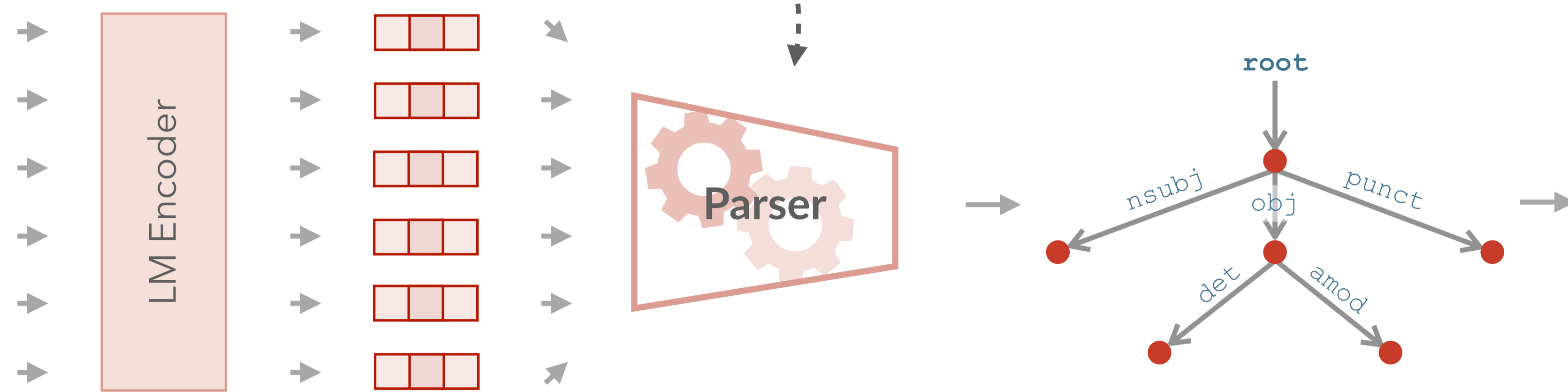


# Probing for Labeled Dependency Trees

Max Müller-Eberstein, Rob van der Goot and Barbara Plank

Training a full biaffine attention parser (BAP) on mBERT: **183M** parameters.

