

Deep Learning Bias Wind Prediction at ZRH airport

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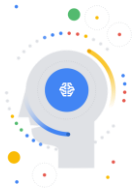
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Bise POC Success Criteria



First ML Model

First ML model for Bise created



Outperform existing Approaches

ML model to beat heuristics in precision, recall, f1 score



Extend Model Timeframe

Extend model to 6h horizon + beat heuristics



Make it easy with Technology

Evaluate new VertexAI time series technology

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- What is Bise?
- Different Forecasting Approaches

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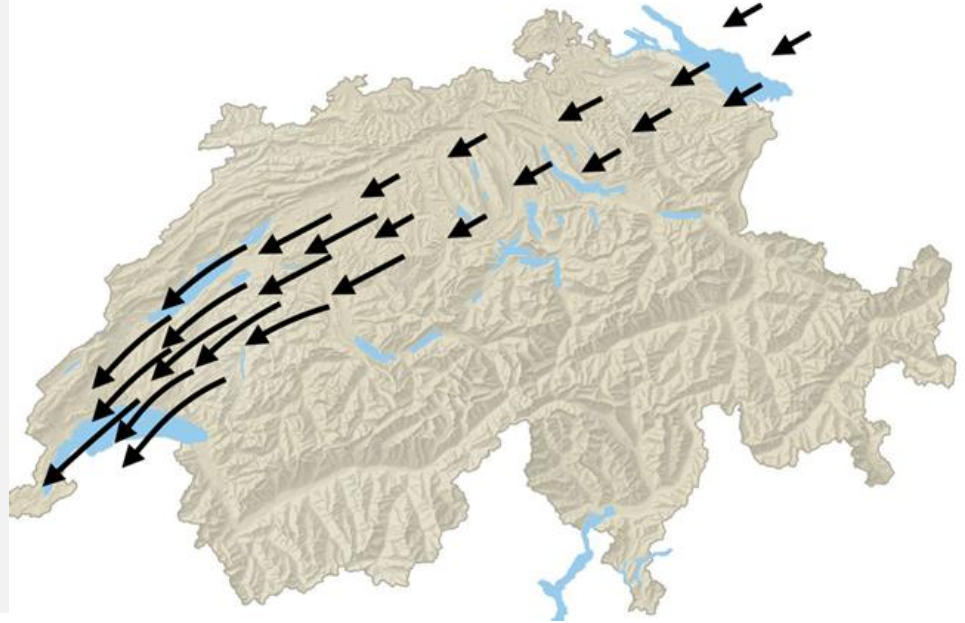
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What is BISE?

- Cold, mostly dry wind from the north-east to south-west
 - Canalization through the Alps and the Jura Mountains towards the west
 - Summer: good weather
 - Winter: high fog, strong haze
 - Strongest at Lake Geneva: Wind speed: avg. 60 km/h, Gust peak: >100 km/h
- ➔ 30% decrease of the ZRH airport capacity for landings & take-offs



What does this mean for the ZRH airport?



30% decrease in the ZRH airport capacity
 Potential delays - on departure and arrival



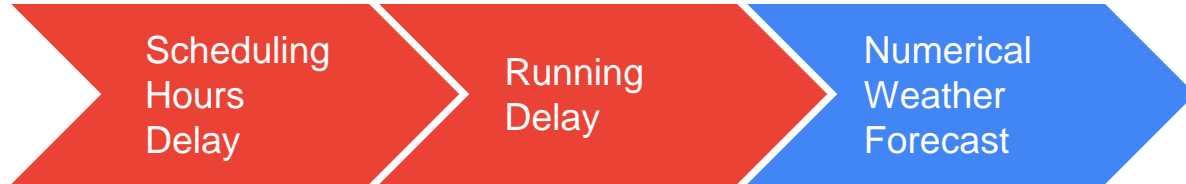
Nordanflugkonzept	Ostanflugkonzept	Südanflugkonzept
Landungen von Norden, Starts Richtung Westen und Süden, bei Bise Richtung Osten	Landungen von Osten, Starts Richtung Norden	Landungen von Süden, Starts Richtung Norden und Westen
Landungen: Piste 14 und 16 Starts: Piste 28 und 16 Piste 10 (Bise)	Landungen: Piste 28 Starts: Piste 32 und 34	Landungen: Piste 34 Starts: Piste 32, 34 teilweise Piste 28
Generelle Anwendung: <ul style="list-style-type: none"> • 07:00 - 21:00 Uhr Mo - Fr • 09:00 - 20:00 Uhr Sa und So, baden-württembergische Feiertage <p>→ bei Biswind auch morgens und abends</p>	Generelle Anwendung: <ul style="list-style-type: none"> • 21:00 - 23:30 Uhr Mo - Fr • 20:00 - 23:30 Uhr Sa und So, baden-württembergische Feiertage <p>→ bei Westwind auch tagsüber</p>	Generelle Anwendung: <ul style="list-style-type: none"> • 06:00 - 07:00 Uhr Mo - Fr • 06:00 - 09:00 Uhr Sa und So, baden-württembergische Feiertage <p>→ abends, wenn Ostanflüge nicht möglich sind (Bise, schlechte Sicht etc.) oder tagsüber wenn weder Nord- noch Ostanflüge möglich sind</p>

Forecasting Approaches

1. Numerical weather forecasting

- Global optimization
- Run every couple of hours + time to compute
- Lead time in days

