# Welcome Note VarDial 2025

VarDial Workshop and Shared Task Organizers

Abu Dhabi, UAE, January 2025

# The VarDial Workshop Series

- VarDial 2014 at COLING in Dublin, Ireland
- LT4VarDial 2015 at RANLP in Hissar, Bulgaria
- VarDial 2016 at COLING in Osaka, Japan
- VarDial 2017 at EACL in Valencia, Spain
- VarDial 2018 at COLING in Santa Fe, United States
- VarDial 2019 at NAACL in Minneapolis, United States
- VarDial 2020 at COLING virtually in Barcelona, Spain
- VarDial 2021 at EACL virtually in Kiev, Ukraine
- VarDial 2022 at COLING in Gyeongju, Korea (and online)
- VarDial 2023 at EACL in Dubrovnik, Croatia (and online)
- VarDial 2024 at NAACL in Mexico City, Mexico (and online)
- VarDial 2025 at COLING in Abu Dhabi, UAE

#### Schedule

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9:00 - 9:30 — Opening and Findings of the Evaluation Campaign
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- 9:30 9:50 Shared Task Participants Poster Boosters
- **9:50 10:15** Oral presentation 1
- **10:15 10:30** Poster boosters I
- **10:30 11:00 —** Coffee break
- **11:00 12:00** Invited talk: Fajri Koto
- 12:00 12:30 Poster boosters II
- **12:30 14:00** Lunch break
- **14:00 15:00** Poster session
- **15:00 15:30** Oral presentation 2
- **15:30 16:00** Coffee break
- **16:00 17:15** Oral presentations 3-5
- **17:15 17:30** Closing remarks

# VarDial 2025 Organizers

#### Workshop organizers:

- Yves Scherrer
- Tommi Jauhiainen
- Marcos Zampieri
- Preslav Nakov
- Nikola Ljubešić
- Jörg Tiedemann

# Organizers of the NorSID shared task:

- Yves Scherrer
- Rob van der Goot
- Petter Mæhlum

# Overview of the VarDial 2025 Evaluation Campaign

The NorSID Shared Task:

Norwegian Slot, Intent and Dialect Identification

#### Slot and Intent Detection

Add VarDial to the calendar for tomorrow 9 AM

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**Intent:** AddCalendar

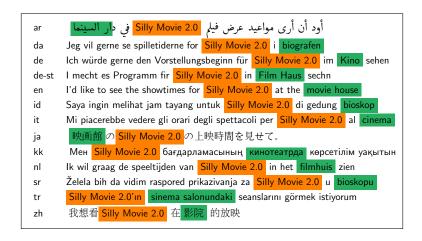
#### Slot and Intent Detection

Add VarDial to the calendar for tomorrow 9 AM

Intent: AddCalendar

**Slots:** Add [VarDial]<sub>Event</sub> to the calendar for [tomorrow 9 AM]<sub>Datetime</sub>

### xSID (2021)



Van der Goot, R., et al.: From Masked Language Modeling to Translation: Non-English Auxiliary Tasks Improve Zero-shot Spoken Language Understanding. Proceedings of NAACL 2021.

## SID4LR (Shared Task at VarDial 2023)

EN	Remind me to go to the dentist next Monday
IT	Ricordami di andare dal dentista lunedì prossimo
<b>NAP</b>	Ricuordam' 'e 'i addo dentista lunnerì prossimo
DE	Erinnere mich am nächsten Montag zum Zahnarzt zu gehen
GSW	Du mi dra erinnere nöchscht Mänti zum Proffumech zga
DE-ST	Erinner mi in negschtn Muntig zin Zohnorzt zu gian

Aepli, N., et al.: *Findings of the VarDial Evaluation Campaign 2023*. Proceedings of VarDial 2023.

