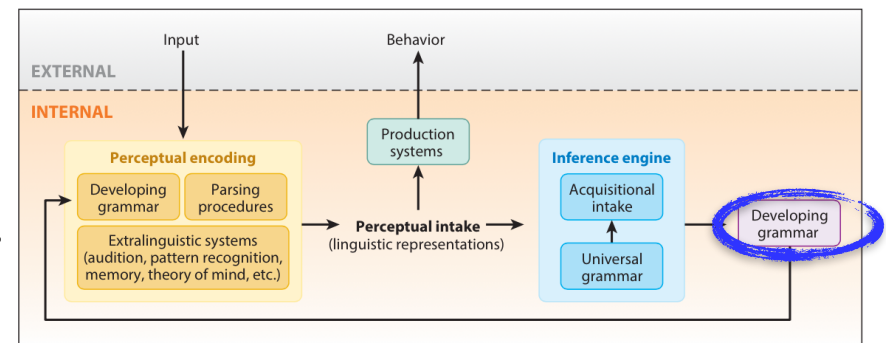
Verbs that can be used **transitively** (aren't purely intransitive) can be **passivized**, though children in English seem to only be able to recognize verbs in passives around age three: Gordon & Chafetz 1990, O'Brien et al. 2006, Crain et al., 2009, Nguyen et al. 2016.

Transitive, single object "Jack ___ it."

Passivizable "It was ___-en."
Patient-like



+ = bite, eat, forget, kick, understand, ...
- = cough, laugh, sleep, sneeze, ...



Lidz & Gagliardi 2015

Target state: Useful verb classes

Verbs allowing the **intransitive** use (no object) are recognized as early as 28 months: Scott & Fisher 2009.

Transitive, single object “Jack ___ it.”

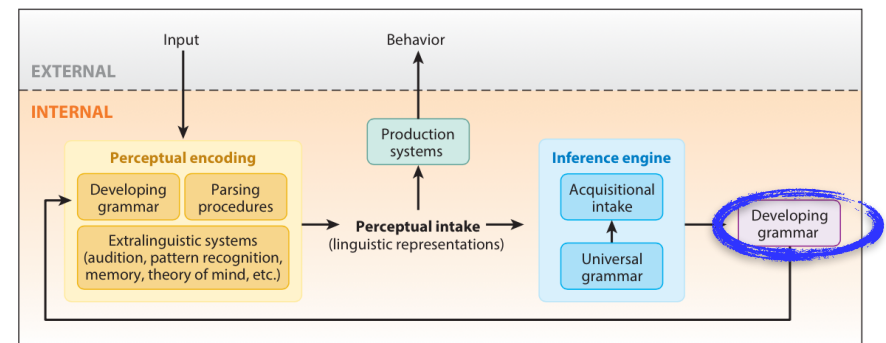
Passivizable “It was ___-en.”
Patient-like

Intransitive “Jack ___.”



+ = chirp, eat, jump, understand, ...

- = buy, give, thank, want ...



Lidz & Gagliardi 2015

Target state: Useful verb classes

Verbs allowing the **ditransitive** use (two objects: indirect and direct) are recognized by age three: Gropen, Pinker, Hollander, Goldberg, & Wilson 1989, Snyder & Stromswold 1997, Campbell & Tomasello 2001, Conwell & Demuth 2007, Thothathiri & Snedeker 2008.

Transitive, single object “Jack ___ it.”

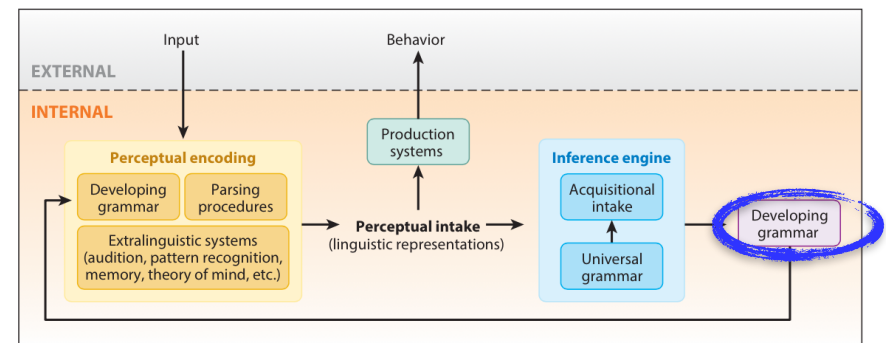
Passivizable “It was ___-en.”
Patient-like

Intransitive “Jack ___.”

Transitive, double object “Jack ___ Lily the thing.”



+ = allow, bring, pour, send, ...
- = bite, eat, laugh, sleep, understand...



Lidz & Gagliardi 2015

Target state: Useful verb classes

Children seem to begin forming a class of verbs used as **unaccusatives** by age two: Déprez & Pierce 1993, Snyder & Stromswold 1997, Bungler & Lidz 2004.

Transitive, single object “Jack ___ it.”

Unaccusative

“Jack ___.”
Patient-like

Passivizable “It was ___-en.”
Patient-like

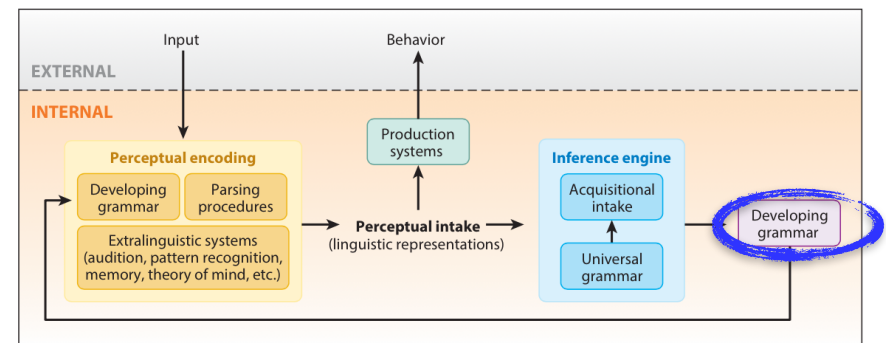
Intransitive “Jack ___.”

Transitive, double object “Jack ___ Lily the thing.”



+ = bounce, break, freeze, melt, ...

- = call, find, help, see, ...



Lidz & Gagliardi 2015

Target state: Useful verb classes

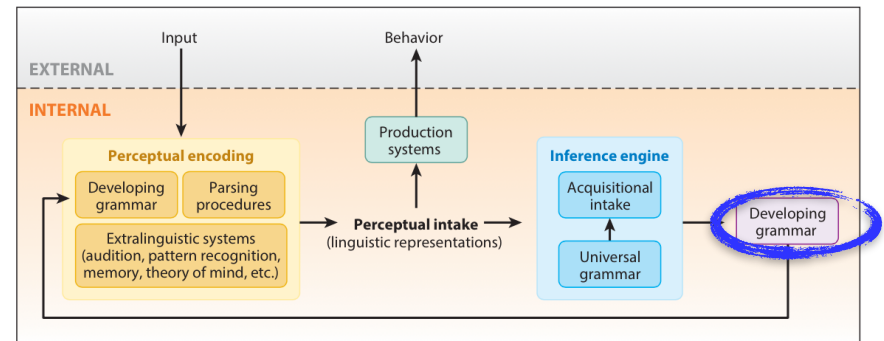
Children seem to begin forming a class of verbs used as **unergatives** by age two: Bunger & Lidz 2008.

- Transitive, single object “Jack ___ it.”
- Passivizable “It was ___-en.”
- Intransitive “Jack ___.”
- Transitive, double object “Jack ___ Lily the thing.”

- Unaccusative “Jack ___.”
Patient-like
- Unergative “Jack ___.”
Agent-like



- + = cry, dance, listen, play, ...
- = bounce, follow, push, shake, ...



Lidz & Gagliardi 2015

Target state: Useful verb classes

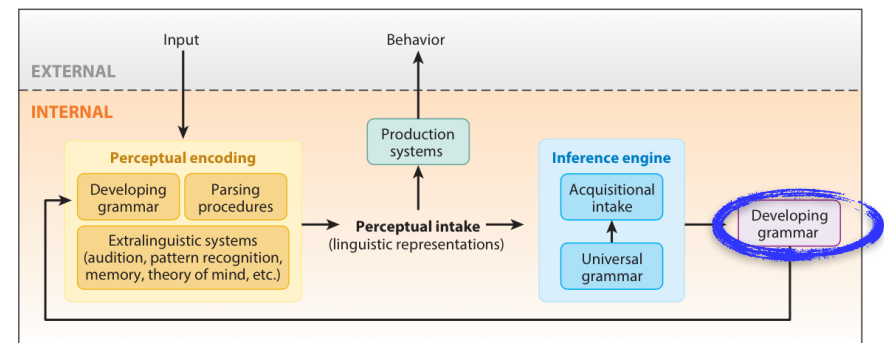
Children seem to begin forming a class of verbs that take *that-complements* by age three: Kidd, Lieven, & Tomasello 2006.

- Transitive, single object “Jack ___ it.”
- Passivizable “It was ___-en.”
Patient-like
- Intransitive “Jack ___.”
- Transitive, double object “Jack ___ Lily the thing.”

- Unaccusative “Jack ___.”
Patient-like
- Unergative “Jack ___.”
Agent-like
- that-complement* “Jack ___ that Lily’s nice.”

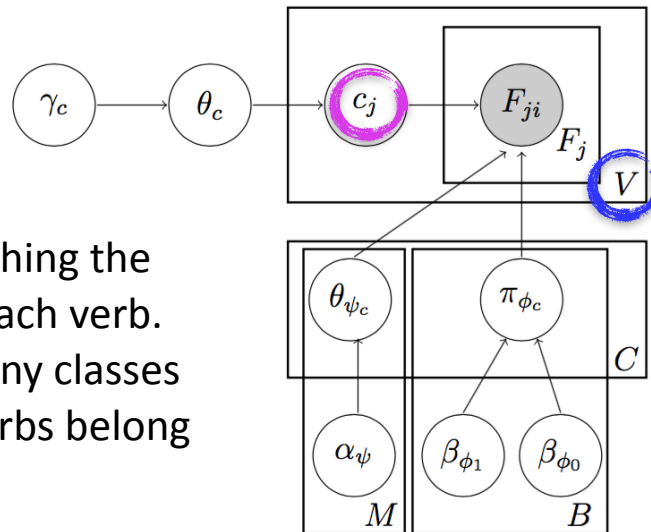


+ = care, decide, know, learn...
- = bounce, follow, push, shake,...



Lidz & Gagliardi 2015

Target state: Evaluating the results



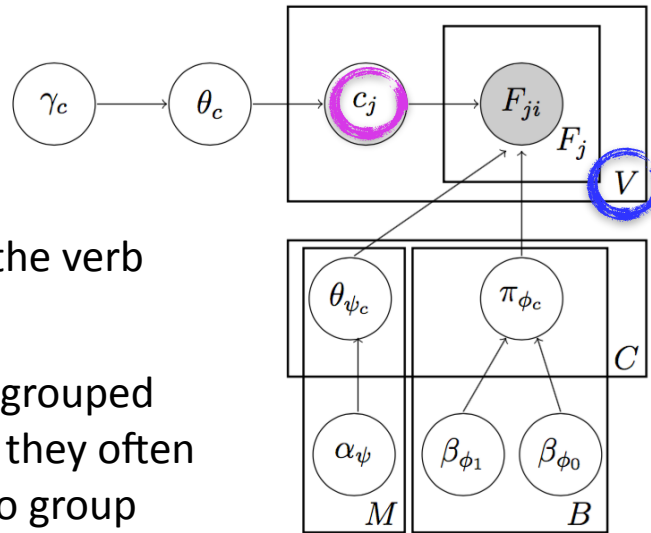
Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives

Remember: The **class** is the main thing the learner is trying to figure out for each verb. The learner doesn't know how many classes there are beforehand, or which verbs belong to which.

