



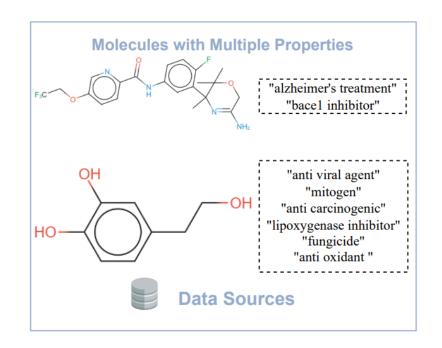
L+M-24: Building a Dataset for Language+Molecules @ ACL 2024

- Existing datasets were:
 - 1. Small and scraped from existing databases
 - 2. Large but noisy and constructed by performing entity linking on the scientific literature
 - 3. Built by converting property prediction datasets to natural language using templates
- The goal was to focus on 3 key benefits of natural language:
 - Compositionality
 - Functionality
 - Abstraction
- We grouped properties into four key categories of interest:
 - Biomedical
 - Light and Electricity
 - Human Interaction and Organoleptics
 - Agriculture and Industry

Compositionality, Functionality, and Abstraction



- To focus on functionality and abstraction, we used specific source datasets containing (mostly) natural language annotations rather than numerical properties.
- To add compositionality, we composed multiple properties together using natural language templates.
 - Some property combinations were held-out of the training set. This may have caused certain properties to be poorly represented in the training data.





Compositional Captions

'The molecule is both a alzheimer's treatment and a bace1 inhibitor.

The molecule is a mitogen and lipoxygenase inhibitor, belonging to the anti oxidant class, and is characterized as anti viral agent, anti carcinogenic, and fungicide.



GPT 4 Written Templates



- Properties including odor, taste, polymerization, and decomposition were taken from PubChem.
 - Descriptions are written free-form and do not fit into categories
- PubChem Examples:
 - PLEASANT GREEN, FRUITY, RUM ODOR

- PUNGENT ODOR SUGGESTIVE OF YLANG-YLANG
- Mercaptan-like odor

- When subjected to high temperatures alone or in the presence of catalysts or strong oxidizing agents it is possible that violent polymerization will take place.
- Woody, floral, berry, fruity with powdery nuances

Data Sources - Chemical Function



 Properties were extracted from the patent literature and standardized. We performed additional processing and kept properties in the following categories:

- "X-icide"
- "anti-X"
- "X treatment"
- "X modulators"
- "X inhibitors"
- "X agonists"
- "X antagonists"
- "light",
- "electricity.

CheF Examples:

- Inhibiting c-Met kinase
- Antimicrobial / Iron(II)-dependent / Biofilm
- herbicide / antidiabetic / urea



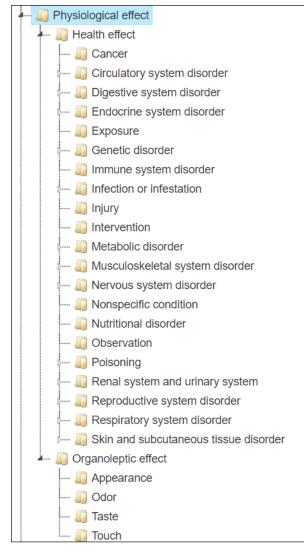
Data Sources - ChemFOnt: Chemical Function Ontology



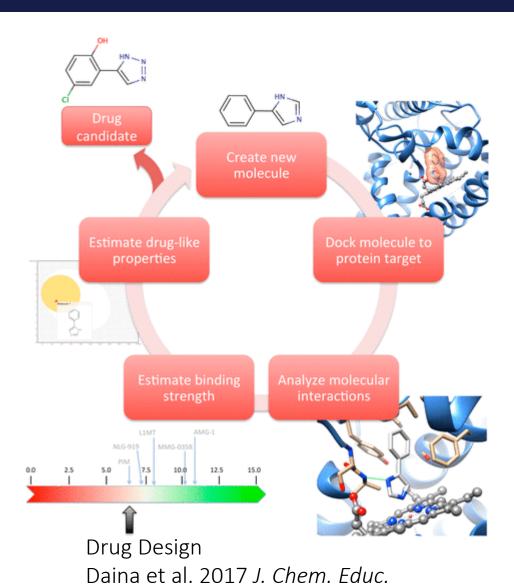
• We considered organoleptic effects, compound roles, and health effects

ChemFOnt Examples:

Insectiphile (HMDB: HMDB0034927)







