A Speech-to-Text Interface for MammoClass

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Outline

- MammoClass
- Development of STT Interface for MammoClass
- Web Speech API applied to MammoClass
- Conclusions and Future Work

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MammoClass

Classification of a mammogram based in a set of mammography findings

MammoClass Classification of a mammogram based in a reduced set of mammography findings To obtain a prediction in terms of malignancy for a certain mass is only necessary to provide the values of the findings, annotated through the Breast Imaging Reporting and Data System (BIRADS), in the form bellow. It is also possible to get a prediction of the attribute mass density in case this feature is not known. The output will indicate the probability of a certain mass being benign or malignant. In the latter case it is suggested that the patient should perform a biopsy. The probabilities are computed using machine learning models built as described in: . P.Ferreira, N. A. Fonseca, I. Dutra, R. Woods, and E. Burnside, Predicting Malignancy from Mammography Findings and Surgical Biopsies, submitted. **Enter Data** Patient's age Mass size Breast Composition | Select a value Mass shape | Select a value | V Mass clockface location | Select a value | v Mass margins (1) Select a value V Mass margins (2) Select a value V

MammoClass

Classification of a mammogram based in a set of mammography findings

Enter Data

Patient's age 22
Mass size 10
Breast Composition Almost entirely fat ▼
Mass shape Lobular ▼
Mass clockface location 2.0 ▼
Mass margins (1) Circumscribed ▼
Mass margins (2)
Mass margins worst Mass Margins (1) ▼
Mass density Iso/Low ▼
Side Left ▼
Quadrant Lower Inner ▼
Depth Anterior ▼
Predist Reset
Result
Predicted mass density: iso (98.4%)
Prediction: mass benign with a probability of 99.9%.

MammoClass – How is it done?

• To **obtain a prediction** in terms of malignancy for a certain mass is only necessary to provide the values of the findings through forms.

 The output will indicate the probability of a certain mass being benign or malignant. In the latter case it is suggested that the patient should perform a biopsy. The **probabilities** are **computed using machine learning models** built as described in:

Ferreira, P., Fonseca, N.A., Dutra, I., Woods, R., Burnside, E.:

Predicting Malignancy from Mammography Findings and Image-Guided Core Biopsies.

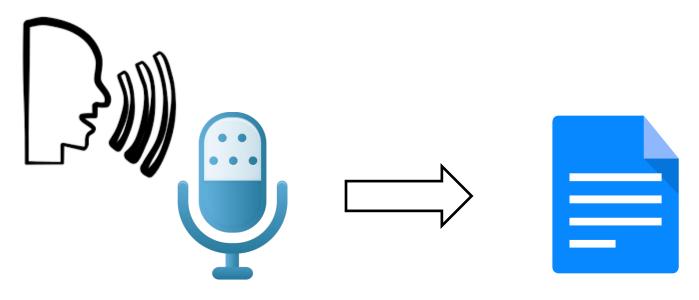
In: Int. Journal of Data Mining and Bioinformatics, 2015.

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What is Speech-to-Text?

Type of software that **takes audio content** and **transcribes it** into written **words** in a word processor or other display destination



What is Speech-to-Text?

> Advantages:

 Valuable to anyone who needs to generate a lot of written content without a lot of manual typing

annwyd asleep avolding bably beach bed beer best big blise birthday biscuits biolo book buy bush buyling ac saice art hecknigh, dreese chicken chocalete christmas coffee Cold colman considering contemplating conversation copy and considering contemplating contemplat

 Useful for people with disabilities that make it difficult for them to use a keyboard

Speech-to-Text Interface for MammoClass



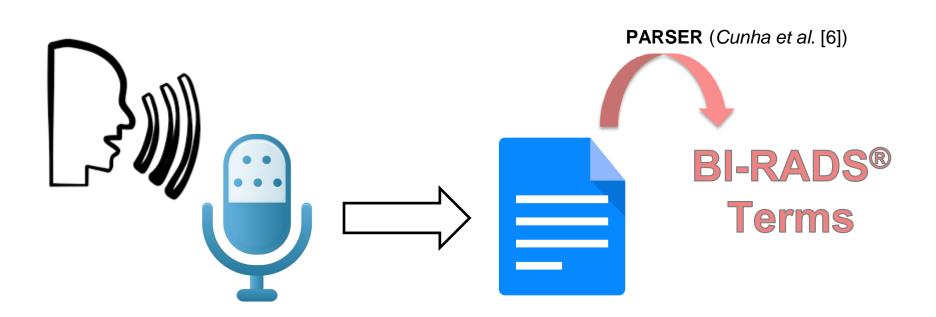
Speech-to-Text Interface for MammoClass

 Some works in the literature do not favor the use of speech recognition technology in the area of radiology and report a high error rate on the resulting recognized texts [1], [2], [3], [4], [5]

 These works focus on the text itself rather than on relevant words that could be extracted from the text to build structered data for posterior automatic studies

Speech-to-Text for MammoClass

What's the innovation?