Target state: Evaluating the results



Each **verb** belongs to some **class** which determines its linguistic behavior.

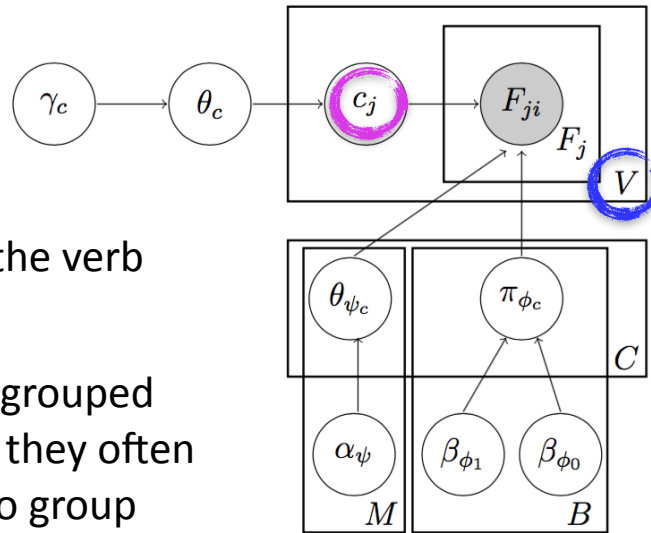
FALL

unaccusatives

Question: How **homogeneous** are the verb classes each learner infers?

That is, when we look at the verbs grouped together into an inferred class, are they often the same kind of verb? It's **useful** to group together verbs of the same kind.

Target state: Evaluating the results



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL *unaccusatives*

Question: How **homogeneous** are the verb classes each learner infers?

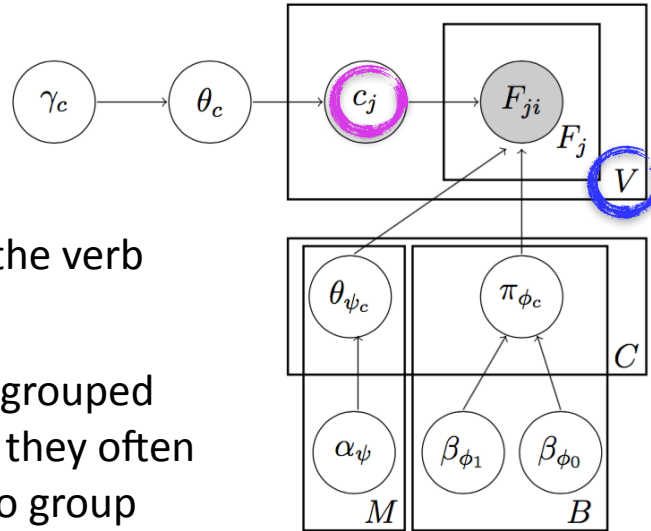
That is, when we look at the verbs grouped together into an inferred class, are they often the same kind of verb? It's **useful** to group together verbs of the same kind.

Implementation:
Random Index

☹️ ☺️
0.0 ≤ RI ≤ 1.0

Intuition: Get **credit** for **putting things together** that belong together and **keeping things apart** that don't belong together.

Target state: Evaluating the results



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives

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Implementation:

Random Index

For each pair of verbs in the inferred classes:

verb_i verb_j

☹️ ☺️
0.0 ≤ RI ≤ 1.0

Intuition: Get **credit** for **putting things together** that belong together and **keeping things apart** that don't belong together.

Inferred Class

Same class

Different class

True

Same class

True Positive

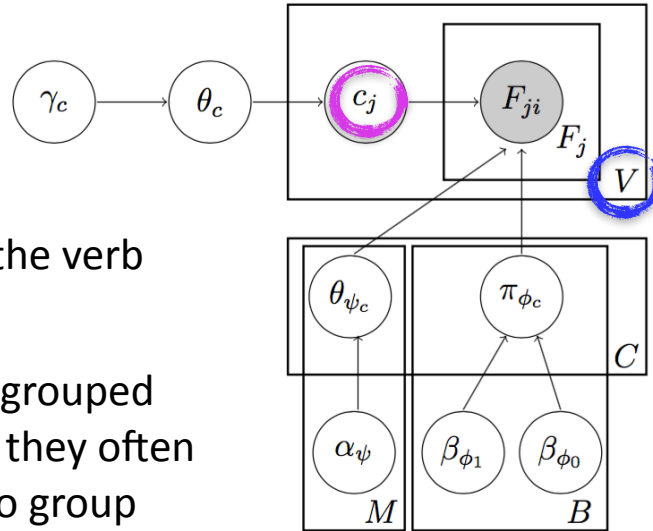
False Negative

Different class

False Positive

True Negative

Target state: Evaluating the results



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL

unaccusatives

Question: How **homogeneous** are the verb classes each learner infers?

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True

Same class

Different class

Inferred Class

Same class

Different class

True Positive

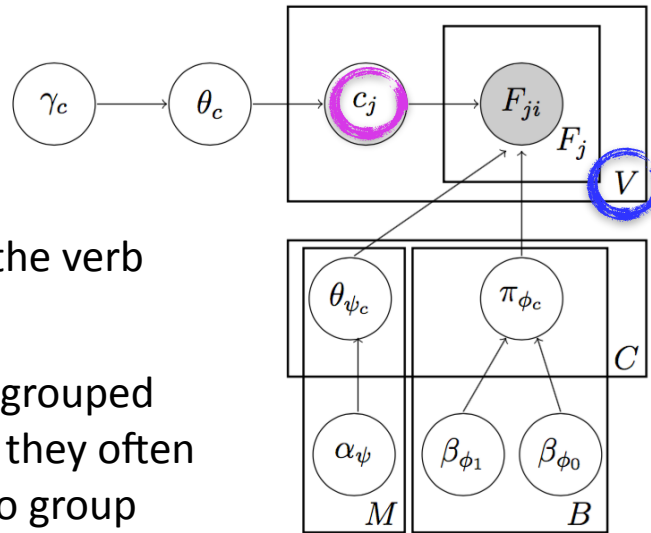
False Negative

False Positive

True Negative

$$\frac{\text{True Positives} + \text{True Negatives}}{\text{True Positives} + \text{True Negatives} + \text{False Positives} + \text{False Negatives}}$$

Target state: Evaluating the results



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL *unaccusatives*

Question: How **homogeneous** are the verb classes each learner infers?

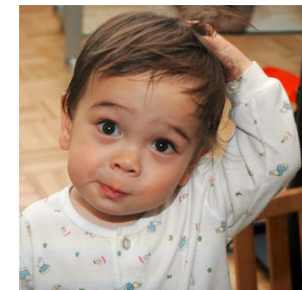
That is, when we look at the verbs grouped together into an inferred class, are they often the same kind of verb? It's **useful** to group together verbs of the same kind.

Implementation:
Random Index

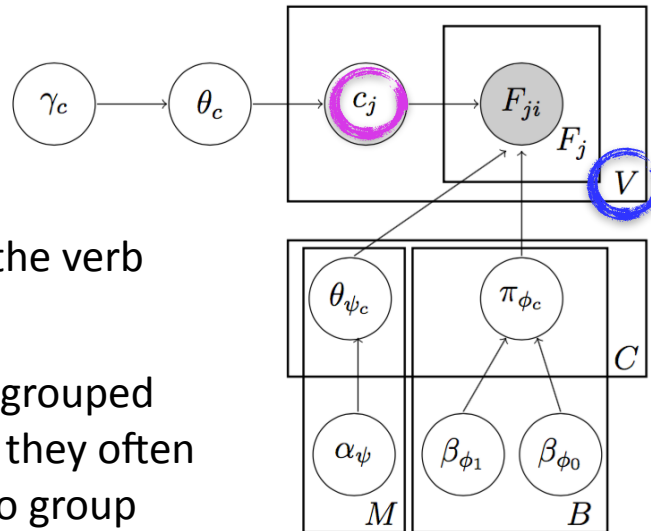
☹️ ☺️
0.0 ≤ RI ≤ 1.0

Intuition: Get **credit** for **putting things together that belong together** and **keeping things apart that don't belong together**.

But how do we know we're doing better than chance?



Target state: Evaluating the results



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL *unaccusatives*

Question: How **homogeneous** are the verb classes each learner infers?

That is, when we look at the verbs grouped together into an inferred class, are they often the same kind of verb? It's **useful** to group together verbs of the same kind.






Implementation:
Adjusted Random Index



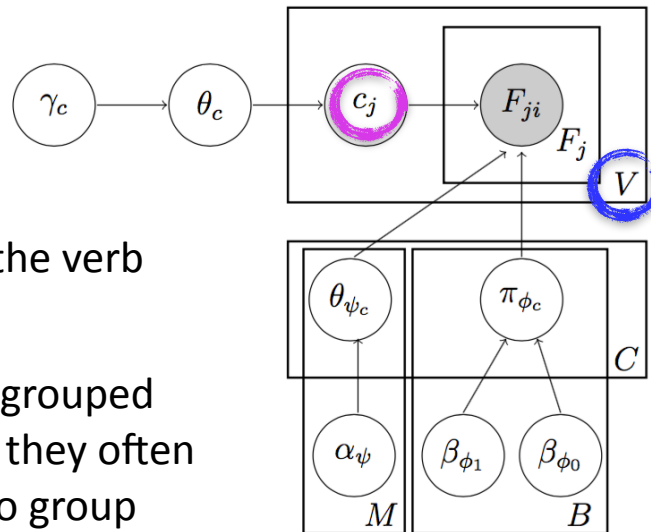

 $-1.0 \leq \text{ARI} \leq 1.0$



Compared against the expected value of the Random Index:

- 1.0 = perfect classification 
- >0 = better than chance 
- 0 = chance performance 
- <0 = worse than chance 
- 1.0 = perfectly awful performance 

Target state: Evaluating the results



Each **verb** belongs to some **class** which determines its linguistic behavior.

FALL *unaccusatives*

Question: How **homogeneous** are the verb classes each learner infers?

That is, when we look at the verbs grouped together into an inferred class, are they often the same kind of verb? It's **useful** to group together verbs of the same kind.




Implementation:
Adjusted Random Index




 $-1.0 \leq \text{ARI} \leq 1.0$

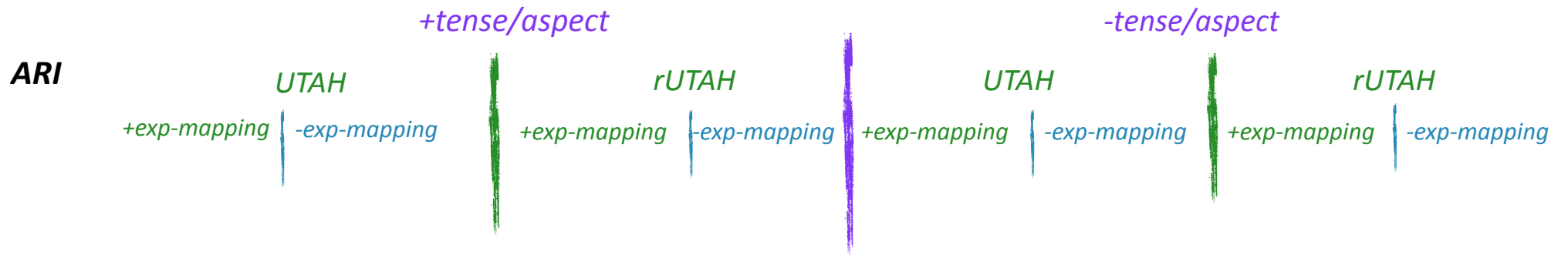


Compared against the expected value of the Random Index:

Useful	1.0 = perfect classification >0 = better than chance	
Not useful	0 = chance performance <0 = worse than chance -1.0 = perfectly awful performance	 

Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.



Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.



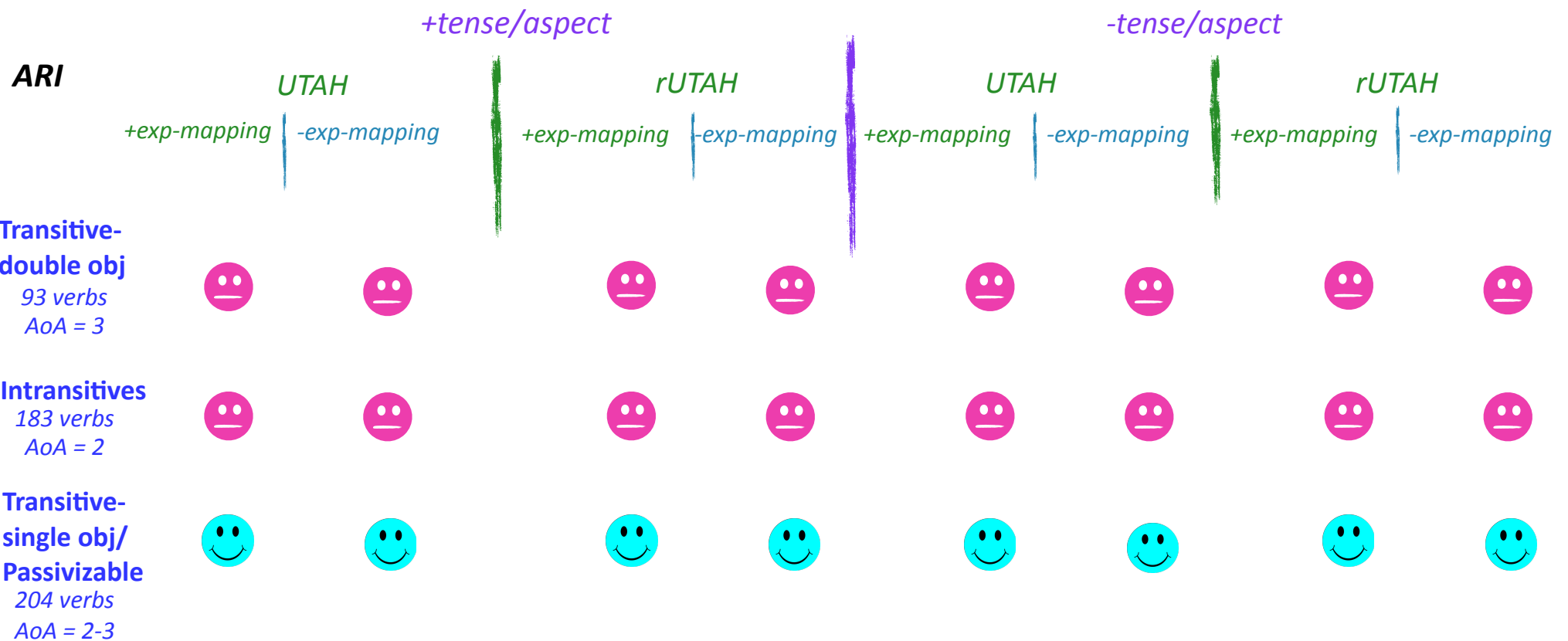
**Transitive-
double obj**
93 verbs
AoA = 3

Intransitives
183 verbs
AoA = 2

**Transitive-
single obj/
Passivizable**
204 verbs
AoA = 2-3

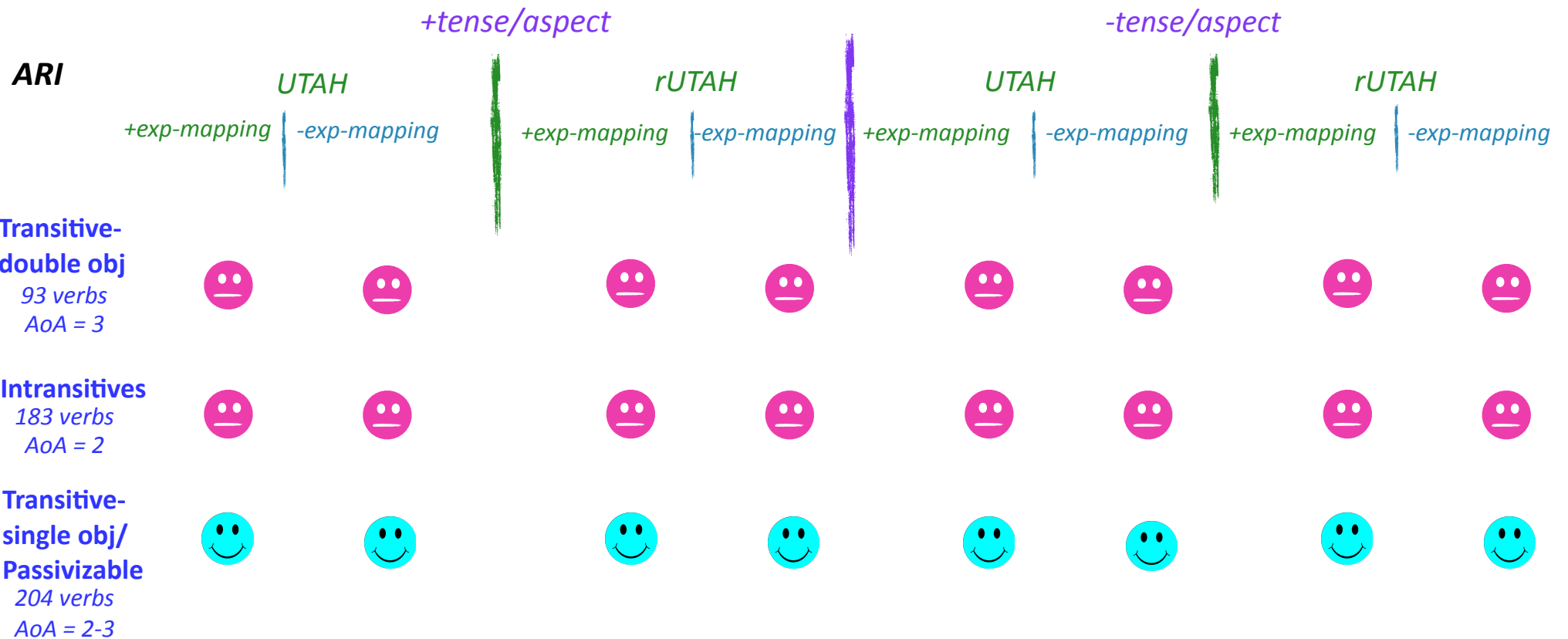
Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.



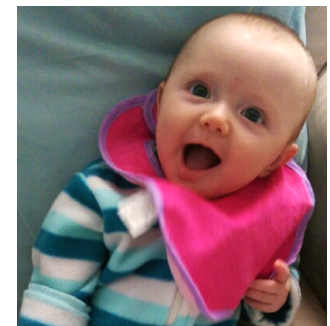
Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.



Learning which verbs **allow a single object** (and so are passivizable) is **easy** no matter which assumptions you use.

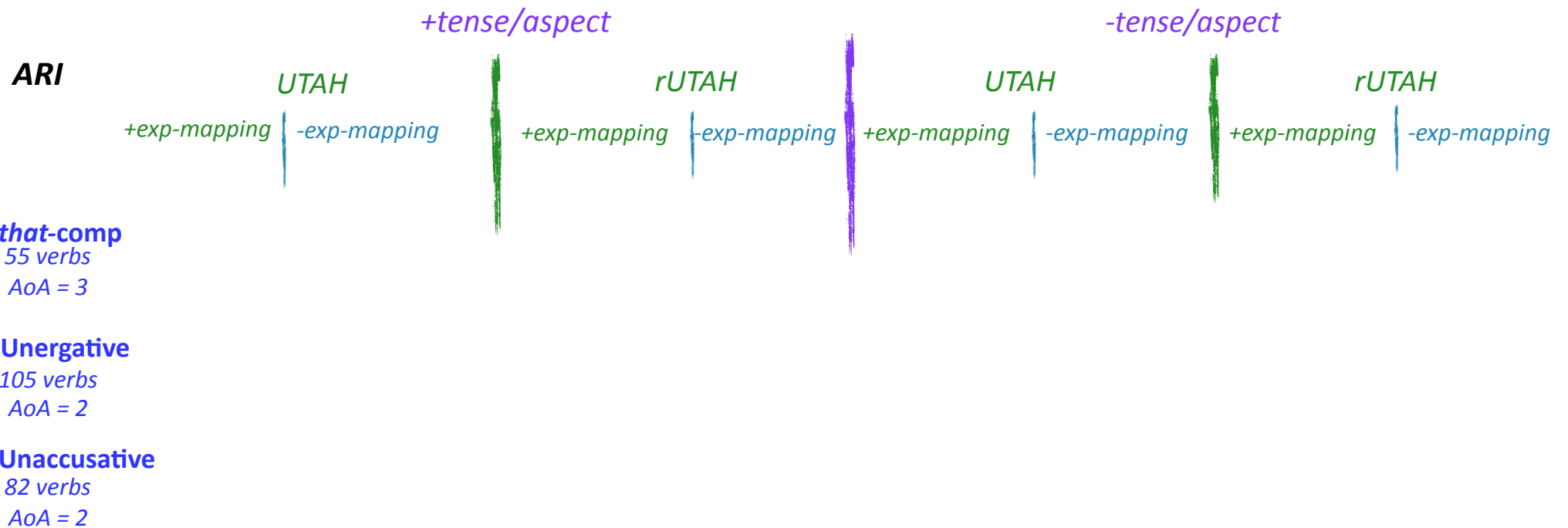
But learning which verbs **allow no objects or two objects** is **hard**, no matter which assumptions you use.



