

## Machine Learning Assisted Discovery of Photoluminescent Materials

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**Predict quantitative information** 









### I) Classification:

Prediction of the photoluminescence color for Li<sub>2</sub>BaSiO<sub>4</sub>:Eu,Ce

#### II) Regression:

Prediction of the CIE coordinates of the photoluminescence for  $Ca_{14}Zn_6Ga_{10}O_{35}$ :Mn,Tm





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365 nm / 254 nm).



#### Li<sub>2</sub>BaSiO<sub>4</sub>:Eu,Ce (LBSO:Eu,Ce)

























#### Machine learning to visualize the experimental conditions to target the white luminescence

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#### Machine learning to visualize the experimental conditions to target the white luminescence





Machine learning to identify the experimental conditions and rationalize the photoluminescence colors







Evolution of the emission intensities associated with  $Ce^{3+}$  ( $\lambda_{em} = 400$  nm, blue line),  $Eu^{2+}$  ( $\lambda_{em} = 510$  nm, green line), and  $Eu^{3+}$  ( $\lambda_{em} = 704$  nm, red line) vs. the temperature of oxidation for LBSO:0.5%Eu, 0.5%Ce





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#### **Prediction of the CIE coordinates**







#### CIE coordinates changes according to the excitation wavelength





**Predict quantitative information** 







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Parameters / sample n°	1	2	3	4	5	6	7	8	9	10	11	12
[Mn]	1	1,1	1,1	1,5	1,5	1	1	1,5	1,5	1,5	1	1
[Tm]	2	2,8	2,8	3	3	1,5	1,5	1,5	6	6	3	3
Duration of TTR at 900°C (h)	1	1	12	12	6	2	12	12	12	24	2	6
<u>λexc (nm)</u>	280-396	280-396	280-396	280-296	280-296	280-296	280-296	280-296	280-296	280-296	280-296	280-296

CIE 1931



4 parameters for the modulation of the resulting emission color



XGBoost model (eXtreme Gradient Boosting)

#### **Prediction of the CIE coordinates**



Validation on 20% of the resulting data ( = 165 experimental points)

# Experimental Profil Profil

#### Representation of the resulting emission color

Experimental validation









Conclusion

ML to identify the key parameters influencing the property.









- ML to identify the key parameters influencing the property.
- ML to predict accurately the property as function of multiple variables.

#### $Ca_{14}Zn_6Ga_{10}O_{35}$ :Mn,Tm



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Romain GénoisHailong YuanEstelle GlaisPhD studentPhD studentPostdoc

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# THANK YOU FOR YOUR ATTENTION



# An endless search with short term challenges



Different stoichiometries, structures of organic molecules, ....

Number of combinations > Number of atoms in universe

Classification of the photoluminescence color

#### INSTITUT DES MATÉRIAUX DE NANTES JEAN ROUXEL

# 88 samples prepared from the eight synthesized samples by oxidation of the dopants under a gradient of temperature



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Strategy

 $\boldsymbol{\bigstar}$  Trichromatic emission : three emission centers