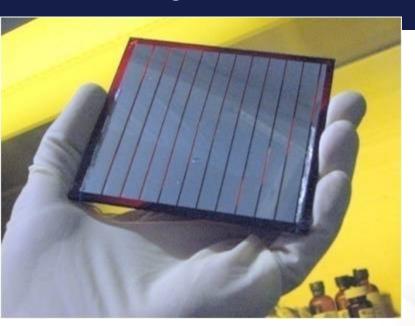
Salient Properties (a sample):

- Anti neoplastic used in the treatment of cancer
- Glaucoma Treatment
- Asthma preventive
- Pyrogenic induces feaver
- Circulatory stimulant increases blood flow
- Glucocorticoid receptor modulator an experimental drug class with anti-inflammatory properties
- Tyrosine kinase inhibitor inhibits tyrosine kinases, with important applications in leukemia
- Serotonin agonist binds and activates serotonin receptors, increasing the amount of the neurotransmitter in the body

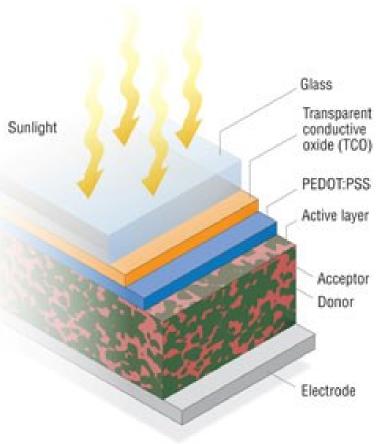


Light and Electricity: Organic Photovoltaics





Organic solar cells are lightweight, flexible, and cheap, but current cells are still inefficient and unstable.



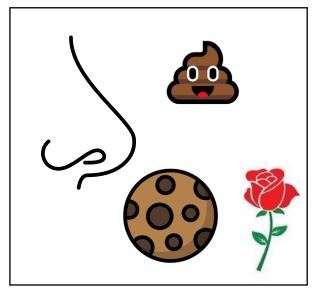
Department of Energy

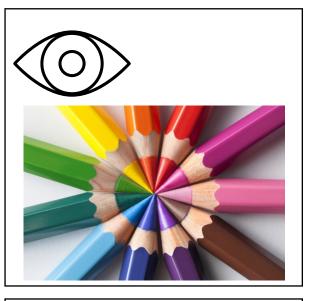
Salient Properties (a sample):

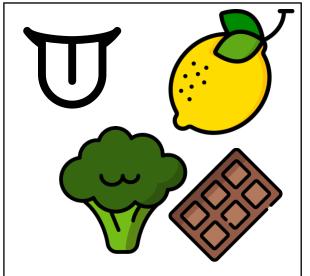
- Organic electroluminescent compound A material which emits light in the presence of a electric current
- Fluorescent Emission of light by a substance that has absorbed light
- Phosphorescent Delayed emission of the absorbed light
- Photoacid generator Compounds that produce acids upon exposure to light
- Photopolymerization A chemical reaction linking small monomers when exposed to light
- Photochromic Undergoes a reversible change in color upon exposure to light

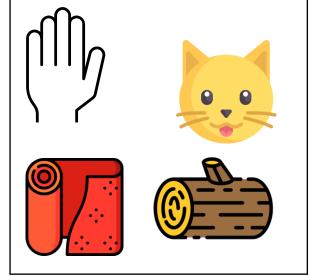
Human Interaction and Organoleptics











Salient Properties (a sample):

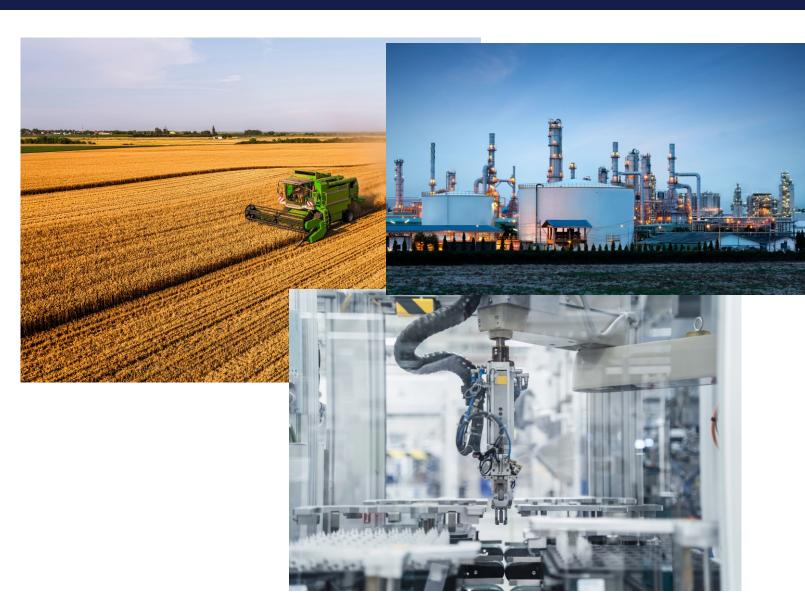
- Bitter
- Nephrotoxic agent Toxic to the kidneys
- Artichoke
- Red cedar
- Cucumber seed