Pearl & Sprouse in progress

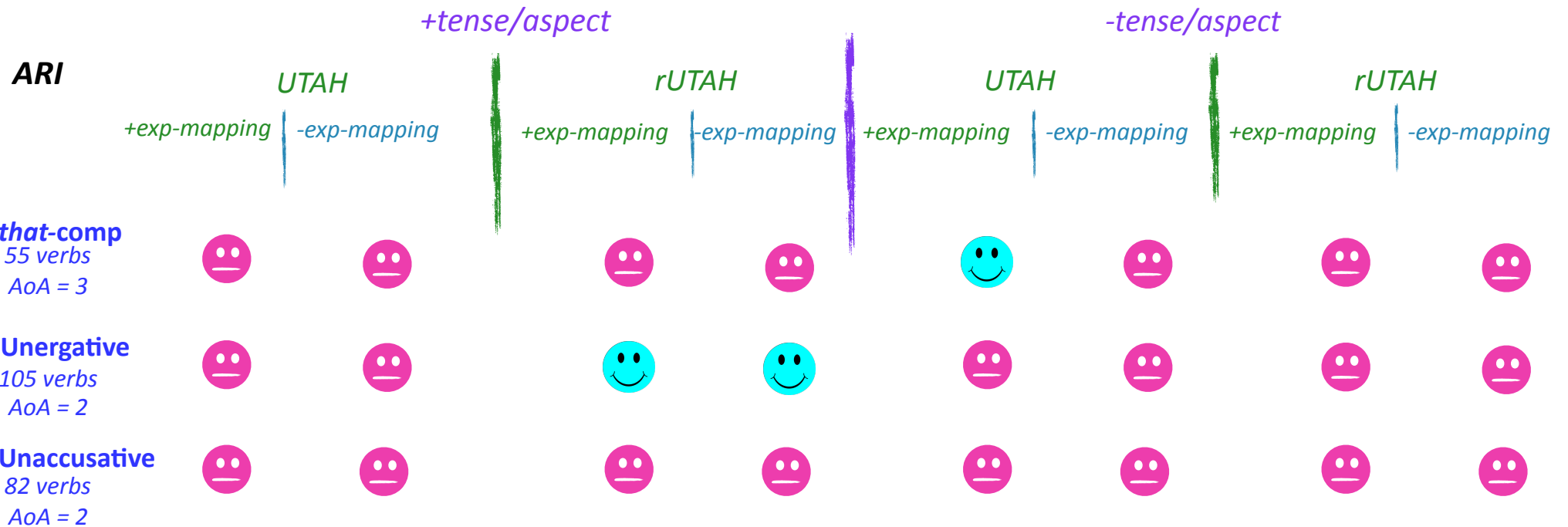
Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.



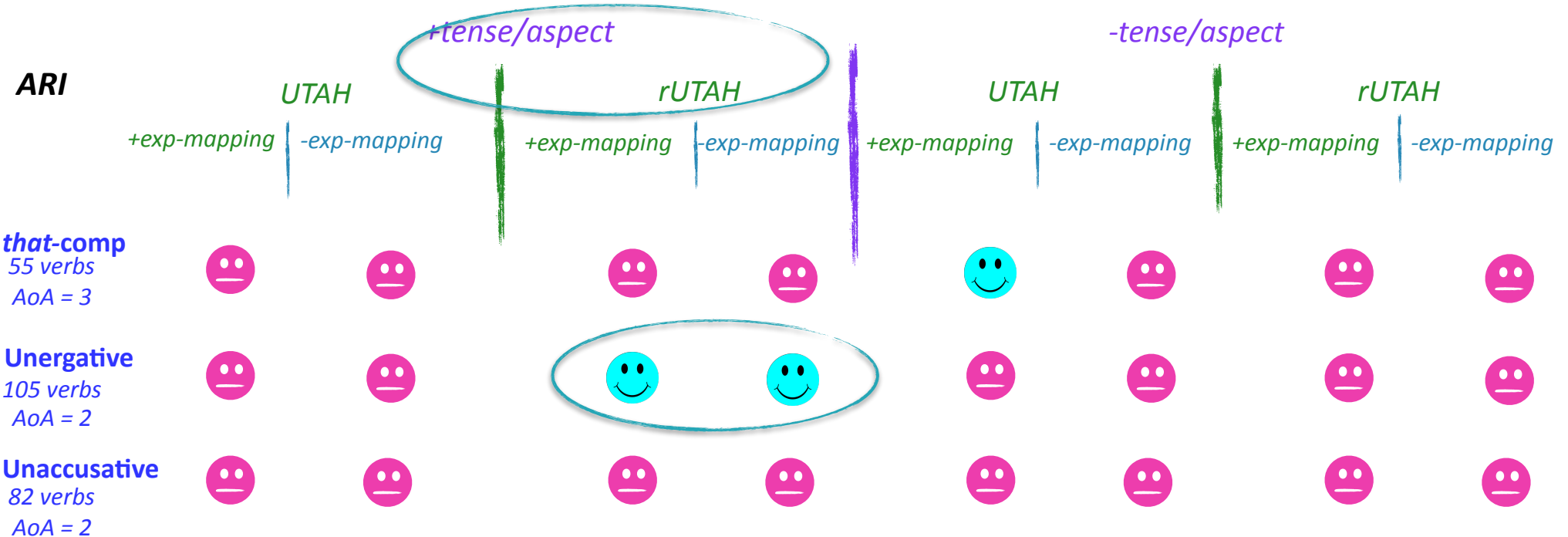
Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.



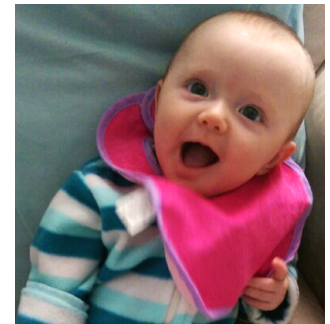
Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.



Now we see some differences:

While distinguishing **unaccusatives** is **hard no matter what**, distinguishing **unergatives** is **fine** if the **rUTAH** intermediate representation is used with **surface tense/aspect morphology** in the syntactic frames.



Pearl & Sprouse in progress

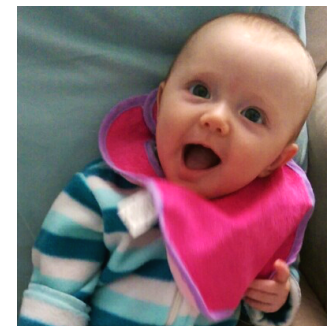
Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.

ARI	+tense/aspect				-tense/aspect			
	UTAH		rUTAH		UTAH		rUTAH	
	+exp-mapping	-exp-mapping	+exp-mapping	-exp-mapping	+exp-mapping	-exp-mapping	+exp-mapping	-exp-mapping
that-comp 55 verbs AoA = 3	☹️	☹️	☹️	☹️	😊	☹️	☹️	☹️
Unergative 105 verbs AoA = 2	☹️	☹️	😊	😊	☹️	☹️	☹️	☹️
Unaccusative 82 verbs AoA = 2	☹️	☹️	☹️	☹️	☹️	☹️	☹️	☹️

Now we see some differences:

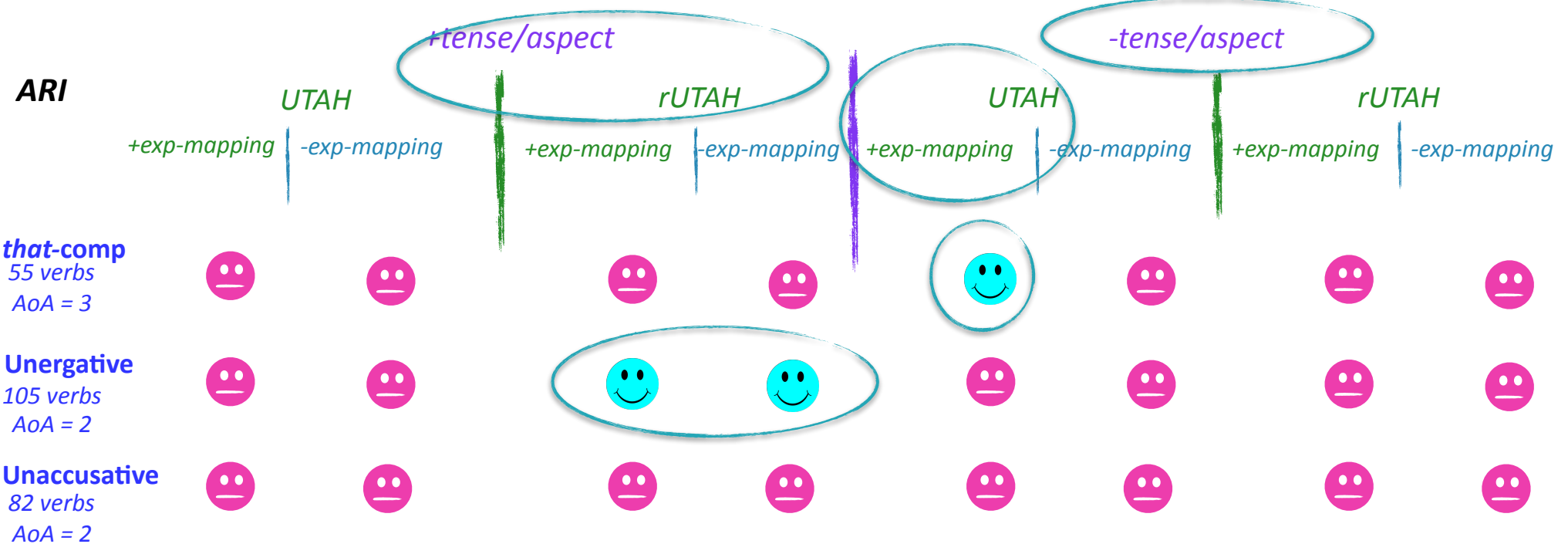
However, only using the **UTAH** intermediate representation with an **expectation of mapping** between that representation and syntactic positions as well as **ignoring surface tense/aspect morphology** will allow a learner to distinguish **that-complement** verbs from these data.



Pearl & Sprouse in progress

Target state: Evaluating the results

Distinctions made by two to three years of age, based on behavioral data.

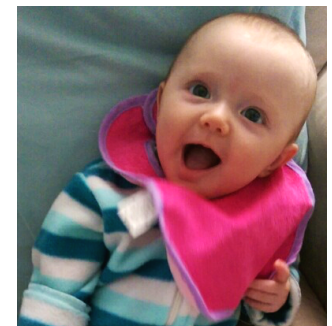


Big picture:

Three of these eight strategies seem to have a leg up on the rest when it comes to making the distinctions children should from these data.

Implication: These combinations of learning assumptions may be more on the right track than the others.

- rUTAH, +exp-mapping, +tense/aspect
- rUTAH, -exp-mapping, +tense/aspect
- UTAH, +exp-mapping, -tense/aspect



Pearl & Sprouse in progress

Learning strategy options

Syntax

She **melted** the ice with a blow dryer.

Subject

Object

Indirect Object

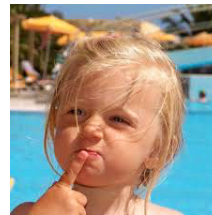


-tense/aspect info ← Choice 3 → +tense/aspect info

+exp-mapping:
movement is salient
because mapping to
syntax is fixed

Choice 2

-exp-mapping:
syntax mapping
distributions are
tracked



Mapping to Syntax

Two good
variants

UG knowledge

UTAH ← Choice 1 → rUTAH

Intermediate
representations



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

Thematic roles map to one
of three categories.

Thematic roles are ordered
with respect to each other.

thematic-roles

(likely derived from lower-level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Learning strategy options

Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object



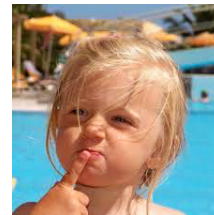
-tense/aspect info ← Choice 3 → +tense/aspect info

+exp-mapping: ← Choice 2
 movement is salient
 because mapping to
 syntax is fixed

-exp-mapping:
 syntax mapping
 distributions are
 tracked

Mapping to Syntax

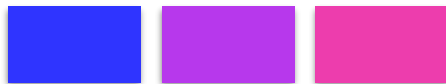
Another good
 variant



UG knowledge

UTAH ← Choice 1 → rUTAH

Intermediate
 representations



Agent > Experiencer >
 Theme > Patient >
 (Source, Goal, Instrument)

Thematic roles map to one
 of three categories.

