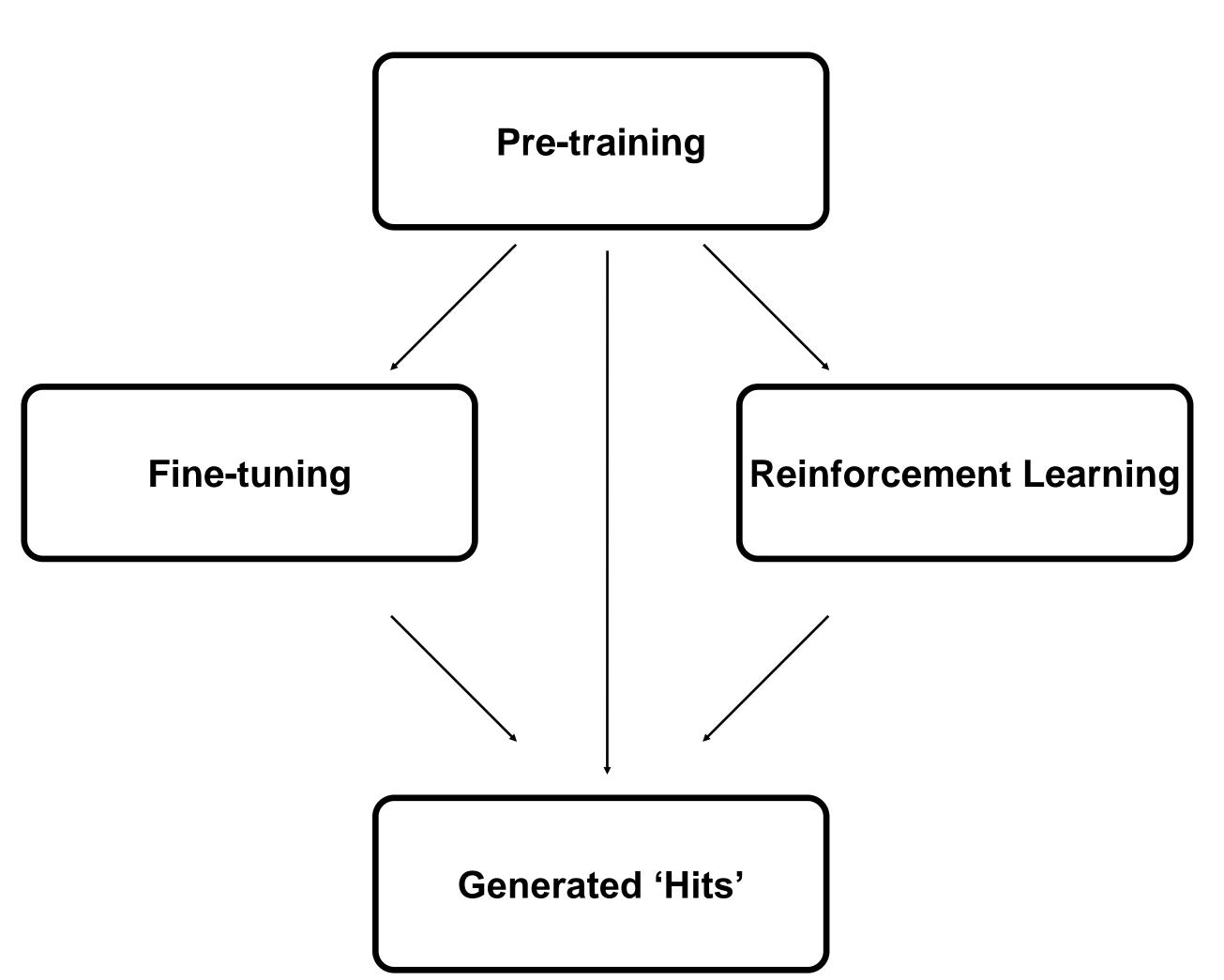
Generative Modeling



Pre-training: Learn SMILES grammar and broad patterns, e.g. frequency of halogens, typical molecular weight, etc. Large, diverse dataset preferred