Interesting reading: A principal odor map unifies diverse tasks in olfactory perception

Agriculture and Industry





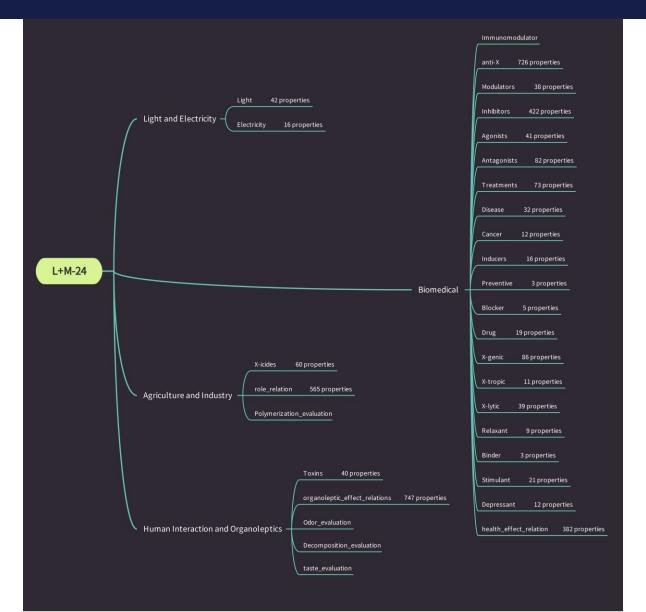
Salient Properties (a sample):

- Pesticide An agent which kills pests
- Acidity regulator Food additives which change or maintain acidity
- Crustacicide An agent which kills crustaceans
- Fertilizer improves growth and productiveness of plants
- Lubricant A substance which helps reduce friction between surfaces



THRUST 1 Property Breakdown





Group	Property-Molecule Pair Count					
Total	1512865					
Biomedical	776712					
anti-X	24884					
Modulators	2787					
Inhibitors	23257					
Agonists	1161					
Antagonists	3172					
Treatments	53070					
Disease	316380					
Cancer	41456					
Inducers	31					
Preventive	0					
Blocker	47					
Drug	260					
X-genic	172					
X-tropic	17					
X-lytic	84					
Relaxant	40					
Binder	4					
Stimulant	60					
Depressant	52					
health_effect_relations	309532					
Light and Electricity	14077					
Light	11069					
Electricity	3008					
Human Interaction	27457					
Toxins	1070					
organoleptic_effect_relations	20501					
Agric. and Industry	694619					
X-icides	809					
role_relation	693648					



THRUST 1 Baseline Model Examples



Input	MoIT5-small	MoIT5-base	MolT5-large	Meditron	Ground Truth
The molecule is a luminescent member of the		\bigcirc - \bigcirc	3000		3008008
organic light-emitting class.	Invalid				
The molecule is both a platelet aggregation inhibitor and a cell adhesion inhibitor.	Cc1ccc(-c2ccc(-c3ccc(-c4ccc(-c5ccccc5)cc5)cc4)cc3)cc2)cc1 Invalid	XIOIOL	CC(C)(C)OC(=O)NC(C(=O) N1CCCC1C(=O)NC(Cc1ccccc1) C(=O)NC(Cc1ccccc1)C(=O)NC (Cc1ccccc1)C(=O)NC(Cc1ccccc1) C(=O)NC(Cc1ccccc1)C(=O)O' Invalid	, ~ l l	~~~
The molecule is a muscarinic agonist that impacts pain treatment and is both alzheimer's treatment and anxiety treatment.	Cc1ccc(-c2ccc(-c3ccc(-c4ccc(-c5cccc5)cc5)cc4)cc3)cc2)cc1 Invalid	xiololo	+0000+	Sou Sou	NH ₂
The molecule is a jak2 inhibitor and is cancer treatment.		◇ - \$ ~ \$	\$ 5000+)
The molecule is both a anti psychotic and a nmda antagonist.	Cc1ccc(-c2ccc(-c3ccc(-c4ccc(-c5cccc5)cc4)cc3)cc2)cc1 Invalid	400	+20040+	404	40
The molecule is a factor ixa inhibitor, a factor xa inhibitor, and anti thrombotic.	Cc1ccc(-c2ccc(-c3ccc (-c4ccc(-c5ccc(-c6ccc cc6)cc6)cc4)cc4)cc3)cc2)cc1 Invalid	400	trodolog	010	, 710
The molecule is a flavoring agent and a nutrient, as well as nutty and green.	2 december 1		£2,400 x	- XXX	