Looking for the suitable tool...

What tool to choose?

- Free
- Support the **Portuguese** language





Tested Tools

- Free Voice to Text
 - (+) Can be used to send emails, create documents by just dictating
 - (-) Does not support Portuguese
- <u>Talking Desktop</u>
 - (+) Makes **text recognition**. Has functions to **recognize dictated text** about **weather** conditions to emit meteorological warnings
 - (-) Limited controls and slow reaction time
 - (-) Does not support Portuguese
- Bragon Naturally Speaking Home (Premium)
 - (+) Has a very functional and user-friendly interface / Has specific vocabulary (eg. medical)
 - (-) Does not support Portuguese

Tested Tools (cont.)



Freesr Speech Recognition



- (+) Has the ability to recognize multiple dictated texts
- (-) Does not support Portuguese
- Simon **
 - (+) Open Source software available for Windows and Linux
 - (-) Does not support Portuguese



Web Speech API



- (+) Google API that allows the programmer to obtain a translation of voice to text
- (+) Supports Portuguese!!



<u>VoiceNote</u>



- (+) Extension for Google Chrome
- (+) Supports Portuguese!!

Tested Tools - Table of Comparison

Software	Free	Price	Languages	Platform	
Free Voice to Text	Yes	0\$	Eng, Spa, Fre, Jpn	Windows	
Talking Desktop	#	47\$	Eng, Spa, Fre, Ger	Windows	
Dragon Naturally Speaking Home	(I)	199\$	Eng	Windows	
Speech Recognition	Trial	NA	Eng	Windows	
*	Yes	0\$	Eng	Linux, Windows	
	Yes	0\$	+	All	
	Yes	0\$	+	All	

Candidate Tools



VS



Web Speech API

VoiceNote

Candidate Tools

• **Report:** No actual estudo, observamos padrão mamográfico de densidades fibroglandulares dispersas, pela pequena quantidade de parênquima mamário.

no **atual** estudo observamos **pedro** mamográfico de densidades **fibroglandular** dispersas pela pequena quantidade de parênquima mamário.



No actual estudo observamos **pedro monográfico** de densidades **fibroglandular** dispersas pela pequena quantidade de parênquima mamário.



Candidate Tools

• **Report:** No actual estudo, observamos padrão mamográfico de densidades fibroglandulares dispersas, pela pequena quantidade de parênquima mamário.

<u>no atual estudo observamos pedro mamográfico de densidades</u> fibroglandular <u>dispersas pela pequena quantidade de parênquima mamário</u>.



No actual estudo observamos pedro monográfico de densidades fibroglandular dispersas pela pequena quantidade de parênquima mamário.



Verdict

- Results very similar
- We believe VoiceNote was built using the Web Speech API

Chosen Tool:



Allows greater freedom since it is an API eg. Can be integrated easy way in any element of a Web page



Web Speech API tested with



TABLE II BI-RADS TERMS AND THEIR TRANSLATION TO PORTUGUESE

		round	arredondada, redonda
mass shape		oval	oval, ovóide, alongada, ovalar
mass shape		lobular	lobular, Polilobular
		irregular	irregular
	7	circumscribed	circunscrita, bem definida, bem delimitada, regular
		microlobular	microlobular
mass margins		obscure	obscura, obscurecida
		indistinct	indistinta, imprecisa, indefinida, mal definida
		spiculated	espiculada
		high	alta, elevada densidade
mass density	\	equal	igual, isodensa, homogênea
	\	low	baixa, tênue
		predominantly fatty	contém gordura
breast density		scattered fibroglandular	fibroglandular
		heterogeneously dense	heterogênea
		extremely dense	densa, muito densa, densidade alta da mama

Testing 86 Individual BI-RADS Terms

- We classify each returned result as:
- **(C) Correct** If the original term and the recognized term are **exactly alike**
- (AC) Almost Correct If the original term and the term returned by Web Speech API are almost identical
- (I) Incorrect If the original term and the term returned by the API are completely different