<u>Fine-tuning:</u> Bias toward patterns of interest by exposing model only to compounds satisfying certain properties, e.g. containing a desired substructure. Need a dataset of such compounds, possibly from virtual screening

Reinforcement Learning: Similar to fine-tuning except the dataset used for biasing is generated on-the-fly. Only need a scoring function specifying the properties of interest

Sample Hit Libraries

Pre-training datasets:

- ChEMBL (1.5M),
- Enamine diverse REAL drug-like (20M)

Fine-tuning datasets:

Virtually screened hits from Enamine 20M (100k)

Library A

Pre-trained on ChEMBL

Hits from filtered, unbiased model using pharmacophore scoring function

~15k

Library B

Virtual screening hits from Enamine 20M

~15k

Library C

Pre-trained on ChEMBL

Use Library B to finetune generative model

Filter generated compounds for hits

~15k

Library D

Pre-trained on Enamine 20M

Hits from filtered, unbiased model using pharmacophore scoring function

~1.5k

Library E

Pre-trained on ChEMBL

Reinforcement learning using pharmacophore scoring function

~15k

Library A: ChEMBL pre-trained, filtered

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13707_Library_A	14207_Library_A	7929_Library_A	370_Library_A	4838_Library_A	9620_Library_A	8691_Library_A	7519_Library_A	12009_Library_A	7206_Library_A
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12583_Library_A	1113_Library_A	7424_Library_A	3011_Library_A	6085_Library_A	10391_Library_A	570_Library_A	641_Library_A	3004_Library_A	9012_Library_A
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1191_Library_A	13837_Library_A	1824_Library_A	13783_Library_A	10672_Library_A	2227_Library_A	1524_Library_A	8372_Library_A	14165_Library_A	1993_Library_A
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5521_Library_A	6221_Library_A	10941_Library_A	1354_Library_A	5459_Library_A	648_Library_A	1496_Library_A	59_Library_A	7176_Library_A	14011_Library_A

Library B: Enamine filtered

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4338_Library_B	7087_Library_B	65_Library_B	5293_Library_B	942_Library_B	6687_Library_B	9175_Library_B	12229_Library_B	1606_Library_B	80_Library_B

Library C: ChEMBL pre-trained, library B fine-tuned

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821_Library_C	2337_Library_C	1812_Library_C	11284_Library_C	7390_Library_C	1094_Library_C	5342_Library_C	11600_Library_C	7121_Library_C	6472_Library_C
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4530_Library_C	8926_Library_C	14699_Library_C	12044_Library_C	14500_Library_C	13111_Library_C	2644_Library_C	4372_Library_C	13479_Library_C	6499_Library_C