Thematic roles are ordered
 with respect to each other.

thematic-roles

(likely derived from lower-level conceptual info) =
 Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Target state: Evaluating the results



But wait! Maybe children haven't figured out *every* verb in these classes by age three...

that-comp
55 verbs
AoA = 3

Unergative
105 verbs
AoA = 2

Unaccusative
82 verbs
AoA = 2

**Transitive-
double obj**
93 verbs
AoA = 3

Intransitives
183 verbs
AoA = 2

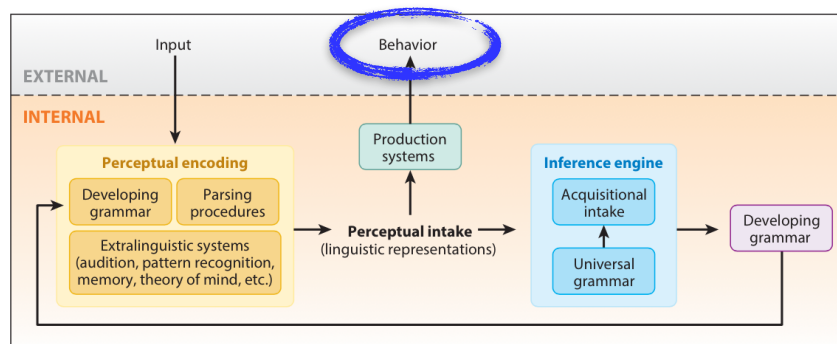
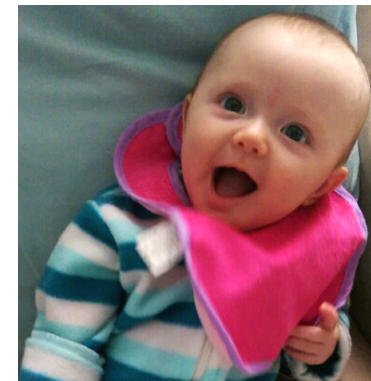
**Transitive-
single obj/
Passivizable**
204 verbs
AoA = 2-3

Target state: Evaluating the results



But wait! Maybe children haven't figured out *every* verb in these classes by age three...

Perhaps we should focus on the specific ones that have been **behaviorally attested** in children by age three.



Lidz & Gagliardi 2015

Pearl & Sprouse in progress

that-comp
7 verbs
AoA = 3

Unergative
105 verbs
AoA = 2

Unaccusative
5 verbs
AoA = 2

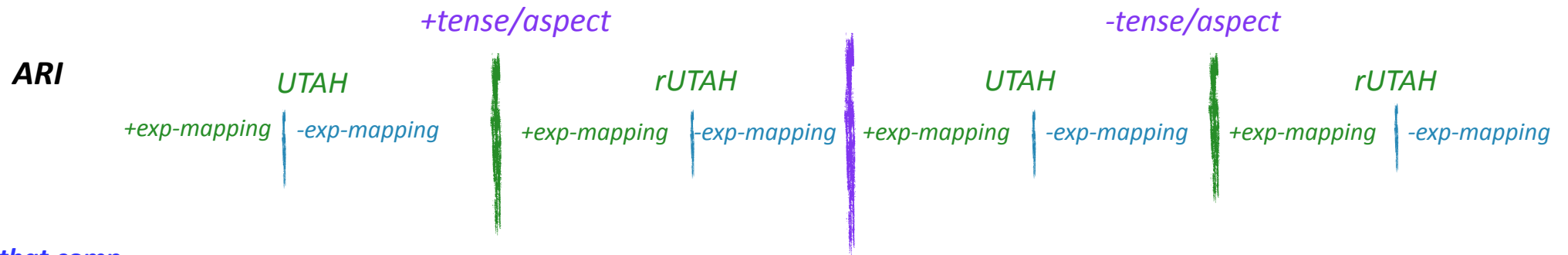
**Transitive-
double obj**
13 verbs
AoA = 3

Intransitives
183 verbs
AoA = 2

**Transitive-
single obj/
Passivizable**
24 verbs
AoA = 2-3

Target state: Evaluating the results

Matching the specific distinctions attested in behavioral studies (experimental & spontaneous speech).



that-comp
7 verbs

Unaccusative
5 verbs

Transitive-2obj
13 verbs

Passivizable
24 verbs

Target state: Evaluating the results

Matching the specific distinctions attested in behavioral studies (experimental & spontaneous speech).

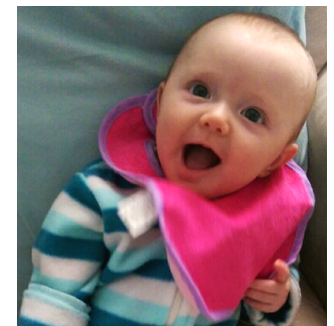
ARI	<i>+tense/aspect</i>				<i>-tense/aspect</i>			
	UTAH		rUTAH		UTAH		rUTAH	
	<i>+exp-mapping</i>	<i>-exp-mapping</i>	<i>+exp-mapping</i>	<i>-exp-mapping</i>	<i>+exp-mapping</i>	<i>-exp-mapping</i>	<i>+exp-mapping</i>	<i>-exp-mapping</i>
<i>that-comp</i> 7 verbs	😊	😊	😊	😊	😊	😊	😊	😊
<i>Unaccusative</i> 5 verbs	😊	😊	😊	😊	😊	😊	😊	😊
<i>Transitive-2obj</i> 13 verbs	😊	😊	😊	😊	😊	😊	😊	😊
<i>Passivizable</i> 24 verbs	😊	😞	😊	😊	😊	😊	😊	😊

Target state: Evaluating the results

Matching the specific distinctions attested in behavioral studies (experimental & spontaneous speech).

ARI	+tense/aspect				-tense/aspect			
	UTAH		rUTAH		UTAH		rUTAH	
	+exp-mapping	-exp-mapping	+exp-mapping	-exp-mapping	+exp-mapping	-exp-mapping	+exp-mapping	-exp-mapping
that-comp 7 verbs	😊	😊	😊	😊	😊	😊	😊	😊
Unaccusative 5 verbs	😊	😊	😊	😊	😊	😊	😊	😊
Transitive-2obj 13 verbs	😊	😊	😊	😊	😊	😊	😊	😊
Passivizable 24 verbs	😊	😞	😊	😊	😊	😊	😊	😊

Things don't look so hard anymore (except for passivizable verbs for one strategy variant). That's probably the only one we would rule out.



Pearl & Sprouse in progress

Seven out of eight good variants

Learning strategy options

Syntax

She melted the ice with a blow dryer.

Subject

Object

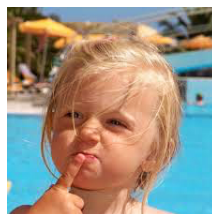
Indirect Object



-tense/aspect info ← Choice 3 → +tense/aspect info

+exp-mapping: movement is salient because mapping to syntax is fixed

-exp-mapping: syntax mapping distributions are tracked



Mapping to Syntax

UG knowledge

UTAH ← Choice 1 → rUTAH

Intermediate representations



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

Thematic roles map to one of three categories.

Thematic roles are ordered with respect to each other.

thematic-roles

(likely derived from lower-level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

Seven out of eight
good variants

Learning strategy options

How do we winnow this down?

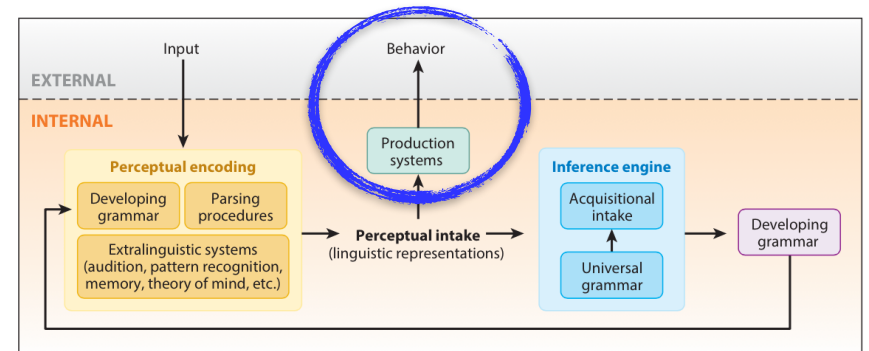


Seven out of eight
good variants

Learning strategy options

How do we winnow this down?

Maybe we need more **behavioral data** about which specific verb distinctions children make at this age. This could then distinguish between these strategies.



Lidz & Gagliardi 2015

Seven out of eight
good variants

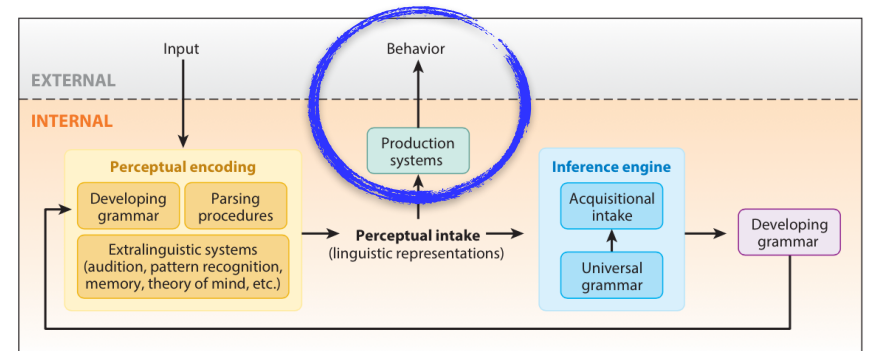
Learning strategy options

How do we winnow this down?

Maybe we need more **behavioral data** about which specific verb distinctions children make at this age. This could then distinguish between these strategies.

Example:

Verbs that are **ditransitive** and **passivizable** like *feed* and *give*



Lidz & Gagliardi 2015

Seven out of eight good variants

Learning strategy options

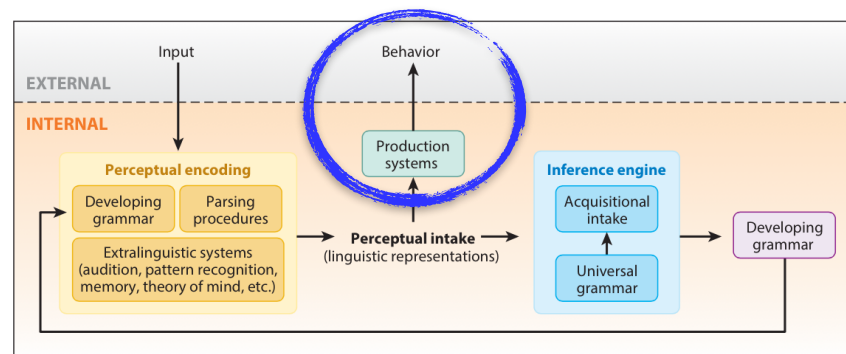
How do we winnow this down?

Maybe we need more **behavioral data** about which specific verb distinctions children make at this age. This could then distinguish between these strategies.

Example:

Verbs that are **ditransitive** and **passivizable** like *feed* and *give*

Do three-year-olds treat them the same?



Lidz & Gagliardi 2015

If **yes**, compatible with these:

- +tense/aspect, rUTAH, +exp-mapping
- +tense/aspect, rUTAH, -exp-mapping
- tense/aspect, UTAH, +exp-mapping
- tense/aspect, UTAH, -exp-mapping
- tense/aspect, rUTAH, +exp-mapping
- tense/aspect, rUTAH, -exp-mapping

NP V-ing NP



NP V NP

Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)

+exp-mapping:
movement is salient
because mapping to syntax is fixed

-exp-mapping:
syntax mapping distributions are tracked

If **no**, compatible with these:

- +tense/aspect, UTAH, +exp-mapping
- +tense/aspect, UTAH, -exp-mapping



NP V-ing NP

+exp-mapping:
movement is salient
because mapping to syntax is fixed

-exp-mapping:
syntax mapping distributions are tracked

Learning strategy options

How do we winnow this down?