Testing **86** Individual **BI-RADS Terms**

ambos os lados	ambos os lados	С
distorção arquitectural	distorção arquitectural	C
desorganização arquitectural	desorganização arquitectural	С
ducto dilatado	ducto dilatado	C
carcinoma ductal	carcinoma ductal	С
Gânglio linfático intramamário	Gânglio linfático intramamário	C
Gânglio mamário	Gânglio mamário	С
densidade assimétrica	densidade assimétrica	C
assimetria mamária	assimetria mamária	C
densidade focal	densidade focal	C
foco nodular	foco nodular	C
foco assimétrico	foco assimétrico	C
retracção cutânea	respiração cutânea	I
repuxamento da pele	repuxamento da pele	С
retracção do mamilo	retracção do mamilo	С
inversão mamilar	inversão mamilar	С
espessamento cutâneo	espessamento cutâneo	С
edema	edema	С
eritema	eritema	С
mastite	mastite	С
espessamento trabecular	espessamento trabecular	С
lesão cutânea	lesão cutânea	С
ulceração cutânea	ulceração cutânea	С
adenopatia axilar	adenopatia axilar	С
adenomegalia axilar	adenomegalia axilar	С
cavado axilar	cavado axilar	С
axila positiva	axila positiva	С
Gânglio axilar suspeito	Gânglio axilar suspeito	С

Testing 86 Individual BI-RADS Terms

Experiments:

Performed by <u>4 people</u> - two female and two male Each of these people used <u>3 different devices</u>:

- Laptop with an external microphone NGS brand
- □ Same **laptop** with **built-in microphone**
- Smartphone

TABLE III Performance per experiment

Device	Person	C(%)	AC(%)	I(%)	C+AC(%)
	A	67.4	8.2	24.4	75.6
	→ B	77.9	5.8	16.3	83.7
Laptop with ext micro	C	68.6	9.3	22.1	77.9
	D	66.3	7.0	26.7	73.3
	Av g.	70.1	7.5	22.4	77.6
	A	68.6	93	22.1	77.9
	→ B	74.4	4.7	20.9	79.1
Laptop with int micro	C	67.4	4.7	27.9	72.1
	D	66.3	3.5	30.2	69.8
	Avg.	69.2	5.5	25.3	74.7
	A	69.8	8.1	22.1	77.9
Smartphone	→ B	74.4	7.0	18.6	81.4
	C	70.9	7.0	22.1	77.9
	D	60.5	5.8	33.7	66.3
	Avg.	68.9	7.0	24.1	75.9

TABLE IV Averages per Person

	Person	C(%)	AC(%)	I(%)	C+AC(%)
	A	68.6	8.5	22.9	77.1
→	В	75.6	5.8	18.6	81.4
	С	69.0	7.0	24.0	76.0
	D	64.3	5.5	30.2	69.8

Outline

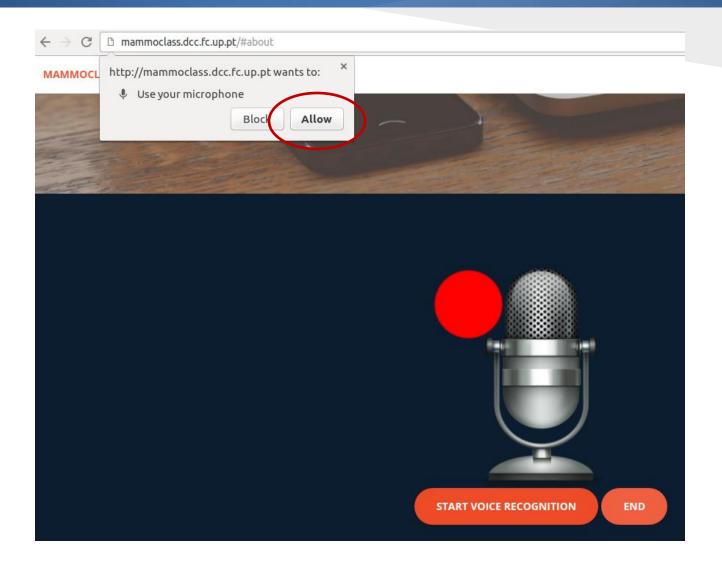
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Web Speech API applied to MammoClass