1



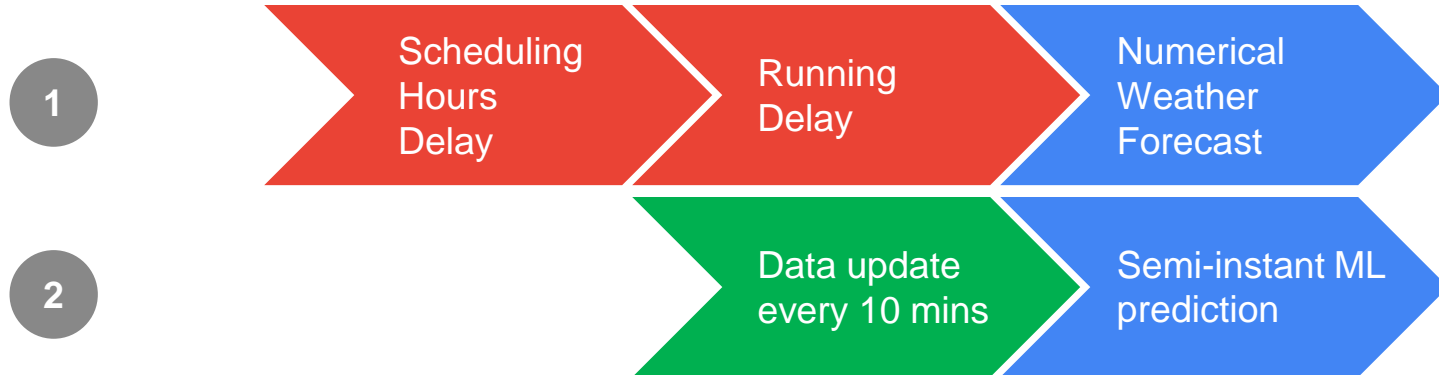
Forecasting Approaches

1. Numerical weather forecasting

- Global optimization
- Run every couple of hours + time to compute
- Lead time in days

2. ML time series prediction

- Focus on local optimization (ZRH airport)
- Semi-instant forecast (only initial training of the model takes hours)
- Lead time in hours



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- Data
- Technology

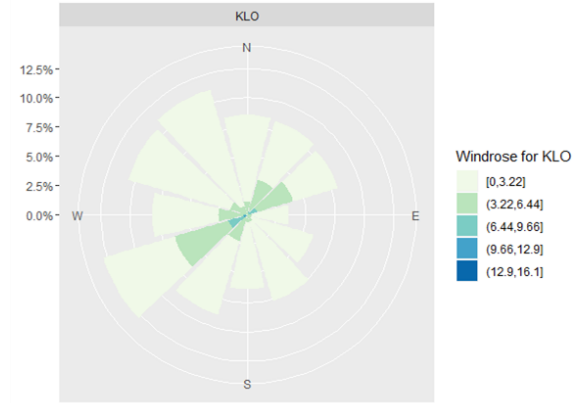
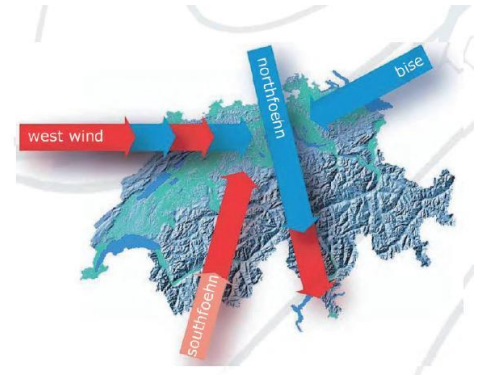
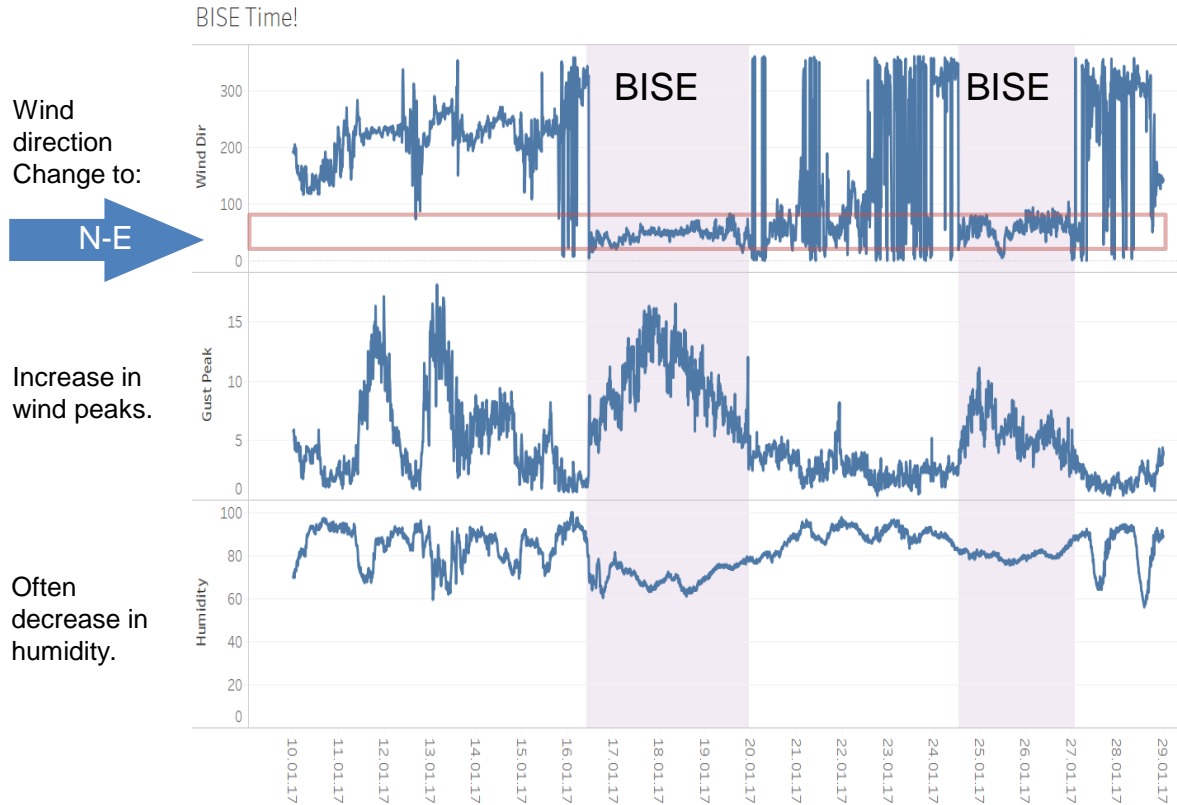
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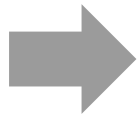
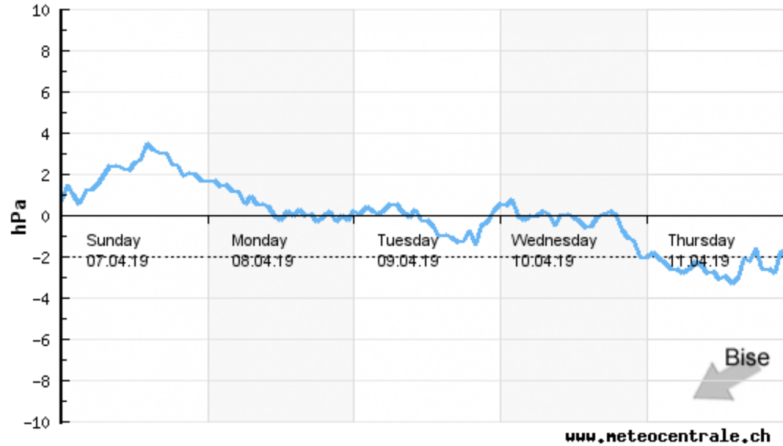
Next Steps