Near future:

Test these learners on a larger data set to [combat potential data sparseness issues](#). (In progress: annotating the Brown-Adam corpus, which has about 20,000 more utterances.)

This also allows a larger age range of child-directed speech, extending up through age four. We can then investigate [performance on predicate distinctions children make at later ages](#).



Learning strategy options

How do we winnow this down?



Near future:

Test these learners on a larger data set to combat potential data sparseness issues. (In progress: annotating the Brown-Adam corpus, which has about 20,000 more utterances.)

This also allows a larger age range of child-directed speech, extending up through age four. We can then investigate performance on predicate distinctions children make at later ages.

Teaser: Even on these data directed at children under three, one strategy consistently does better at capturing the distinctions children will make at older ages.

(psych-subject experiencer verbs, psych-object experiencer verbs, raising-object verbs, raising-subject verbs, control-subject verbs, non-finite *to* complement verbs)

This is the same one that did better on *that*-complementizer verbs.

UTAH, +exp-mapping, -tense/aspect



+exp-mapping:
movement is salient
because mapping to syntax is fixed

NP V NP

What next?



Further future:

Alternative theories: Are there other options for linking thematic role information to syntactic structure that we can explore in this framework? What about linking conceptual information, if we're not so sure thematic roles are there?

thematic-roles

(likely derived from lower-level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...



Pearl & Sprouse in progress

What next?



Further future:

More sophisticated syntactic cues: What kind of structure is necessary for children to know in order to capture some of the more sophisticated distinctions they make at later ages? (It's likely a simple syntactic skeleton won't be enough...)

She melted the ice → NP V_{past} NP
The ice melted → NP V_{past}
The ice was melted → NP V_{past_participle}
The ice was melting → NP V_{progressive_participle}



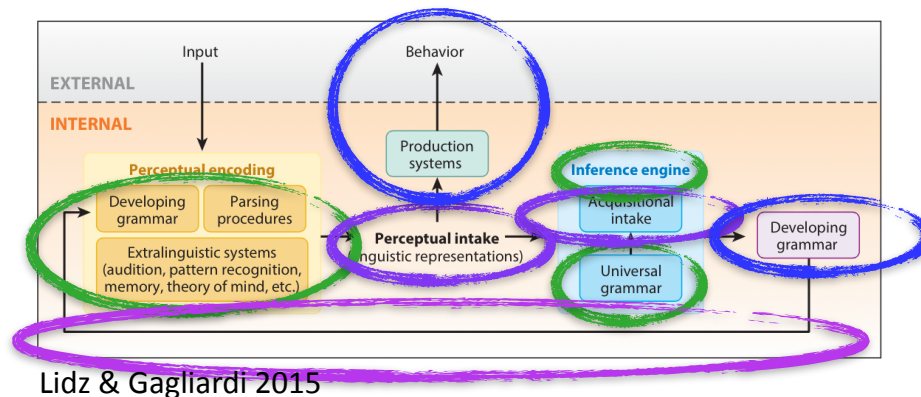
What next?



Further future:

More realistic assumptions about children:

- What if children only have some thematic roles available initially (and some syntactic structure), which they later build on? Do these theories still work/not work?
- What happens when we embed these theories in a learning model that learns incrementally (or at least in stages) and has cognitive constraints? For example, children might have one set of assumptions at age two, but a different set at age three based on the knowledge they've acquired.



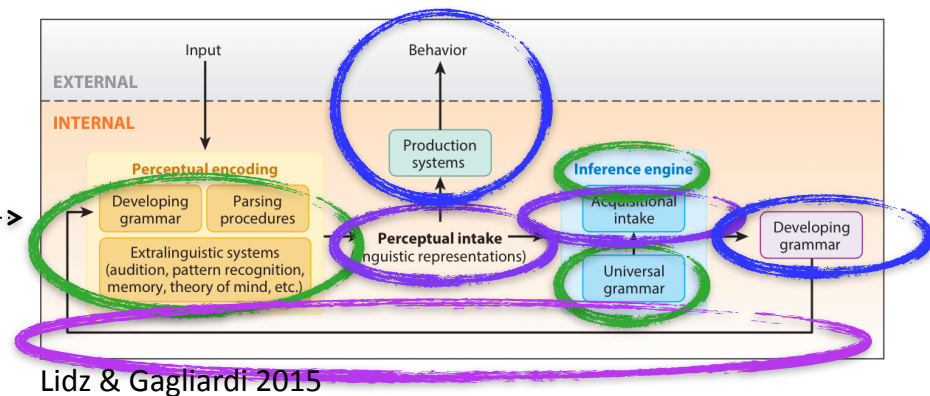
Pearl & Sprouse in progress

Big picture:

Understanding how children make syntactic generalizations

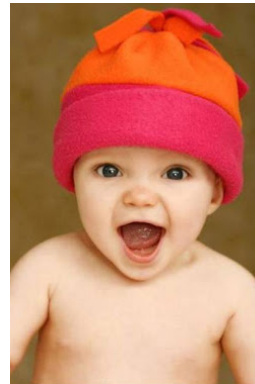
Precisely defining the components of a learning problem is necessary for making progress on how children solve that learning problem, which requires insights from many different empirical methods. This approach allows us to connect theories of linguistic representation and theories of language acquisition.

Given a specific initial state, a learner must use the data intake to reach the target state by the end of the learning period.



Biggest picture:
Computational acquisition modeling
for building integrated theories of acquisition

This technique is a useful tool — so let's use it to inform our theories of syntactic **representation** and **acquisition**!

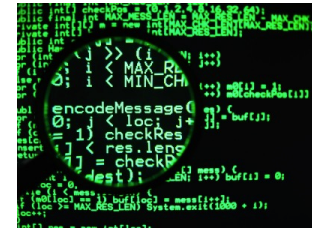
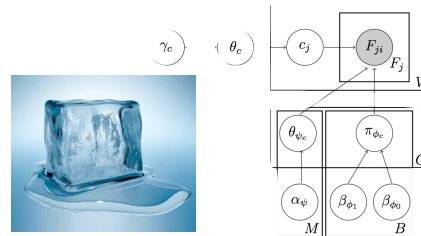


Thank you!

Jon Sprouse



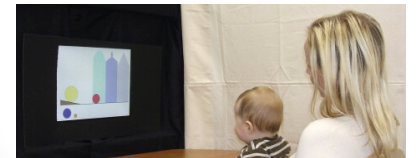
McGill University
Linguistics 2016



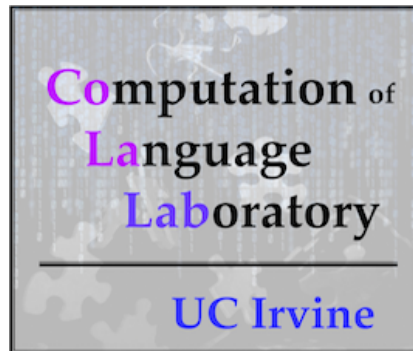
done-to

The ice melted.
The penguin climbed.

doer



Agent > Experiencer >
Theme > Patient >
(Source, Goal, Instrument)



This work was supported in part
by NSF grant BCS-1347028.



Special thanks to Abbie Thornton, Alandi Bates, Emily Yang, and BreAnna Silva for CHILDES Treebank corpus annotation.

Brown-Eve+Valian+Adam4yrs

Acquisitional intake: Input data

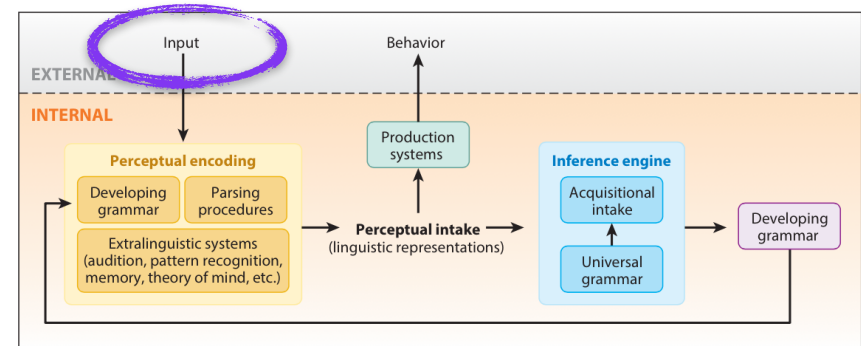
Brown-Eve+Valian+Adam4yrs



Data come from the Brown-Eve corpus (Brown 1973), the Valian corpus (Valian 1991), and the Brown-Adam corpus (Brown 1973) directed at age four, with syntactic & thematic annotations provided by the CHILDES Treebank (Pearl & Sprouse 2013).

This corpus (Brown-Eve+Valian+Adam4yrs) contains speech directed at 23 children between the ages of 18 and 58 months.

There are ~45,000 utterances total, comprised of ~224,000 word tokens. Of the 603 verb lexical items that appear, 253 occur 5 or more times.



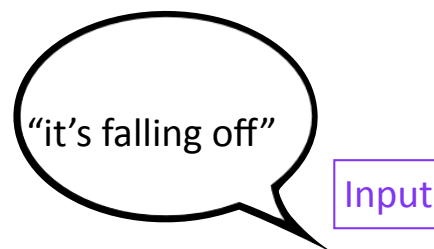
Lidz & Gagliardi 2015

Acquisitional intake: Input data

Brown-Eve+Valian+Adam4yrs



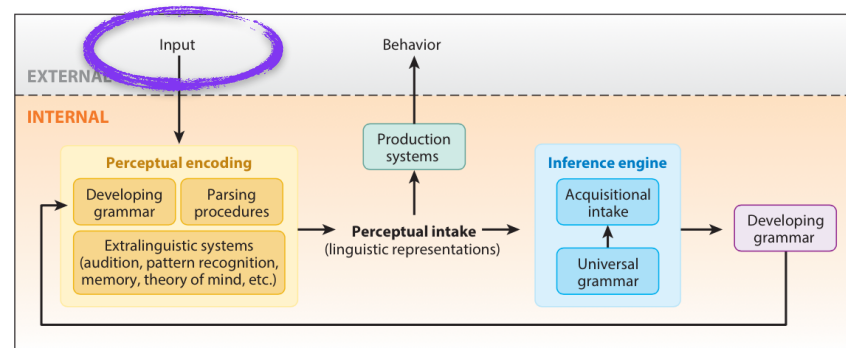
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This corpus (Brown-Eve+Valian+Adam4yrs) contains speech directed at 23 children between the ages of 18 and 58 months.

There are ~45,000 utterances total, comprised of ~224,000 word tokens. Of the 603 verb lexical items that appear, 253 occur 5 or more times.

Focus on learning the predicate categories for these for now.
Intuition: Frequent enough to be useful to distributionally learn from.