Results - Molecule Captioning

		Overall	Translation						
Team			Metric Increase	BLEU-2	BLEU-4	ROUGE-1	ROUGE-2	ROUGE-L	METEOR
avaliev	RAG_SIM_098	27.08	6.37	73.8	1 53.04	80.06	60.17	57.5	77.45
qizhipei	BioT5+_large_voting	14.66	6.45	75.5	8 54.77	7 79.41	59.89	57.46	75.43
protonunfold	SciMind	12.39	5.77	75.6	6 54.98	78.24	58.42	56.34	74.76
NLPeople	Ensembled	12.3	5.68	75.5	4 54.83	78.1	58.47	56.37	74.57
hecao	bioagent	10.95	5.57	74.1	1 53.84	78.73	59.38	57.06	74.08
xwk89	mistral_4b9_e1	10.8	5.56	74.3	8 54.08	78.49	59.49	56.82	73.91
mengmeng	Mistral	10.37	5.48	75.0	4 54.56	78.1	58.81	56.42	73.73
langmolecules	Meditron	10.34	5.47	75.1	6 54.72	77.97	58.75	56.33	73.69
NLPeople	Rank_model_1	9.94	4.8	74.7	3 54.2	77.3	57.7	55.53	73.26
	ALMol~10%DataTrai								
dimitris	ned	9.61		74.7					
xygui	MDEG	9.43	2.96	73.9	8 53.33	3 75.08	54.39	52.58	72.21
danielshao	SMol+LPM	7.83	2.7	72	2 52.02	74.76	56.1	53.34	71.57
xwk89	mistral_e1	7.45	3.46	73.1	8 53.24	75.29	56.74	54.39	71.75
duongttr	Mol2Lang-VLM	4.52	4.11	73.4	3 53.19	76.72	57.67	55.43	72.05
langmolecules	MolT5-Large	4.22	3.64	73.6	3 53.2	75.79	56.47	54.42	72.16
	phi3-knowchem-sft-								
bluesky333	beam1	2.87	0.81	70.5	6 50.83	72.61	53.61	52.01	69.01
langmolecules	MolT5-Base	1.06	1.2	69.8	3 50.56	73.34	54.55	52.86	69.86
langmolecules	MolT5-Small	0	0	66.8	2 48.29	72.8	54.44	53.33	68.14
guiyike	yike	-16.64	-43.45	12	8 6.37	7 25.9	13.83	24.1	20.11

Dverall



Results - Molecule Captioning - Property F1 (%)