Weather Data



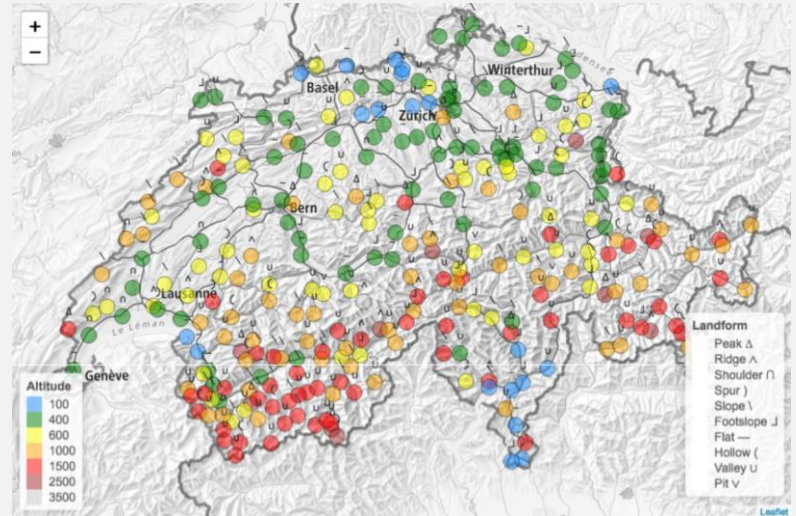
Weather Data

Current prediction based on pressure difference between two points (from MeteoSwiss simulation)



Project idea: Create a machine learning system to predict Bise wind in ZRH

Idea for the future implementation:
Utilize information from actual measurements to create more granular prediction with a model based on historical data.



Inputs

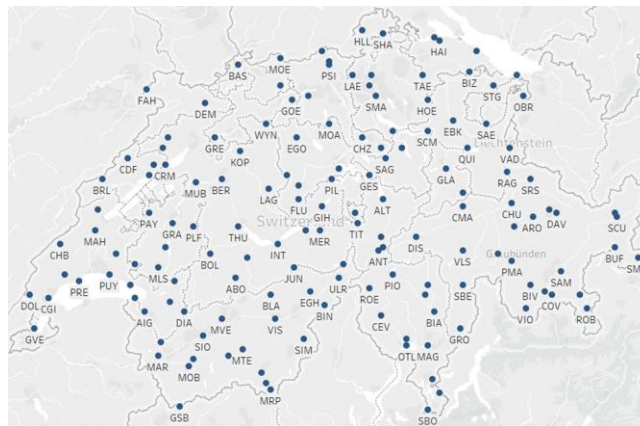
1 Meteorological data

- Wind:
 - Direction
 - Strength
 - Gusts
- Pressure (3 different fields)
- Temperature
- Humidity

5 years of data, 10 min resolution

2 Station location data (~150)

- Name
- Position (2 / 4 fields)
- Altitude
- Available measured parameters



Statistics

- High impact, but a rare event – imbalanced sample.
- Depending on the selection criteria for the Bise, it can be in range of 0.6% to 3.2%.

Bise according to >10knots headwind on runway 10

windSpeedDir100gt5.15

Null	129
False	261,194
True	1,478

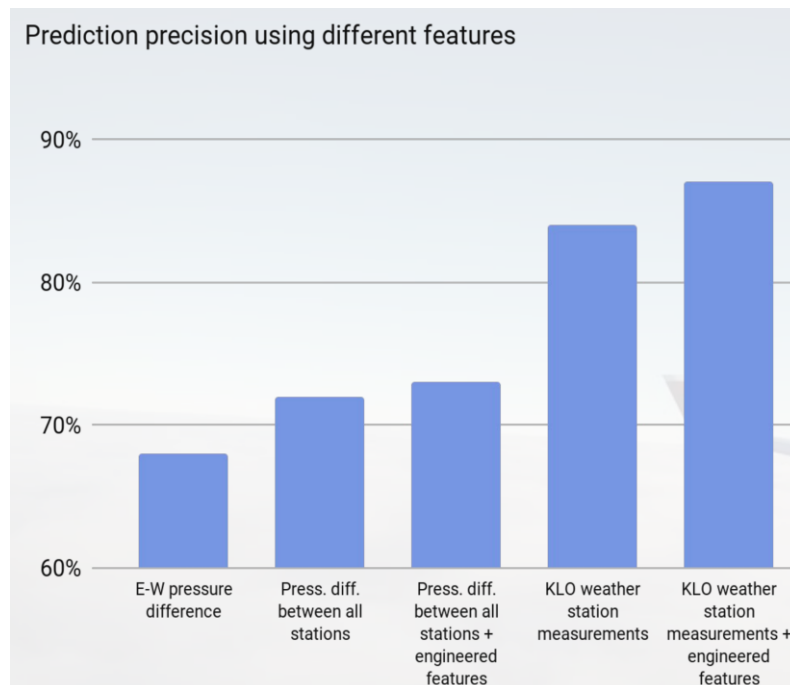
Bise according to >10knots headwind gust on runway 10