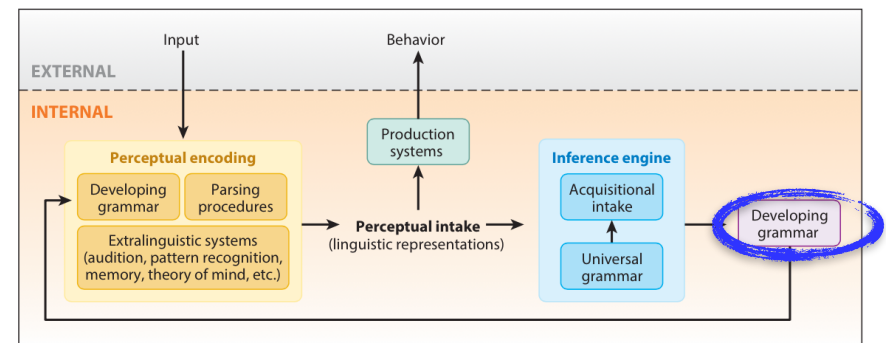


Lidz & Gagliardi 2015

Target state: Useful verb classes

Adult knowledge is the eventual target state for acquisition, and there are a variety of verb distinctions that have different syntactic and/or thematic role implications. Do some of these distinctions fall out directly by using the syntactic, conceptual, and semantic-syntactic cues we're using?

Given the input data we have from the Brown-Eve-Valian+Adam corpus (which is directed at children age 4;10 and younger), we can include distinctions children seem to have made by age five when we learn from those data.



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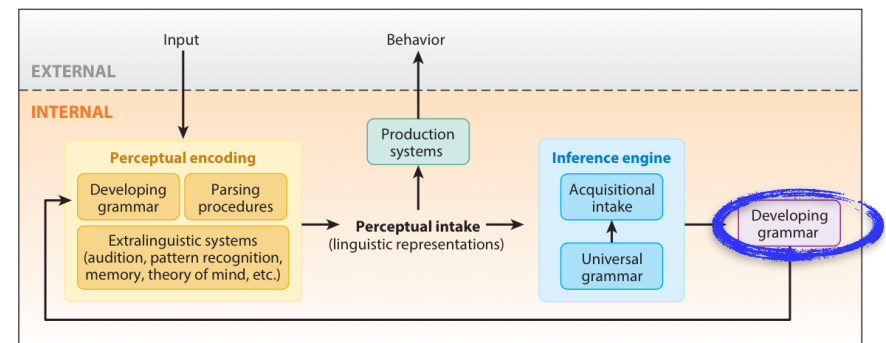
Target state: Useful verb classes

Children seem to figure out **object-experiencer psych verbs** before **subject-experiencer psych verbs** in English, though they seem to sort them both out by age 4 or 5 (Hartshorne, Pogue, & Snedeker 2015).

Psych, object experiencer “It ___ Jack.”
Causer Experiencer



+ = bother, confuse, scare, worry...
- = fall, go, kick, stare...



Lidz & Gagliardi 2015

Target state: Useful verb classes

Children seem to figure out **object-experiencer psych verbs** before **subject-experiencer psych verbs** in English, though they seem to sort them both out by age 4 or 5 (Hartshorne, Pogue, & Snedeker 2015).

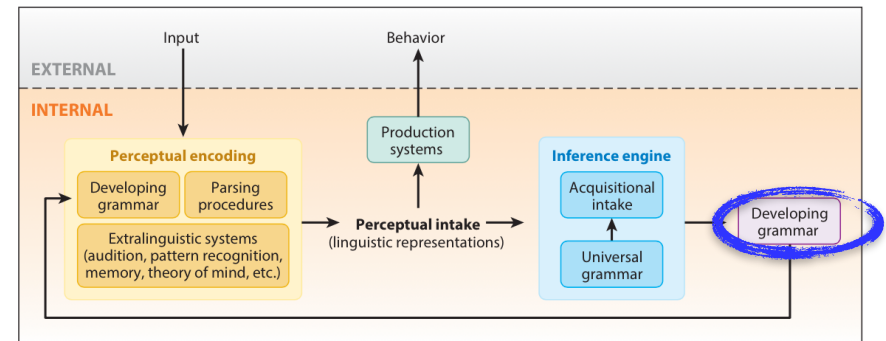
Psych, object experiencer “It ___ Jack.”
Causer Experiencer

Psych, subject experiencer “Jack ___ it.”
Experiencer SubjectMatter



+ = like, love, miss, want...

- = fall, go, kick, stare...



Lidz & Gagliardi 2015

Target state: Useful verb classes

By 4 to 5 years old, English children can use animacy information when distinguishing between **control-object** and **raising-object** verbs (Kirby 2009, 2010, 2011).

Psych, object experiencer “It ___ Jack.”
Causer Experiencer

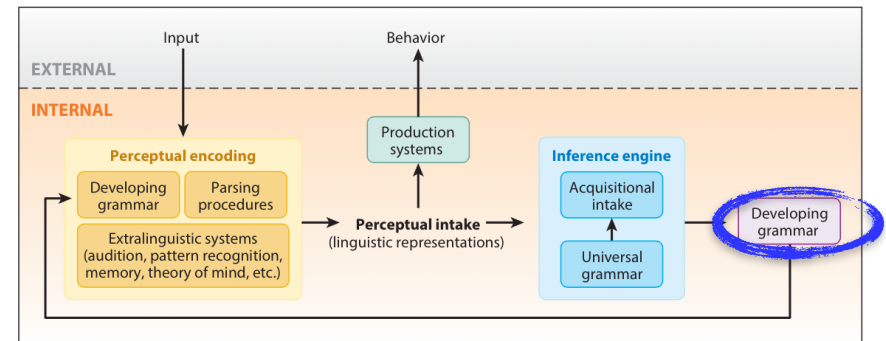
Psych, subject experiencer “Jack ___ it.”
Experiencer SubjectMatter

Control-object “Jack ___ her to win.”
Agent-like Goal-like



+ = ask, tell, teach, thank...

- = fall, go, kick, stare...



Lidz & Gagliardi 2015

Target state: Useful verb classes

By 4 to 5 years old, English children can use animacy information when distinguishing between **control-object** and **raising-object** verbs (Kirby 2009, 2010, 2011).

Psych, object experiencer “It ___ Jack.”
Causer Experiencer

Psych, subject experiencer “Jack ___ it.”
Experiencer SubjectMatter

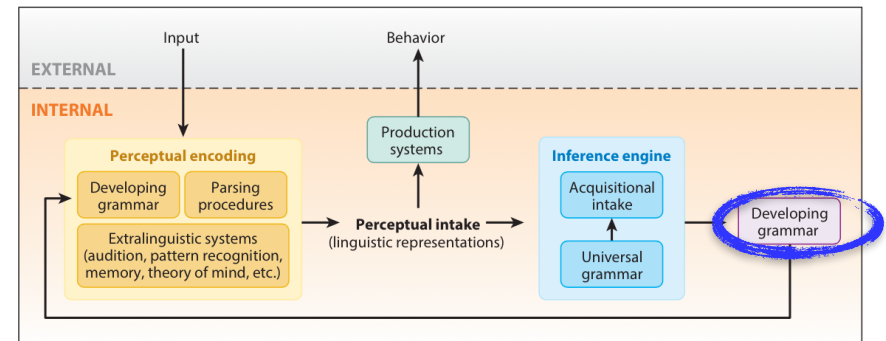
Control-object “Jack ___ her to win.”
Agent-like Goal-like

Raising-object (ECM) “Jack ___ her to win.”
Agent-like



+ = knew, mean, need, take...

- = fall, go, kick, stare...



Lidz & Gagliardi 2015

Target state: Useful verb classes

By 4 to 5 years old, English children have figured out that inanimate subjects can distinguish between **raising-subject** and **control-subject** verbs (Becker 2006, 2007, 2009, 2014). In particular, raising-subject verbs allow inanimate subjects. So, they've likely figured out these classes.

Psych, object experiencer “It ___ Jack.”
Causer Experiencer

Raising-subject

“Jack ___ to win.”
Agent-like

Psych, subject experiencer “Jack ___ it.”
Experiencer SubjectMatter

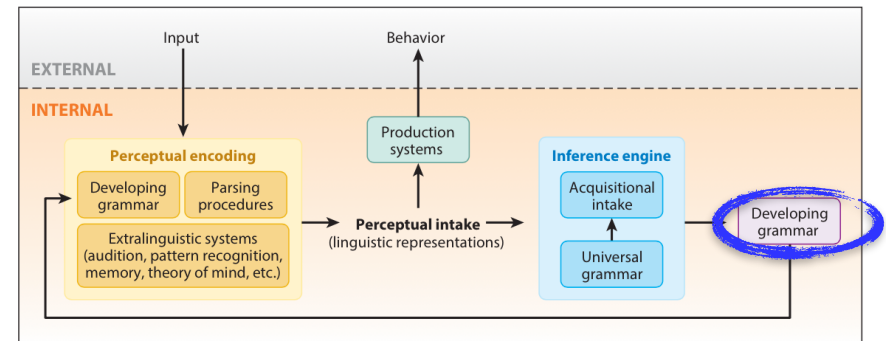
Control-object “Jack ___ her to win.”
Agent-like Goal-like

Raising-object (ECM) “Jack ___ her to win.”
Agent-like



+ = begin, happen, seem, use...

- = fall, go, kick, stare...



Lidz & Gagliardi 2015

Target state: Useful verb classes

By 4 to 5 years old, English children have figured out that inanimate subjects can distinguish between raising-subject and control-subject verbs (Becker 2006, 2007, 2009, 2014). In particular, raising-subject verbs allow inanimate subjects. So, they've likely figured out these classes.

Psych, object experiencer "It ___ Jack."
Causer Experiencer

Raising-subject "Jack ___ to win."
Agent-like

Psych, subject experiencer "Jack ___ it."
Experiencer SubjectMatter

Control-subject "Jack ___ to win."
Agent-like₁ Agent-like₂

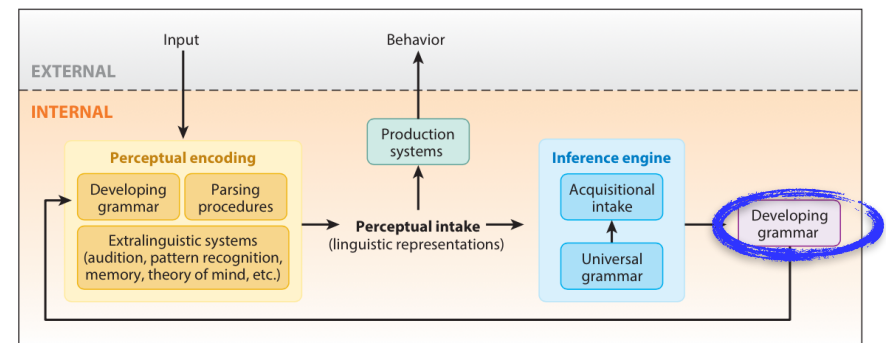
Control-object "Jack ___ her to win."
Agent-like Goal-like

Raising-object (ECM) "Jack ___ her to win."
Agent-like



+ = decide, like, try, want...

- = fall, go, kick, stare...



Lidz & Gagliardi 2015

Target state: Useful verb classes

By 5 years old, English children use *whether/if*-complement taking verbs in their spontaneous speech (Diessel & Tomasello 2001), which may indicate they've formed a class of these verbs.