RAG_SIM_ avaliev 088	Held-out													Light +	Agr. +	Human		Property F-	Metric	Overall		
avalley 088 27.08 33.99 26.99 27.9 4.18 3.55 72.32 0 6.74 71.92 72.71 44.52 11.07 68.09 55.7 65.49 41.53 52.07 BIOTS+_Isr BIOTS+_I	sease Combos	Disease	Cancer	Agonists	Treatments	Antagonist	Modulators	anti-X	Inhibitors	Electricity	Light	Toxins	X-icides	Electro	Industry	Interaction	Biomedical	1 Increase	Increase	Increase		Team
glathjelle ge voting 14.66 17.39 13.76 19.76 4.01 3.07 28.2 0 6.25 31.28 25.12 20.55 3.64 37.02 30.77 23.89 16.5 99.47 protonumfor old SciMind 12.3 14.6 11.51 18.17 3.94 2.93 21 0.04 6.2 23.95 18.05 18.06 2.55 30.14 25.36 19.42 14.81 57.54 NI.People Ensembled 12.3 14.5 11.63 17.86 4.02 2.97 21.67 0.04 6.33 24.74 18.59 17.02 2.77 30.96 25.89 18.61 15.1 53.69 hecao bloagent 10.95 12.74 9.94 16.86 3.9 2.76 16.24 0 6.43 18.12 14.37 18.84 2.48 26.78 23.41 16.33 14.02 50.28 mistral_ab xwk89 9_e1 10.8 12.54 9.93 16.98 3.85 2.81 16.1 0 6.26 17.15 15.05 15.2 2.12 31.67 19.11 14.34 14.59 51.17 mengmeng Mistral and the strain of the st	66.15 69	7 66	53 52.07	41.53	7 65.49	55.	7 68.09	11.07	. 44.52	72.71	71.92	0 6.7		72.32	3.55	4.18	27.9	26.99	33.99	27.08		avaliev
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mistral_ab	67.44 69	9 67	.1 53.69	15.1	18.61	25.89	7 30.96	2.77	17.02	18.59	24.74	4 6.3	0.0	21.67	2.97	4.02	17.86	11.63	14.5	12.3	Ensembled	NLPeople
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les Meditron 10.34 11.96 9.7 16.87 3.75 2.83 15.36 0.04 6.07 18.19 12.53 14.71 2.17 24.72 20.61 14.35 10.54 53.74 Rank_mod NLPeople el 1 9.94 11.66 9.88 16.5 3.85 2.87 16.28 0.08 5.97 19.29 13.28 13.91 2.14 23.71 17.88 13.87 12.27 50.82 ALMoi-10 %DataTrain dimitris ed 9.61 11.39 10.05 15.75 3.74 2.78 17.94 0.21 6 17.11 18.77 12.39 1.83 17.34 18.53 11.14 11.16 51.66 xygui MDEG 9.43 11.59 8.74 16.08 3.57 2.72 12.6 0 5.86 12.68 12.52 13.2 1.87 29.96 19.61 12.47 14.08 54.55 danielshao SMol+LPM 7.83 9.54 8.55 14.97 2.39 2.62 14.2 0 3.19 16.38 12.01 10.9 1.61 22.13 17.43 9.66 9.93 40.41 xwk89 mistral_el 7.45 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Mol2Lang-duongttr VLM 4.57 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MOITS-les Large 4.27 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem	67.45 69	1 67	5 53.61	10.55	14.34	21.2	2 24.43	2.12	14.74	12.64	18.49	4 6.0	0.0	15.57	. 2.83	3.81	16.72	9.73	12	10.37		
Rank_mod NLPeople el_1 9,9 11.66 9,88 16.5 3.85 2.87 16.28 0.08 5.97 19.29 13.28 13.91 2.14 23.71 17.88 13.87 12.27 50.82 ALMoi*To %DataTrain dimitris ed 9,63 11.39 10.05 15.75 3.74 2.78 17.94 0.21 6 17.11 18.77 12.39 1.83 17.34 18.53 11.14 11.16 51.66 xygui MDEG 9,43 11.59 8.74 16.08 3.57 2.72 12.6 0 5.86 12.68 12.52 13.2 1.87 29.96 19.61 12.47 14.08 54.55 danielshao SMoi+LPM 7.83 9.54 8.55 14.97 2.39 2.62 14.2 0 3.19 16.38 12.01 10.9 1.61 22.13 17.43 9.66 9.93 40.41 xwk89 mistral_el 7.45 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Moi2Lang-duongttr VLM 4.52 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MoiT5-les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem																						langmoleci
NLPeople el_1 9.94 11.66 9.88 16.5 3.85 2.87 16.28 0.08 5.97 19.29 13.28 13.91 2.14 23.71 17.88 13.87 12.27 50.82 ALMol~10 %DataTrain dimitris ed 9.61 11.39 10.05 15.75 3.74 2.78 17.94 0.21 6 17.11 18.77 12.39 1.83 17.34 18.53 11.14 11.16 51.66 xygui MDEG 9.43 11.59 8.74 16.08 3.57 2.72 12.6 0 5.86 12.68 12.52 13.2 1.87 29.96 19.61 12.47 14.08 54.55 danielshao SMol+LPM 7.83 9.54 8.55 14.97 2.39 2.62 14.2 0 3.19 16.38 12.01 10.9 1.61 22.13 17.43 9.66 9.93 40.41 xwk89 mistral_e1 7.45 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Mol2Lang- duongttr VLM 4.52 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MolT5- les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3- knowchem	67.4 69	1 6	4 53.74	10.54	14.35	20.6	7 24.72	2.17	14.71	12.53	18.19	4 6.0	0.0	15.36	2.83	3.75	16.87	9.7	11.96	10.34		les