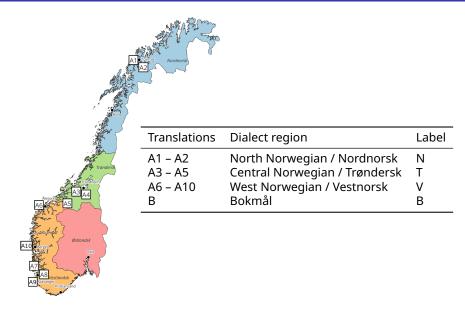
### NoMusic (VarDial 2024)

Extension of the xSID data to 11 Norwegian varieties: Standard Bokmål + 10 dialects (A1 – A10):

English	Set a reminder to go to the grocery store later
Bokmål	Sett på en påminnelse om å gå i butikken etterpå
A1	Minn mæ på at æ skal dra på butikken seinere.
A2	Sett enn påminnelse om å fære tel butikken seinar.
A3	Sett en alarm for å da te matbutikken seinere
A4	Sett en påminnelse om å gå te matbutikken seinar
A5	Sett en påminnelse for å gå t butikken seinar
A6	Sett en påminnelse om å stikke på butikken seinere.
A7	Sett på en påminnelse om å gå t butikken seinare
A8	Lag ein påminnelse om å gå på butikken seinere
A9	Sett ein påminnelse for å dra te matbutikken seinåre
A10	Sett på en påminnelse for å gå på butikken senere.

Mæhlum, P. & Scherrer, Y.: *NoMusic – The Norwegian Multi-Dialectal Slot and Intent Detection Corpus.* Proceedings of VarDial 2024.

## NoMusic (VarDial 2024)



### The NorSID Shared Task

#### Three subtasks:

- Slot identification
  - BIO span annotation task (40 slot types)
  - Metric: span F1 score
- Intent identification
  - Text classification task (18 intent labels)
  - Metric: accuracy
- Dialect identification
  - Text classification task (4 labels)
  - Metric: weighted F1 score on deduplicated data

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#### Four participating teams:

Team	Slots	Intents	Dialects
HiTZ	✓	✓	<b>√</b>
MaiNLP	$\checkmark$	$\checkmark$	
LTG	$\checkmark$	$\checkmark$	
CUFE		✓	✓

Dataset	Size	S	I	D
Manually annotated English xSID <b>training</b> set	43k	✓	✓	-
Machine-translated xSID training sets (12 languages, e.g. German, Dutch, Danish, Nor-	43k	✓	✓	-
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NorDial (dialectal tweets, not annotated)		-	-	<b>(√)</b>
NordicTweetStream (geotagged tweets, not necessarily dialectal)		-	-	(√)
Nordic Dialect Corpus + LIA corpus (dialecto- logical transcriptions, different genre)		-	-	(√)

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Concatenation of the 11 NoMusic <b>validation</b> sets (allowed for <b>training</b> )	3300	✓	✓	✓
Concatenation of the 11 NoMusic <b>test</b> sets	5500	<b>√</b>	$\checkmark$	<b>√</b>

# Slots and Intents - Participants and Approaches

• Baseline: Multi-task mBERT fine-tuned on English xSID training data

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- Multi-task > single-task models (HiTZ)
- Norwegian/Scandinavian  $\approx$  multilingual base models (HiTZ, MaiNLP, LTG, CUFE)
- Norwegian > English training data for intents (HiTZ, MaiNLP, LTG)
- English > Norwegian training data for slots (HiTZ, MaiNLP, LTG)

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- Improved label projection and translation of the Norwegian training data (LTG)
- Noise injection to simulate spelling and dialectal variation dialectal variation
- Training on auxiliary tasks (NER, POS, Dep, DID) \( \Pi \) (MaiNLP)
- Combining layers of models fine-tuned on different datasets (MaiNLP)