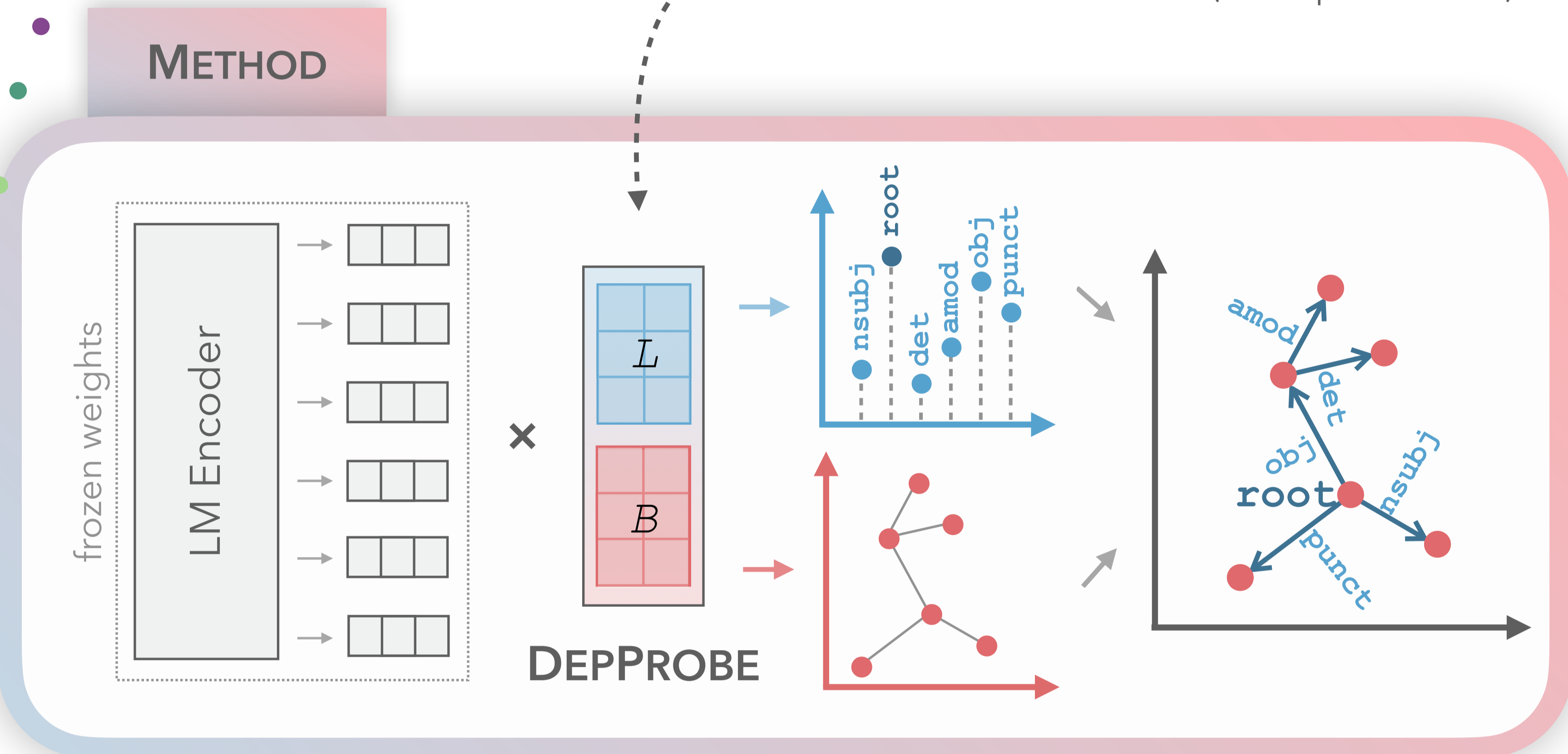


How accurately can these **169** transfer settings be predicted **w/o training a full parser**?

	AR	EN	EU	FI	HE	HI	IT	JA	KO	RU	SV	TR	ZH
AR	83	32	19	32	41	15	39	8	13	44	38	20	11
EN	39	99	37	51	54	33	78	19	30	66	75	51	39
EU	20	39	84	48	30	33	32	17	34	43	43	37	30
FI	29	44	40	89	38	32	47	16	35	61	61	38	32
HE	43	54	33	46	90	21	69	12	28	59	58	31	24
HI	15	39	42	43	24	82	31	35	34	43	44	36	28
IT	52	69	34	55	59	25	93	14	32	67	74	34	27
JA	6	16	21	17	7	40	12	93	32	17	15	19	17
KO	9	21	23	27	17	18	20	15	86	26	24	31	13
RU	50	52	35	54	55	27	65	13	32	94	59	33	31
SV	37	71	40	55	48	31	70	17	32	63	89	35	33
TR	11	29	33	41	22	23	24	15	33	36	33	70	19
ZH	19	45	31	41	29	30	35	19	34	46	45	32	86

**BAP**  
Dozat and Manning (2017)

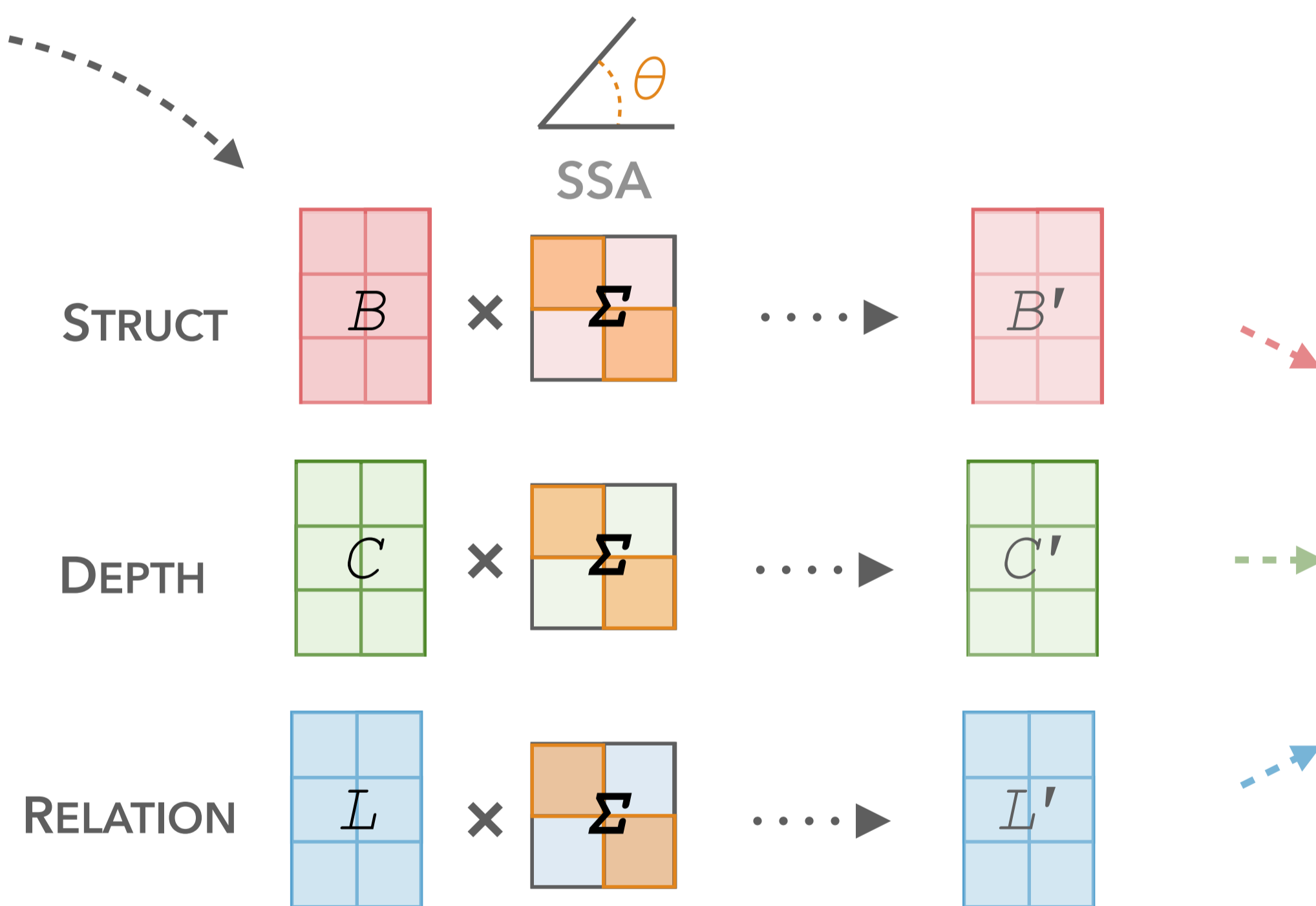
Extracting labeled + directed dependency trees w/ DEPProbe: **<3** matrices (**127k** parameters).