windGustDir100gt5.15

Null	129
False	254,284
True	8,388

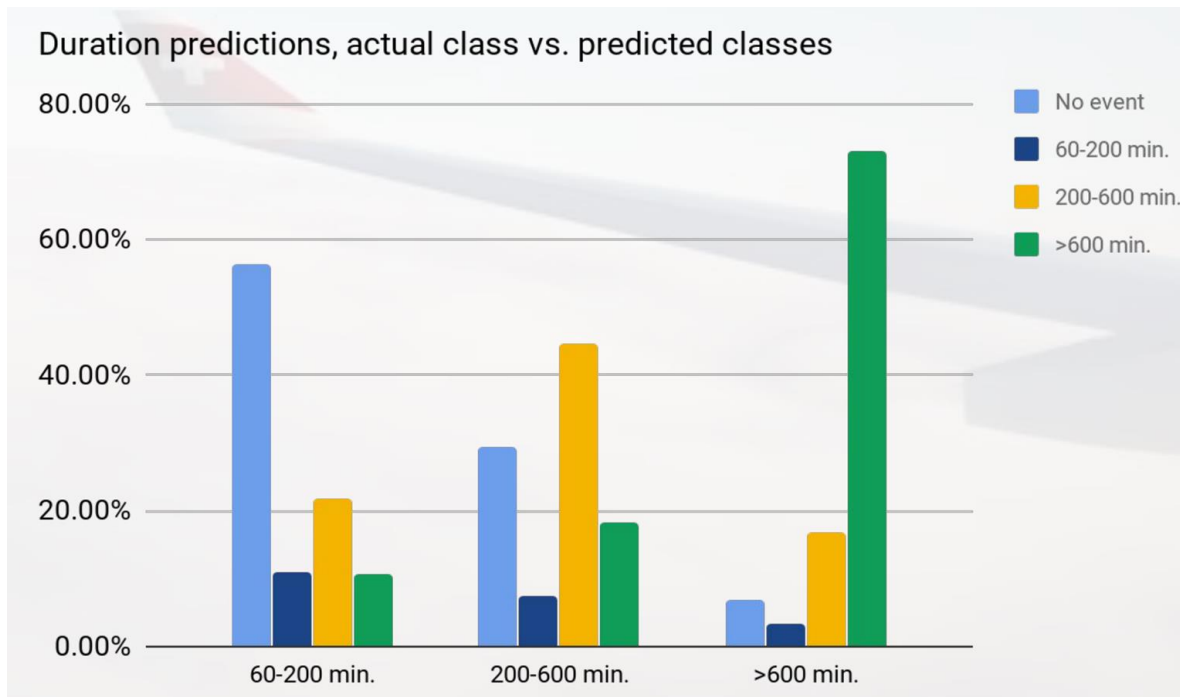
What can we predict?

Next time steps – wind / wind gust values along runway or binary classification



What can we predict?

Event duration



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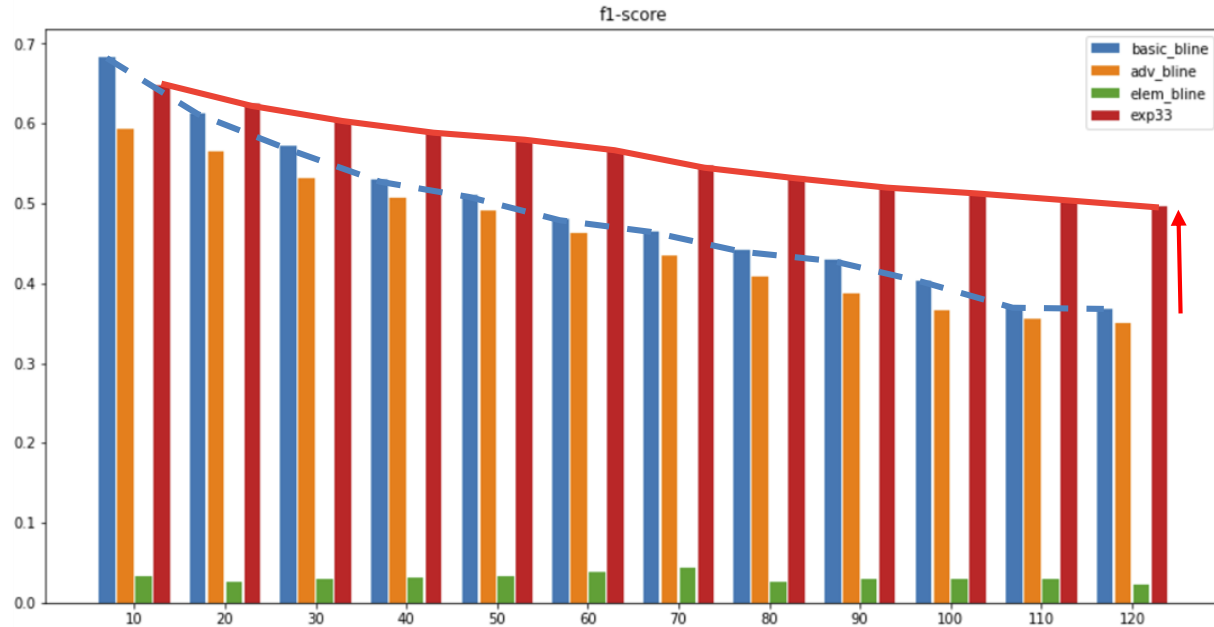
Results

- 2 hour window
- 6 hour window

5

Next Steps

Results - F1 Score 2 hrs

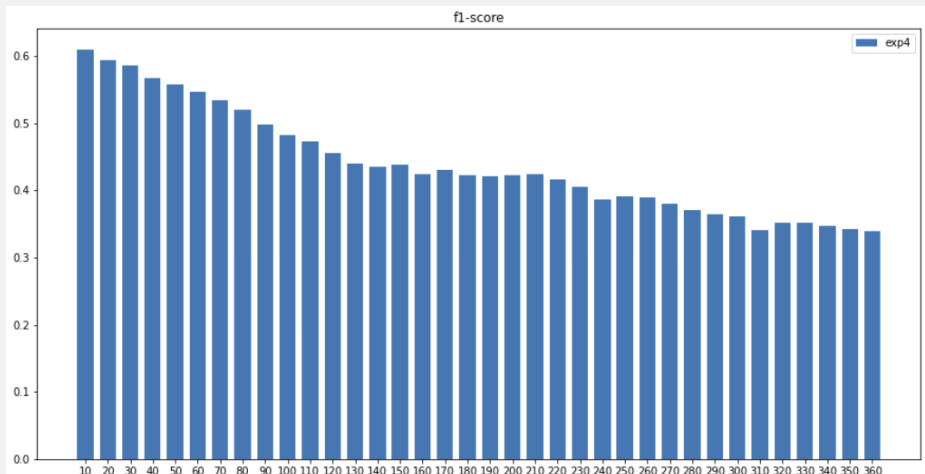


Precision answers “out of all the times gives rise how often is it actually a rise?”.
Since Precision/recall are inversely correlated trade-offs: F1 score is the harmonic mean of precision and recall