Psych, object experiencer "It ___ Jack."
Causer Experiencer

Raising-subject "Jack ___ to win."
Agent-like

Psych, subject experiencer "Jack ___ it."
Experiencer SubjectMatter

Control-subject "Jack ___ to win."
Agent-like₁ Agent-like₂

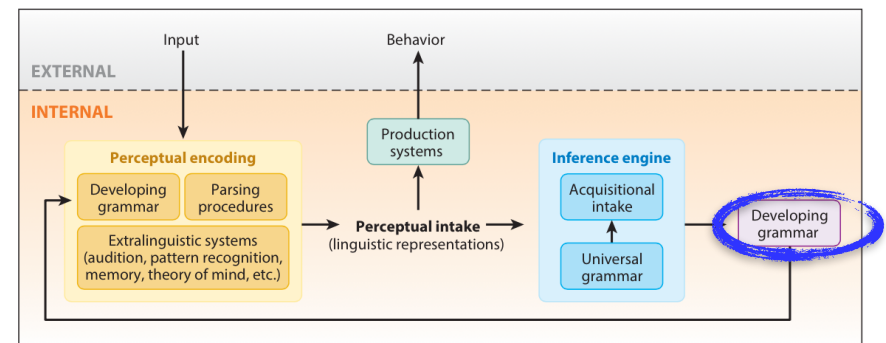
Control-object "Jack ___ her to win."
Agent-like Goal-like

whether/if-complement "Jack ___ whether we won."

Raising-object (ECM) "Jack ___ her to win."
Agent-like



+ = decide, forget, know, wonder...
 - = fall, go, kick, stare...



Lidz & Gagliardi 2015

Thematic roles & how to use them

One idea about how children could use thematic role information: (r)UTAH.

The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis

UG knowledge

UTAH: Baker 1988, Baker 1997, Dowty 1991, Fillmore 1968, Grimshaw 1990, Jackendoff 1987, Perlmutter & Postal 1984, Speas 1990

Each thematic role maps to a specific syntactic position (grammatical role).

control



She **tried** to **melt** the ice.

doer
doer

done-to

*It **tried** that she **melted** the ice.

doer

done-to



The penguin **seemed** to **climb** the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

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Each thematic role maps to a specific syntactic position (grammatical role).

Agent-like = grammatical subject

Agent
Causer
Experiencer
Possessor



("internal cause" = Rappaport-Hovav 1995)

control

She **tried** to melt the ice.

doer
doer

done-to

*It **tried** that she **melted** the ice.

doer

done-to



The penguin **seemed** to climb the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

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Each thematic role maps to a specific syntactic position (grammatical role).

Agent-like = grammatical subject

Agent

Causer

Experiencer (**Baker: only when subject*)

Possessor

(“internal cause” = Rappaport-Hovav 1995)



control

She **tried** to melt the ice.

doer
doer

done-to

*It **tried** that she **melted** the ice.

doer

done-to

She **fears** spiders.

Experiencer



Spiders **frighten** her.

Experiencer



The penguin **seemed** to climb the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

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The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis

UG knowledge

UTAH: Baker 1988, Baker 1997, Dowty 1991, Fillmore 1968, Grimshaw 1990, Jackendoff 1987, Perlmutter & Postal 1984, Speas 1990

Each thematic role maps to a specific syntactic position (grammatical role).

Agent-like = grammatical subject

Patient-like = grammatical object

Patient

Theme

Experiencer

Subject Matter

(“external cause”)



control

She **tried** to **melt** the ice.

doer
doer

done-to

*It **tried** that she **melted** the ice.

doer

one-to



The penguin **seemed** to **climb** the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

Thematic roles & how to use them

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The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis

UG knowledge

UTAH: Baker 1988, Baker 1997, Dowty 1991, Fillmore 1968, Grimshaw 1990, Jackendoff 1987, Perlmutter & Postal 1984, Speas 1990

Each thematic role maps to a specific syntactic position (grammatical role).

Agent-like = grammatical subject

Patient-like = grammatical object

Patient

Theme

Experiencer (*Baker: only when not subject)

Subject Matter

("external cause")



control

She **tried** to **melt** the ice.

doer
doer

done-to

*It **tried** that she **melted** the ice.

doer

done-to

Patient

Theme

Experiencer (*Baker: only when not subject)

Subject Matter

("external cause")

She **fears** spiders.

Experiencer



Spiders **frighten** her.

Experiencer



The penguin **seemed** to **climb** the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

Thematic roles & how to use them

One idea about how children could use thematic role information: (r)UTAH.

The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis

UG knowledge

UTAH: Baker 1988, Baker 1997, Dowty 1991, Fillmore 1968, Grimshaw 1990, Jackendoff 1987, Perlmutter & Postal 1984, Speas 1990

Each thematic role maps to a specific syntactic position (grammatical role).

Agent-like = grammatical subject

Patient-like = grammatical object

Goal-like = grammatical indirect object

Location

Source

Goal

Benefactor

Instrument



control

She **tried** to **melt** the ice with a blow dryer.

doer
doer

done-to

done-with

*It **tried** that she **melted** the ice with a blow dryer.

doer

done-to

done-with



The penguin **seemed** to **climb** the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

Thematic roles & how to use them



Syntax

She melted the ice with a blow dryer.

Subject

Object

Indirect Object

Mapping to Syntax

UG knowledge

The **U**niformity of **T**heta **A**ssignment **H**ypothesis:

Baker 1988, Baker 1997, Dowty 1991, Fillmore 1968, Grimshaw 1990, Jackendoff 1987, Perlmutter & Postal 1984, Speas 1990

UTAH

Intermediate representations



Thematic roles map to one of three categories.

(likely derived from lower level conceptual info) =
Agent, Experiencer, Patient, Theme, Goal, Source, Instrument...

thematic-roles

Thematic roles & how to use them

One idea about how children could use thematic role information: (r)UTAH.

The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis

UG knowledge

rUTAH: Larson 1988, Larson 1990

Thematic roles are ordered relative to each other, with the highest thematic role mapping to the highest grammatical role (subject > object > indirect object).

control



She **tried** to **melt** the ice with a blow dryer.

doer
doer

done-to

done-with

*It **tried** that she **melted** the ice with a blow dryer.

doer

done-to

done-with



The penguin **seemed** to **climb** the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

Thematic roles & how to use them

One idea about how children could use thematic role information: (r)UTAH.

The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis

UG knowledge

rUTAH: Larson 1988, Larson 1990

Thematic roles are ordered relative to each other, with the highest thematic role mapping to the highest grammatical role (subject > object > indirect object).

Basic intuition:

doer (Agent-like) >

done-to (Patient-like) >

done-for/with (Goal-like)



control

She **tried** to **melt** the ice with a blow dryer.

doer
doer

done-to

done-with

*It **tried** that she **melted** the ice with a blow dryer.

doer

done-to

done-with



The penguin **seemed** to **climb** the hill.

doer

done-to

It **seemed** that the penguin **climbed** the hill.

doer

done-to

raising

Thematic roles & how to use them

One idea about how children could use thematic role information: (r)UTAH.

The (relativized) **U**niformity of **T**heta **A**ssignment **H**ypothesis

UG knowledge

rUTAH: Larson 1988, Larson 1990

Thematic roles are ordered relative to each other, with the highest thematic role mapping to the highest grammatical role (subject > object > indirect object).

Basic intuition:

doer (Agent-like) >

done-to (Patient-like) >

done-for/with (Goal-like)



An example implementation:

Agent > Causer > Experiencer > Possessor >

Subject Matter > Causee > Theme > Patient >

(Location, Source, Goal, Benefactor, Instrument)



control

She **tried** to **melt** the ice with a blow dryer.

doer
doer

done-to

done-with

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Note: You don't need to have every role relatively ranked. If some are unranked with respect to each other, the order in which they get mapped to grammatical positions doesn't matter.

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doer

done-to

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This relative ranking can help deal with certain situations, like those involving Experiencers.

She fears spiders.

Experiencer Subject Matter

Experiencer > Subject Matter

Subject Object



Spiders frighten her.

Causer Experiencer

Causer > Experiencer

Subject Object

What next?



Near-ish future:

Other ways to evaluate the output of the modeled learners.

- (1) **Qualitative analysis**: Which verbs of each class is a learner consistently getting right? Are these more important/more useful in some respect? What do the errors look like, and do they look like the kind of thing children do? (Behavioral data on specific verbs gets at this somewhat already.)
- (2) **Utility of inferred classes**: Can we identify a specific acquisition task that depends on verb classes, and see if the inferred classes are useful for that task (Phillips & Pearl 2015, Bar-Sever & Pearl 2016)? This can tell us if they're good classes, even if they don't match adult verb classes.



Learning strategy option refinement: The bigger picture

The Linking Problem: Pearl & Sprouse in progress

Refining ideas about what **implementations of prior knowledge** are consistently useful for acquisition (Ambridge et al. 2014, Pearl 2014):

Not: UTAH & -exp-mapping if using surface tense/aspect information

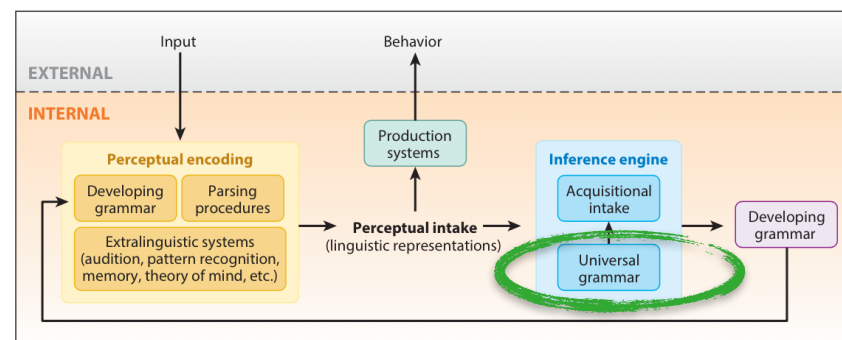


done-to

The ice melted.

The penguin climbed.

doer



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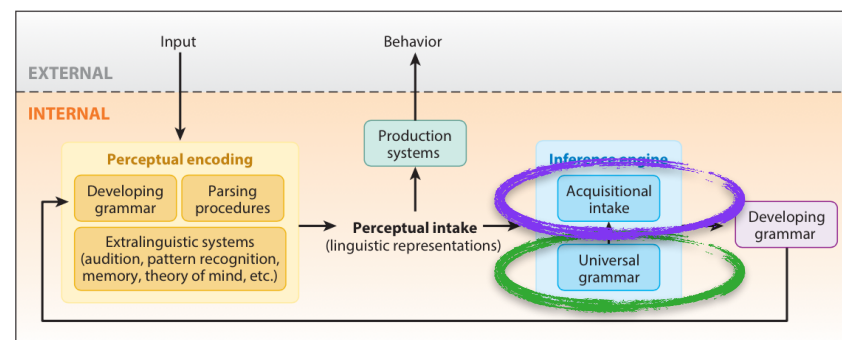


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Larger point: Connection between theories of **linguistic representation** and theories of **language acquisition**

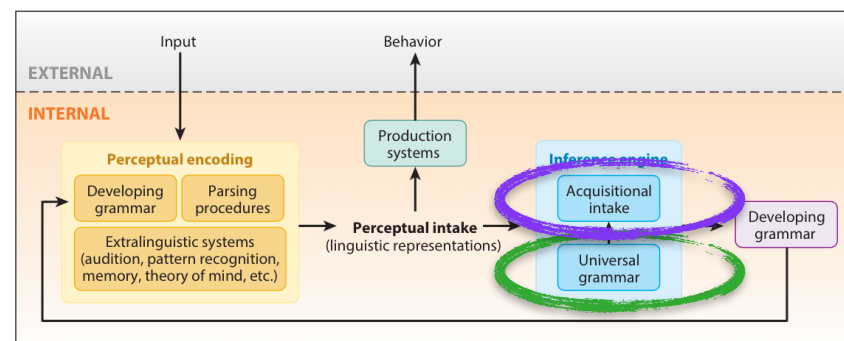


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