DEPProbe selects the best source **94%** of the time.

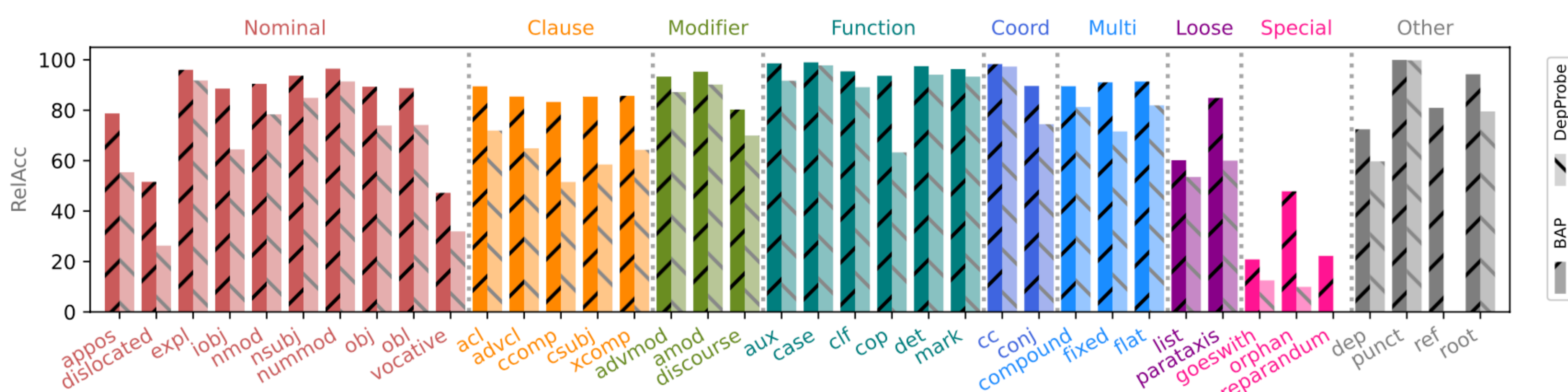
Each probe is **just one matrix**, so all relevant parameters across languages can be **compared at once!**

Which **type of dependency information** is relevant for cross-lingual transfer?



Probe Type	LAS		UAS	
	$\rho$	$\tau_w$	$\rho$	$\tau_w$
SSA-STRUCT	.68	.42	.60	.43
SSA-DEPTH	.62	.34	.53	.35
SSA-REL	<b>.73</b>	<b>.55</b>	.65	.53

SSA Correlation with BAP.  $\rho$  and  $\tau_w$  w.r.t. subspace angles of structural (Hewitt and Manning, 2019; STRUCT), depth (DEPTH) and relational probes (REL).



Relation Accuracy of BAP and DEPProbe compared for all 13 in-language targets, grouped according to the Universal Dependencies taxonomy (de Marneffe et al., 2014).

How much **relational information** do **tuned/untuned embeddings** contain?



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