mammoclass.dcc.fc.up.pt

Web Speech API applied to MammoClass

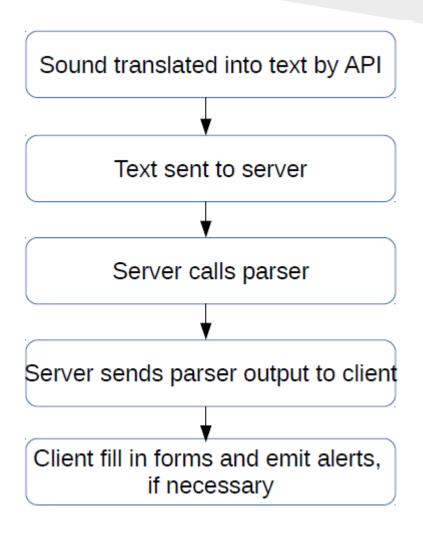


Web Speech API applied to MammoClass

No QSE da mama esquerda observa-se opacidade nodulariforme de contornos <u>espiculados</u> e parcialmente obscuros, medindo aproximadamente 2cm, que corresponde no estudo ecográfico complementar a nodulo sólido, com cerca de 1.5cm

MAMMOCLASS V2			9	STT	DATA MANUALLY	FILL FORM	ABOUT
Patient's age:							
Mass size (mm):							
Breast Composition:	Select a value	¥	The parser does not extract				
Mass Shape:	Select a value	*	This finding does not exist				
Mass Margins:	Spiculated	,					
Mass Density:	Automatic	*	This finding does not exist				
Mass clockface location:	Select a value	٧	The parser does not extract				
Side:	Select a value	٧	The parser does not extract				
Quadrant:	Select a value	*	The parser does not extract				
Depth:	Select a value	٧	The parser does not extract				
		MAKE PREDIC	TION RESET				

Flow Chart – STT API



Conclusions

 We provide to the user an interface where medical reports can be dictated as opposite to input in forms or textboxes

 Although the recognized text sometimes differs from the original written report, the most relevant BI-RADS terms are still recognized

 Implementation of Speech-to-Text interface and all the core to handle the API and send the results to the server

Future Work

 Speech interfaces for long sentences in Portuguese need to be improved

We would like to design and implement our own tools for recognizing
 Portuguese terms, which could be independent of voice type or
 entonation, and that could be trained only on the subset of words used
 in the area of breast cancer

Thanks

Questions?

Appendix

References

- [1] J. du Toit, R. Hattingh, and R. Pitcher, "The accuracy of radiology speech recognition reports in a multilingual south african teaching hospital," BMC Medical Imaging, vol. 15, no. 1, pp. 1–5, 2015. [Online]. Available: http://dx.doi.org/10.1186/s12880-015-0048-1
- [2] S. Basma, B. Lord, L. M. Jacks, M. Risk, and S. A. M., "Error rates in breast imaging reports: comparison of automatic speech recognition and dictation transcription." AJR Am J Roentgenol, vol. 197, pp. 923–927, 2011.
- [3] R. Hoyt and A. Yoshihashi, "Lessons learned from implementation of voice recognition for documentation in the military electronic health record system," Perspectives in Health Information Management, no.7(Winter):1e, 2010.
- **[4]** S. McGurk, K. Brauer, T. V. Macfarlane, and K. A. Duncan, "**The effect of voice recognition software on comparative error rates in radiology reports**," The British Journal of Radiology, vol. 81, pp. 767–770, 2008.
- [5] I. Hammana, L. Lepanto, T. Poder, and M. S. Bellemare, C. Ly, "Speech recognition in the radiology department: a systematic review," HIM J., vol. 44, no. 2, pp. 4–10, 2015.

References (cont.)

[6] Cunha et al. H. Nassif, F. Cunha, I. C. Moreira, R. Cruz-Correia, E. Sousa, D. Page, E. S. Burnside, and I. de Castro Dutra, "**Extracting bi-rads features from portuguese clinical texts**," in IEEE International Conference on Bioinformatics and Biomedicine, BIBM 2012, 2012, pp. 1–4.

State of the Art

[7] Kang et al. Use speech recognition technology in surgical pathology and conclude that it is useful in their anatomic pathology workflow and provides a goo return on investment, error reduction, and cost savings.

Boolean Table

Formação nodular hiperdensa com contornos <u>espiculados</u> com cerca de <u>3 cm</u> na transição dos quadrantes inferiores da mama direita.

Foi detectada tambem uma margem <u>lobular</u>. achados imagiológicos muito sugestivos de malignidade - bi-rads - 5.