ALMol~10 %DataTrain dimitris ed 9.6 11.39 10.05 15.75 3.74 2.78 17.94 0.21 6 17.11 18.77 12.39 1.83 17.34 18.53 11.14 11.16 51.66 xygui MDEG 9.4 11.59 8.74 16.08 3.57 2.72 12.6 0 5.86 12.68 12.52 13.2 1.87 29.96 19.61 12.47 14.08 54.55 danielshao SMol+LPM 7.8 9.54 8.55 14.97 2.39 2.62 14.2 0 3.19 16.38 12.01 10.9 1.61 22.13 17.43 9.66 9.93 40.41 xwk89 mistral_e1 7.4 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Mol2Lang-duongttr VLM 4.5 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MolT5-les Large 4.2 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem	66.01 69	2 66	7 50.01	12.2	12 07	17 0	1 22 71	2.17	12 01	12 20	10.20	o 50	0.0	16 29	2 07	2 00	16.5	0.00	11 66	0.0/	_	NI Pooplo
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dimitris ed 9.61 11.39 10.05 15.75 3.74 2.78 17.94 0.21 6 17.11 18.77 12.39 1.83 17.34 18.53 11.14 11.16 51.66 xygui MDEG 9.43 11.59 8.74 16.08 3.57 2.72 12.6 0 5.86 12.68 12.52 13.2 1.87 29.96 19.61 12.47 14.08 54.55 danielshao SMol+LPM 7.83 9.54 8.55 14.97 2.39 2.62 14.2 0 3.19 16.38 12.01 10.9 1.61 22.13 17.43 9.66 9.93 40.41 xwk89 mistral_e1 7.45 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Mol2Lang-duongttr VLM 4.52 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MolT5-les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem																						
danielshao SMol+LPM 7.83 9.54 8.55 14.97 2.39 2.62 14.2 0 3.19 16.38 12.01 10.9 1.61 22.13 17.43 9.66 9.93 40.41 xwk89 mistral_e1 7.45 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Mol2Lang-duongttr VLM 4.52 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MolT5-les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem	67.09 69	6 67	.6 51.66	11.16	3 11.14	18.5	3 17.34	1.83	12.39	18.77	17.11	1	0.2	17.94	2.78	3.74	15.75	10.05	11.39			dimitris
xwk89 mistral_e1 7.45 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Mol2Lang-duongttr VLM 4.52 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MolT5-les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem	67.1 69	5 6	8 54.55	14.08	12.47	19.6	7 29.96	1.87	13.2	12.52	12.68	5.8	,	12.6	2.72	3.57	16.08	8.74	11.59	9.43	MDEG	xygui
xwk89 mistral_e1 7.45 8.78 8.23 14.44 2.69 2.74 13.05 0 4.33 13.43 12.67 14.23 2 26.09 18.6 12.99 11.75 20.44 Mol2Lang-duongttr VLM 4.52 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MolT5-les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem																						
Mol2Lang- duongttr VLM	64.84 68	1 64.	3 40.41	9.93	9.66	17.43	1 22.13	1.63	. 10.9	12.01	16.38	0 3.19		14.2	2.62	2.39	14.97	8.55	9.54	7.83	SMoI+LPM	danielshao
duongttr VLM 4.52 4.66 5.76 10.73 3.18 2.36 6.78 0 5.5 9.39 4.18 1.08 0.3 0 2.49 1.72 0.69 40.62 langmolecu MolT5- les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3-knowchem	59.5 69	4 59	5 20.44	11.75	5 12.99	18.	26.09	2	14.23	12.67	13.43	0 4.3		13.05	2.74	2.69	14.44	8.23	8.78	7.45	mistral_e1	xwk89
langmolecu MolT5- les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3- knowchem																					_	
les Large 4.22 4.42 5.81 10.33 3.2 2.36 7.36 0 5.58 7.77 6.94 0.65 0.13 0.39 0.11 2.02 0.58 38.34 phi3- knowchem	67.61 69	2 67.	9 40.62	0.69	9 1.72	2.49	3 0	0.3	1.08	4.18	9.39	0 5.		6.78	2.36	3.18	10.73	5.76	4.66	4.52		-
phi3- knowchem	66.97 69	4 66	Q 20 2/	0.59	1 2.02	0.1	2 0.30	0.13	0.65	6 0/1	7 77	n 55		7 36	2 36	2.1	10 22	5 91	1 12	12		
knowchem	00.57	, 00.	0 30.34	0.50	2.02	0.1.	0.59	0.1.	0.03	0.54	7.77	J. J.		7.50	2.30	5.2	10.55	5.61	4.42	4.22	Large	163
	63.71 64	9 63.	21 35.39	0.21	2.04	0.29	3 0	0.23	0.93	4.21	8.01	5.4		6.11	2.13	3.15	10.05	5.36	3.56	2.87		
langmolecu MolT5- les Base 1.06 1.02 3.99 8.27 2.68 2 3 0 4.63 5.52 0.48 0.08 0.04 0 0 1.79 0 18.04	47.64 68	4 47	0 18.0/	() 1 70		1 0	0.04	n na	0.48	5 52	0 46		2	2	2 68	8 27	3 00	1 02	1.06		
langmolecu MolT5-	47.04	7.	5 10.04		1.73		. 0	0.04	0.00	0.40	5.52	7.0		3	2	2.00	0.27	5.93	1.02	1.00		
les Small (0 3.23 7.87 0.27 1.65 3.12 0 0 6.24 0 0.06 0 0 0 1.66 0 17.99	41.09 65	9 41	0 17.99	(1.66		0 0	(0.06	. 0	6.24	0		3.12	1.65	0.27	7.87	3.23	0	(_