### Slots and Intents - Results

#### **Intents** (accuracy %):

Submission	В	N	Т	V	all
LTG 3	98.00	97.20	98.27	98.20	98.02
LTG 1	98.20	97.20	98.33	97.84	97.89
LTG 2	98.20	97.30	98.13	97.84	97.85
HiTZ 2	98.20	97.10	97.60	97.88	97.69
MaiNLP 3	97.80	96.90	98.00	97.68	97.64
MaiNLP 2	97.60	96.20	97.67	97.16	97.16
HiTZ 3	97.80	95.40	97.80	97.24	97.11
HiTZ 1	97.40	95.40	96.93	96.04	96.29
CUFE 1	96.40	93.30	95.80	93.56	94.38
MaiNLP 1	92.80	92.60	93.40	94.00	93.47
Baseline	86.40	82.60	83.33	84.80	84.15
LTG 4*	97.80	96.70	97.73	97.20	97.31

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- Similar errors across teams
- Subtask is close to being solved

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MaiNLP 1	92.80	92.60	93.40	94.00	93.47
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LTG 4*	97.80	96.70	97.73	97.20	97.31

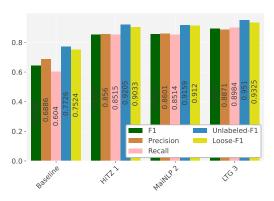
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#### Slots (span F1-score %):

Submission	В	Ν	Т	٧	all
LTG 3	90.94	87.19	89.69	89.49	89.27
LTG 2	89.92	87.89	89.27	89.62	89.25
MaiNLP 2	90.11	79.66	85.18	87.17	85.57
HiTZ 1	91.09	79.00	85.48	86.61	85.37
MaiNLP 1	85.60	82.66	82.99	84.11	83.68
MaiNLP 3	84.37	79.25	81.68	84.01	82.57
LTG 1	84.74	80.09	80.96	83.30	82.22
HiTZ 3	71.15	60.98	66.22	68.18	66.64
Baseline	71.49	60.68	63.23	65.05	64.36
HiTZ 2	56.74	51.94	56.69	56.25	55.66
LTG 4*	91.84	87.56	89.00	89.82	89.38

- Northern dialects seem most difficult
- LTG 4\* includes Norwegian MASSIVE training dataset (not allowed by ST guidelines)

#### Slots - Results



- Precision > recall (except LTG)
- Unlabeled F1 > labeled F1 (difficulties finding the correct label)
- Loose F1 > strict F1 (difficulties finding the exact span boundaries)

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- Encoder models with fine-tuning > decoder models with few-shot prompting or supervised fine-tuning (HiTZ)

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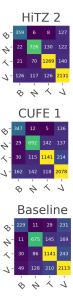
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- Include additional silver-labeled datasets ♥ (HiTZ)

#### Dialects - Results

#### Weighted-average F1-score %:

Submission	В	N	Т	V	all
HiTZ 2	75.40	78.44	85.95	87.45	84.17
HiTZ 3	74.91	77.50	84.29	87.08	83.32
HiTZ 1	74.10	75.72	83.97	86.61	82.71
CUFE 1	68.93	73.38	80.26	84.14	79.64
Baseline	57.38	73.46	77.76	82.59	77.42

- Systems struggle most with identifying Bokmål and Nordnorsk, the two varieties with least data (1 and 2 translators, respectively)
- Confusions between the Western (V) dialects and Bokmål are most common
- Also significant confusion between the non-adjacent dialect areas N and V



## **Takeaways**

- Intent identification is mostly solved, whereas slot and dialect identification show room for improvement:
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  - Is Norwegian dialect writing closer to standard?
  - Is there better cross-lingual transfer from English?

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  - Is Norwegian dialect writing closer to standard?
  - Is there better cross-lingual transfer from English?
- What kind of variation does the Norwegian data actually contain?
  - Is individual speaker variation (punctuation, word choices, translationese, ...) more salient than dialectal variation?

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#### **Announcements**

Tutorial 6: Connecting Ideas in Lower-Resource Scenarios: NLP for National Varieties, Creoles, and Other Low-Resource Scenarios

**Organizers:** Aditya Joshi, Diptesh Kanojia, Heather Lent, Hour Kaing and Haiyue Song

**Time:** Tomorrow Monday, 09:00 - 17:30

**Location:** Conference Hall B (C)

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I have a 3-year postdoc opening at the University of Oslo!

Deadline: 6 April