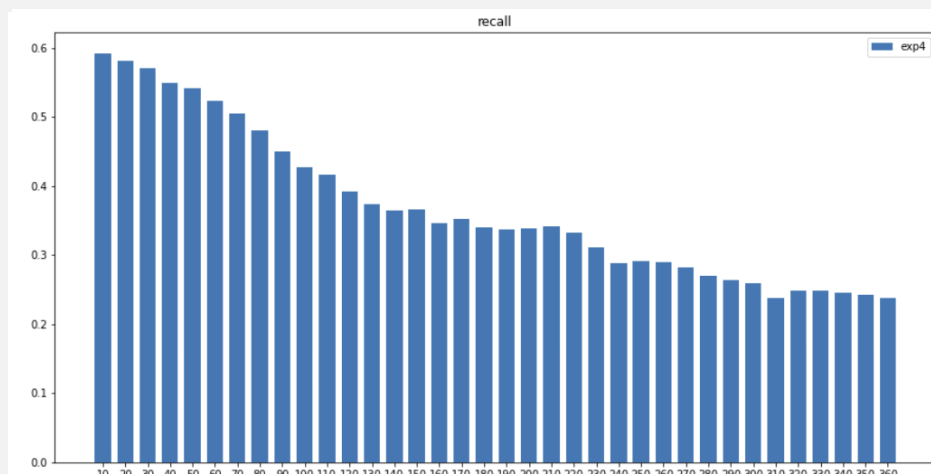
$$\text{Precision} = \frac{tp}{tp + fp}$$
$$\text{Recall} = \frac{tp}{tp + fn}$$

- Predicting Rise with ML is possible
- Good overall precision and recall
- The longer the time scale, the bigger advantage of DL models over baseline

Results - 6 hours



F1-Score

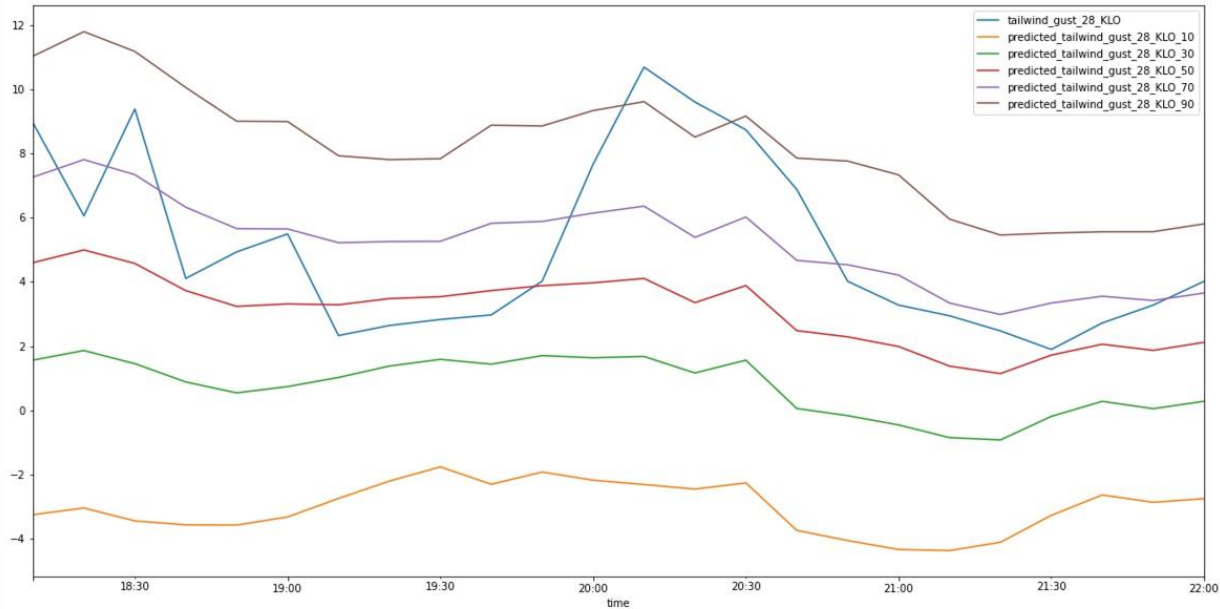


Recall



- Extension beyond 2 hours - works, strong performance at 3h
- The longer the time scale, the bigger advantage of DL models over baseline

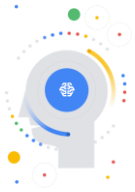
Model Learning Wind Movements - 2 hrs in advance



Behind the Scenes:

- Model uses Wind prediction to determine Bise
- Wind direction predicted 90%+ of times two hrs ahead