			Test Metric								
Team	Model	Overall Increase	Increase	BLEU	Exact Match	Levenshtein	Validity	MACCS FTS	RDK FTS	Morgan FTS	FCD
qizhipei	BioT5+_large	12.97	13.2	73.17	0.01	41.05	100	76.05	68.7	50.05	3.13
protonunfold	SciMind	12.68	3 12.76	73.44	·	40.35	99.78	75.06	67.05	47.82	2.54
avaliev	PLAIN	12.41	12.39	71.82	0.01	43.91	98.98	74.97	66.85	48.92	0.28
mengmeng	Mistral	12.26	5 12.22	70.56	C	43.75	99.4	75.6	67.57	48.62	2.01
langmolecules	Meditron	11.81	11.66	68.84	0.01	. 46.47	99.54	75.59	67.66	48.72	2.44
dimitris	ALMol~10%Data	10.41	9.26	69.74	0.01	. 43.24	92.84	70.22	62.79	42.96	3.05
hecao	bioagent_epoch5	10.21	10.67	61.98	0.02	47.12	99.67	75.94	68.38	46.92	2.17
langmolecules	MolT5-Base	10.07	7 10	67.04	C	45.71	99.89	74.61	63.7	46.29	nan
danielshao	SMol+LPM	8.74	8.45	59.74	0.01	. 55.09	97.66	74.35	66.79	46.6	4.25
hecao	bioagent	6.39	5.36	51.5	0.01	70.67	98.05	74	66.43	45.56	3.74
langmolecules	MolT5-Large	3.65	5 4.78	55.31		56.47	99.12	74.14	63.4	38.54	17.63
erikxiong	PUF	(0	55.44	· C	57.21	81.03	63.06	56.83	36.69	nan
langmolecules	MolT5-Small	C	0	55.44	C	57.21	81.03	63.06	56.83	36.69	nan
ndhieunguyen	Lang2mol-diff	-1.21	-0.75	54.15	C	55.26	100	59.6	32.44	31.98	10.71
guiyike	Nano	-7.39	-6.64	43.51	. С	83.38	100	49.25	37.82	23.52	5.64