Library D: Enamine pre-trained, filtered

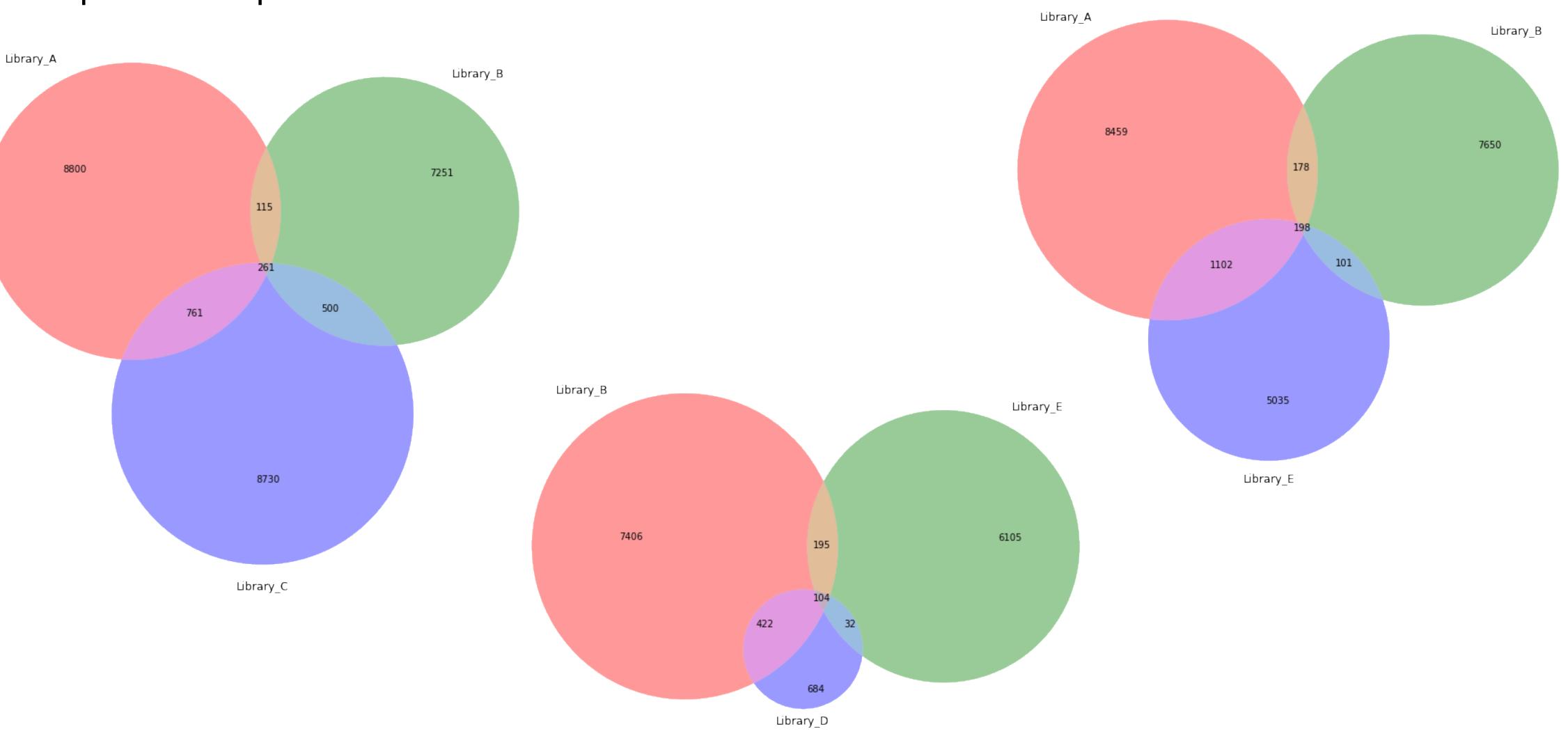
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303_Library_D	1422_Library_D	902_Library_D	27_Library_D	1269_Library_D	1061_Library_D	1429_Library_D	375_Library_D	689_Library_D	1244_Library_D

Library E: ChEMBL pre-trained, reinforcement learning

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5088_Library_E	10085_Library_E	6813_Library_E	6725_Library_E	6569_Library_E	4768_Library_E	2710_Library_E	7711_Library_E	3348_Library_E	5482_Library_E
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6999_Library_E	12702_Library_E	81_Library_E	2677_Library_E	4407_Library_E	11648_Library_E	6248_Library_E	3864_Library_E	2058_Library_E	12590_Library_E
0740		04				-0-tra	Jan	+8-40	Land
3769_Library_E	11300_Library_E	12172_Library_E	7708_Library_E	4277_Library_E	6306_Library_E	13879_Library_E	2923_Library_E	6133_Library_E	4558_Library_E
	auto	DAS				ST. CO	9	PPQ.	
11997_Library_E	11358_Library_E	9918_Library_E	11428_Library_E	7684_Library_E	1139_Library_E	5101_Library_E	11839_Library_E	2073_Library_E	13883_Library_E

Murcko Scaffold Overlap

Find all unique Murcko scaffolds present in each library, compare overlap between libraries



Library A: ChEMBL pre-trained, filtered

Library C: ChEMBL pre-trained, library B fine-tuned Library D: Enamine pre-trained, filtered Library E: ChEMBL pre-trained, reinforcement learning

Generally low overlap between scaffolds in each library