POC Success Criteria



First ML Model

First ML model for Bise created



Outperform existing Approaches

ML model to beat heuristics in precision, recall, f1 score



Extend Model Timeframe

Extend model to 6h horizon + beat heuristics



New Technology

Evaluate new VertexAI time series technology



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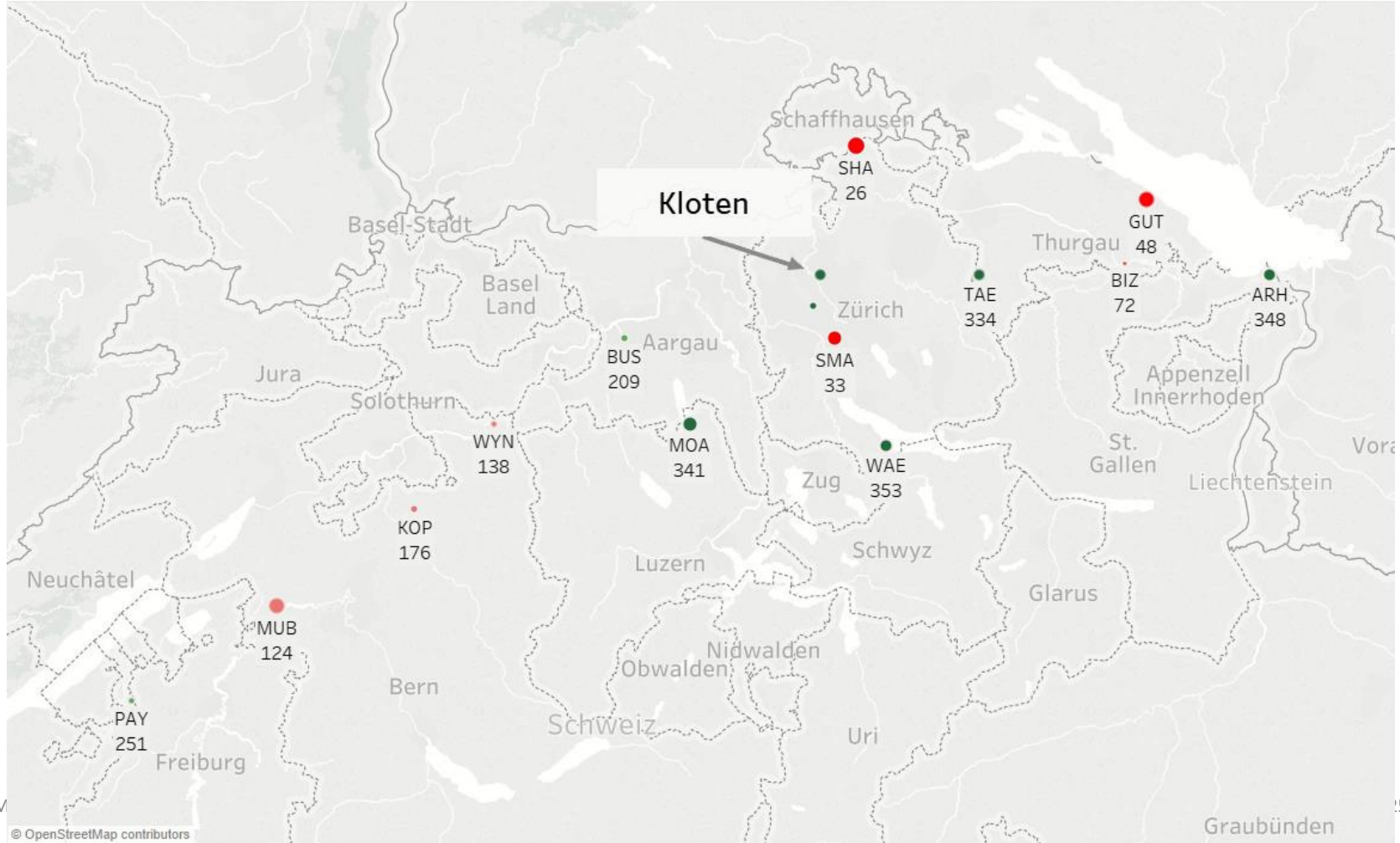


- 1 Productionalization [from **POC** to **MVP**]
- 2 Possible implementation as a **dashboard** for operators
- 3 Usability / **user acceptance** needs to be assessed
- 4 Approach scalable to other **time series** problems



Thank you very much
for your attention

Backup slides



Data – legend files

Parameter

Einheit	Beschreibung
dkl010z0 °	Windrichtung; Zehnminutenmittel
fkl010z0 m/s	Windgeschwindigkeit skalar; Zehnminutenmittel
fkl010z1 m/s	Böenspitze (Sekundenböe); Maximum
pp0qffs0 hPa	Luftdruck reduziert auf Meeresniveau (QFF); Momentanwert
pp0qnhs0 hPa	Luftdruck reduziert auf Meeresniveau mit Standardatmosphäre (QNH); Momentanwert
prestas0 hPa	Luftdruck auf Stationshöhe (QFE); Momentanwert
tre200s0 °C	Lufttemperatur 2 m über Boden; Momentanwert
ure200s0 %	Relative Luftfeuchtigkeit 2 m über Boden; Momentanwert