Results - Molecule Generation - Held-out Combos



			Withheld Combo								
Team	Model	Overall Increase	Increase E	LEU	Exact Match	Levenshtein	Validity	MACCS FTS	RDK FTS	Morgan FTS	FCD
qizhipei	BioT5+_large	12.97	12.74	78.48	0.02	43.99	100	86.54	80.43	59.54	4.32
protonunfold	SciMind	12.68	3 12.6	78.99	C	43.39	99.77	86.03	79.16	58.06	3.06
avaliev	PLAIN	12.41	12.42	78.19	C	47.84	99.64	86.3	79.16	59.02	0.35
mengmeng	Mistral	12.26	12.31	78.08	C	46.34	99.48	86.14	78.75	59.39	2.27
langmolecules	Meditron	11.81	11.96	77.11	C	48.29	99.53	86.23	78.91	59.57	2.59
dimitris	ALMol~10%Data	10.41	11.56	76.93	0.01	. 45.68	98.99	85.9	79.35	55.28	3.51
hecao	bioagent_epoch5	10.21	9.75	65.86	C	52.06	99.83	86.55	80.13	55.68	3.24
langmolecules	MolT5-Base	10.07	10.15	72.48	C	50.9	99.84	86.22	78.23	57.56	inan
danielshao	SMol+LPM	8.74	9.03	65.95	C	56.31	99.1	86.14	79.77	56.75	4.36
hecao	bioagent	6.39	7.42	64.6	C	65.63	99.11	85.86	79.58	54.81	4.18
langmolecules	MolT5-Large	3.65	5 2.36	57.74	C	66.94	99.17	83.19	74.77	40.97	'nan
erikxiong	PUF	c	0	56.96	C	72.44	91.89	81.03	73.36	41.6	nan
langmolecules	MolT5-Small	C	0	56.96	C	72.44	91.89	81.03	73.36	41.6	inan
ndhieunguyen	Lang2mol-diff	-1.21	-1.67	59.38	C	63.17	100	73.26	38.6	39.76	6.38
guiyike	Nano	-7.39	-8.15	44.84	C	85.92	100	59.67	47.95	30.91	7.86

Mystery Molecules!

- 137 molecules without ground truth properties were added to the test split.
 - These consisted of:
 - Small molecule drugs approved by the FDA in the 2020s
 - Molecules suggested by the scientific advisory board
- The goal is to create relevant property predictions important molecules by ensembling the different submissions to the task.
 - We plan to make a submitted prediction dataset available for future work to build more sophisticated ensembles.
- In particular, we were interested in looking for "off-label" properties of newly approved drugs
 - We looked at a voting ensemble of properties suggested for these molecules by multiple models.





Rimegepant

- a medication used for the acute treatment of migraine with or without aura in adults and the prophylactic/ preventive treatment of episodic migraine in adults.
- Suggested to be a cgrp receptor antagonist (correct)



Ozanimod

- an immunomodulatory medication for the treatment of relapsing multiple sclerosis and ulcerative colitis
- Suggested to be an antifungal by several models



Bempedoic acid

- Used for treatment of high blood cholesterol levels.
- Predicted by several models as a coating for ship hulls to prevent sea life such as algae and mollusks attaching themselves to the hull.



D-Lactitol monohydrate

- approved by the FDA for use in chronic idiopathic constipation in February 2020.
- Predicted as a nutrient and sweet tasting.
- "Sugar alcohol sweet tastant detectable by humans." -Sigma Aldrich



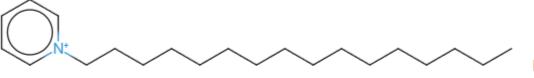
Tucatinib

- An anticancer medication used for the treatment of HER2positive breast cancer
- Predicted as a kinase inhibitor by several models.



Cetylpyridinium bromide

- A surfactant, an antiseptic drug and an EC 2.7.11.18 (myosin-light-chain kinase) inhibitor
- Suggested to be toxic.





Vamorolone

- A corticosteroid used for the treatment of Duchenne muscular dystrophy
- Suggested to an anti-inflammatory (correct)
 - Also predicted to be anti-angiogenic



Palovarotene

- First in class medication for heterotopic ossification
- Suggested to be fluorescent by multiple models



Perfluorohexyloctane

- Used for the treatment of dry eye disease
- Suggested to be a dielectric

(%)

Lenacapavir

- An antiretroviral medication used to treat HIV/AIDS
- Predicted as an anti viral and hiv treatment



Quizartinib

- A kinase inhibitor used to treat Acute Myeloid Leukemia
- Predicted as antiviral by several models

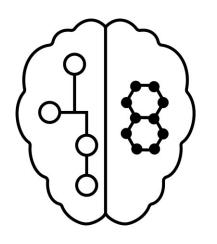


(3)

Thank you to everyone who participated in the shared task!!

- The task was designed to emphasize benefits of natural language in molecule design.
- It considered new metrics for evaluating property prediction in an open-vocabulary setting (specifically, F-1 score)
 - This revealed that performance varies considerably between different types of properties!
- Your submissions have helped us to reveal a number of exciting points
 - Evaluation is still a big challenge!
 - Data splitting when focusing on abstraction, composition, and functionality is challenging and has room for improvement.
 - Inserting relevant external data, especially on molecular interactions, is critical.
 - Model architectures currently favor building from LLMs, but there's significant opportunity for research on novel ways to interpret different data modalities.
- Shared task submitters will be presenting the details of their methods in the upcoming oral and poster presentations!





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