Data – legend files

Stationen

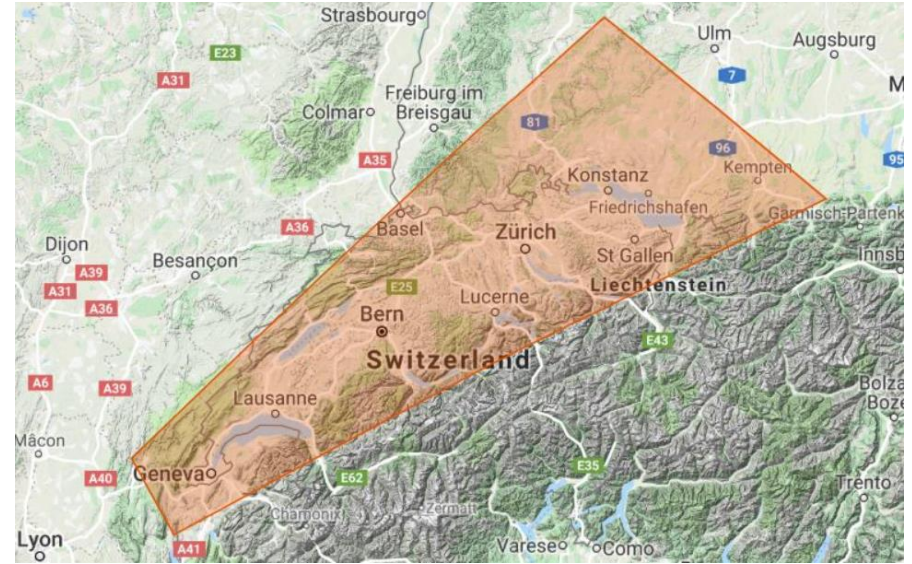
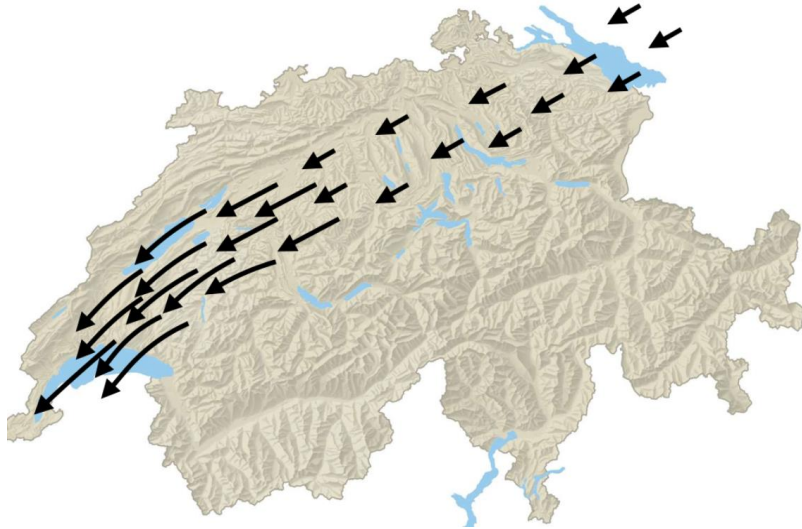
stn	Name	Parameter	Datenquelle	Länge/Breite	Koordinaten [km]	Höhe ü. M. [m]
VIO	Vicosoprano	dkl010z0	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VIO	Vicosoprano	fkl010z0	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VIO	Vicosoprano	fkl010z1	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VIO	Vicosoprano	pp0qnhs0	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VIO	Vicosoprano	presta0	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VIO	Vicosoprano	tre200s0	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VIO	Vicosoprano	ure200s0	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VIO	Vicosoprano	pp0qffs0	MeteoSchweiz	9°38'/46°21'	768485/135866	1089
VLS	Vals	dkl010z0	MeteoSchweiz	9°11'/46°38'	734016/165551	1242
VLS	Vals	fkl010z0	MeteoSchweiz	9°11'/46°38'	734016/165551	1242
VLS	Vals	fkl010z1	MeteoSchweiz	9°11'/46°38'	734016/165551	1242
VLS	Vals	pp0qnhs0	unknown	9°11'/46°38'	734016/165551	1242
VLS	Vals	presta0	unknown	9°11'/46°38'	734016/165551	1242
VLS	Vals	tre200s0	MeteoSchweiz	9°11'/46°38'	734016/165551	1242
VLS	Vals	ure200s0	MeteoSchweiz	9°11'/46°38'	734016/165551	1242
VLS	Vals	pp0qffs0	unknown	9°11'/46°38'	734016/165551	1242

Data

Stn	Parameter / Datenquelle																			
	dkl010z0			fkI010z0		fkI010z1		pp0qffs0		pp0qnhs0			prestas0			tre200s0		ure200s0		
	MeteoSchiweiz			MeteoSchiweiz		MeteoSchiweiz		MeteoSchiweiz		Schweizerische Rettungsflugwach		MeteoSchiweiz			MeteoSchiweiz		MeteoSchiweiz			
									unknown						unknown		unknown		unknown	
ABO	■	1		■	1	■	1		■	1	■	1				■	1	■	1	
AEG	■	1		■	1	■	1		■	1			■	1			■	1	■	1
AIG	■	1		■	1	■	1	■	1		■	1			■	1	■	1	■	1
ALT	■	1		■	1	■	1	■	1		■	1						■	1	
AND	■	1		■	1	■	1		■	1	■	1						■	1	
ANT	■	1		■	1	■	1		■	1	■	1						■	1	
ARH	■	1		■	1	■	1	■	1		■	1						■	1	
ARO	■	1		■	1	■	1		■	1	■	1						■	1	
ATT	■	1		■	1	■	1		■	1	■	1						■	1	
BAS	■	1		■	1	■	1	■	1		■	1				■	1	■	1	
BEH	■	1		■	1	■	1		■	1	■	1				■	1	■	1	
BER	■	1		■	1	■	1	■	1		■	1				■	1	■	1	
BEZ	■	1		■	1	■	1	■	1		■	1				■	1	■	1	
BIA	■	1		■	1	■	1	■	1		■	1				■	1	■	1	
BIE	■	1		■	1	■	1		■	1	■	1				■	1	■	1	
BIN	■	1		■	1	■	1		■	1			■	1			■	1		
BIV	■	1		■	1	■	1		■	1			■	1			■	1		

Datenquelle: MeteoSchiweiz
 Parameter: tre200s0
 Stn: AIG
 Count of Stn: 1

Project proposal



Es wird diskutiert wie man von der Meteoprognose zu einem Entscheid kommt, der dann das DCB Verfahren auslöst. Aus Sicht von Eugen Müller stellt heute das TAF, welches auf einem deterministischen Modell basiert, das am besten geeignete Produkt dar. In Zukunft könnte man auch Prognosen auf Basis von Ensemble Modellen verwenden, die dann auch Eintretenswahrscheinlichkeiten berechnen können. Das TAF wird alle 3h aktualisiert und umfasst einen Prognosezeitraum von 30h. Zum Zeitpunkt der Nachmittagstelefonkonferenz hätte man somit eine Prognose für die Betriebszeit des nächsten Tages. Anhand der TAF Daten würde man dann die Wahrscheinlichkeit, dass das Runway Konzept Bise zur Anwendung kommt ableiten und in Kategorien einteilen.

Beispiel:

Time	0600	0700	0800	0900	1000	1100	1200	1300	1400	1500
Bise Risk										
Bise Key	NIL		Low		Medium		High			Very High

Mit «Bise Risk» ist in dem Zusammenhang das Runway Konzept gemeint.

Die Einteilung in die Gruppen würde auf Basis der TAF Windinformation und der daraus abgeleiteten Rückenwindkomponente für RWY 28 erfolgen. Am Meeting wurde sich initial auf folgende Einteilung verständigt.

- Very High: TWC RWY 28 \geq 12kt
- High: TWC RWY 28 \geq 8kt
- Medium: TWC RWY 28 \geq 4kt
- Low: TWC RWY 28 \geq 0kt
- NIL: TWC RWY 28 $<$ 0kt

Der DCB Bisenprozess würde dann ausgelöst werden, wenn zum Beispiel das Risiko für Bisenkonzept High oder

Speed ⌵

10

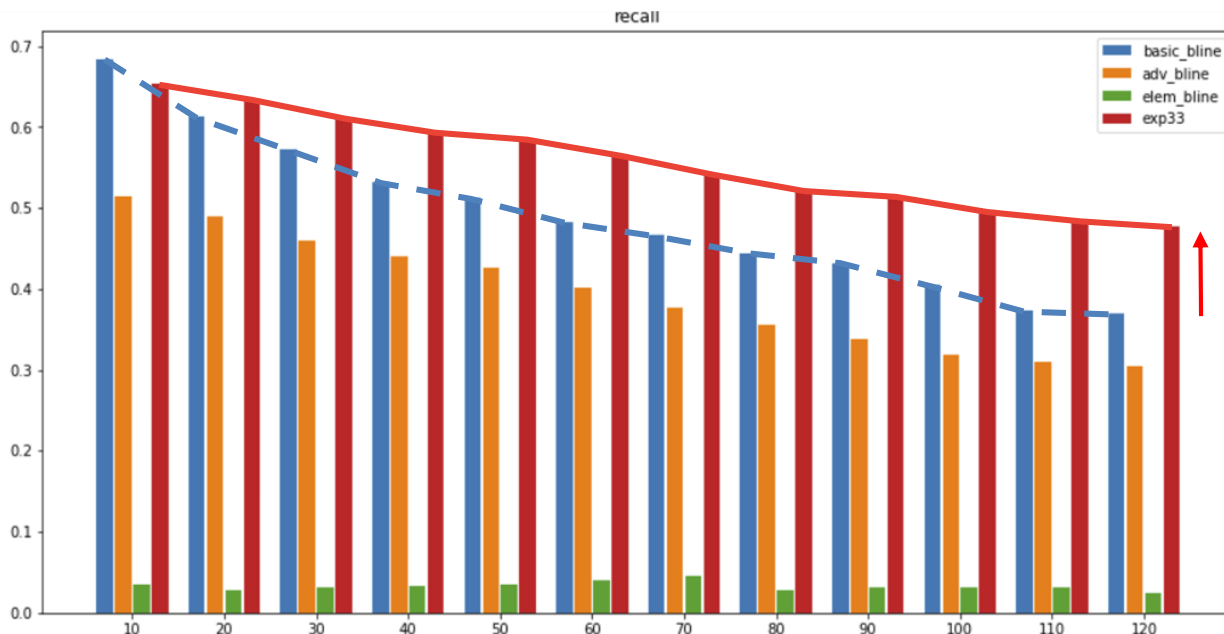
Knot ⌵

=

5.14444

Metre per second ⌵

Results - Recall 2 hrs



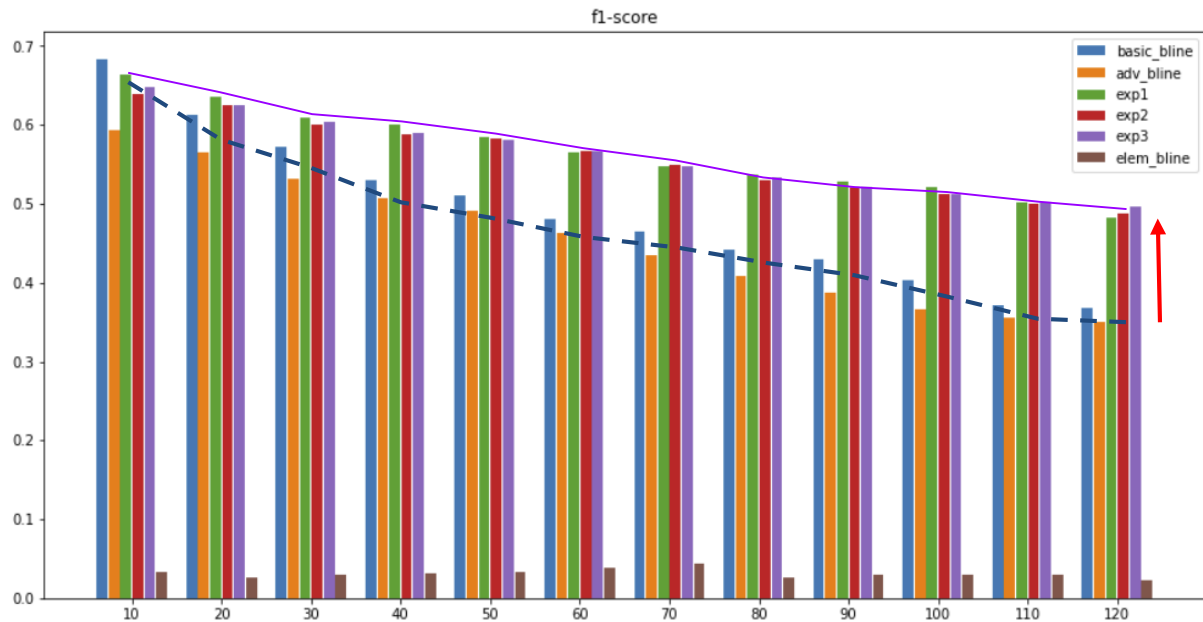
Recall is the number of correct results divided by the number of results that should have been returned.
“What %age of the actual bise periods did were caught?”

$$\text{Precision} = \frac{tp}{tp + fp}$$
$$\text{Recall} = \frac{tp}{tp + fn}$$



- Predicting Bise with ML is possible
- Good overall precision and recall (+10-11 points over heuristics on average- better for longer forecasts)
- The longer the time scale, the bigger advantage of DL models over baseline

Results - F1 Score



Precision answers “ out of all the times gives bise how often is it actually a bise?”.
Since Precision/recall are inversely correlated trade-offs: F1 score is the harmonic mean of precision and recall

$$\text{Precision} = \frac{tp}{tp + fp}$$

$$\text{Recall} = \frac{tp}{tp + fn}$$