

# Knowledge Graph Embeddings for Salmon Lice Treatment Recommendation

From TransE to ComplEx: Understanding KG Models with PyKeen

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# Research Journey: Solving the Puzzle

- ▶ **I. The Challenge:** Parasitic pressure and treatment paths
- ▶ **II. Methodology:** Finding the best model for aquaculture
  - ▶ **Path 1:** Translation (Baseline & Limitations)
  - ▶ **Path 2:** Rotation (Handling 1-to-N relationships)
  - ▶ **Path 3:** Bilinear (Directional & Asymmetric logic)
  - ▶ **Path 4:** Semantics (Leveraging SBERT)
- ▶ **III. Validation:** Model comparison and performance

## Problem Statement:

- ▶ Salmon farming faces parasitic *Lepeophtheirus salmonis* (sea lice)
- ▶ Global economic impact and animal welfare concerns
- ▶ Multiple treatments available: chemicals, cleaner fish, thermal, mechanical

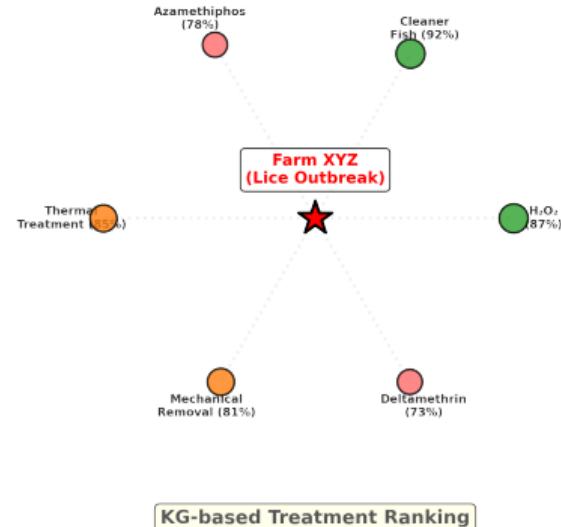
*"Which treatment is best for which specific farm condition?"*

# The Solution: A Knowledge Graph Approach

## Mapping Relationships:

- ▶ Salmon species ↔ Lice strains
- ▶ Treatments ↔ Efficacy
- ▶ Environmental factors ↔ Outbreaks

**Goal:** Structured link prediction for decision support



Treatment recommendation via KG reasoning

# TransE: Translational Embedding

**Definition:**  $h + r \approx t$

- ▶ Relationship  $r$  is a translation vector from head  $h$  to tail  $t$
- ▶ **Strength:** Excellent for hierarchical taxonomies
- ▶ **Limitation:** A single vector cannot point to multiple locations

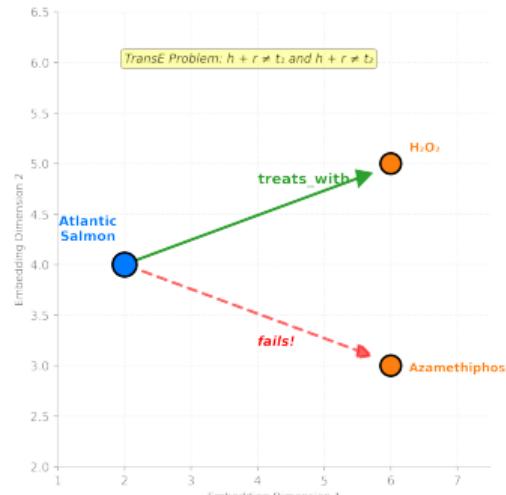


# TransE: Failure on 1-to-N Relations

## The Logic Trap:

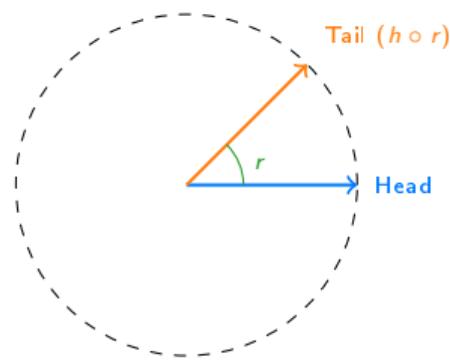
- ▶ (*Salmon, treats, H<sub>2</sub>O<sub>2</sub>*)
- ▶ (*Salmon, treats, Azamethiphos*)
- ▶ If  $h + r = t_1$  and  $h + r = t_2$
- ▶ **Consequence:**  $t_1 \approx t_2$

The model incorrectly treats different chemicals as identical!



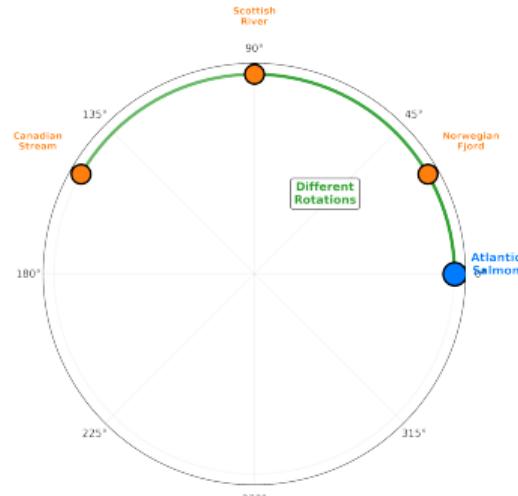
**Definition:**  $t = h \circ r$

- ▶ Map entities into complex space  $\mathbb{C}^d$
- ▶ Relations are rotations on the unit circle
- ▶ **Capability:** Can model complex symmetry and inversion



## Solving the Overlap:

- ▶ Different relations rotate to different points
- ▶  $(\text{Salmon}, \text{migrates\_to}, \text{River\_A})$
- ▶  $(\text{Salmon}, \text{migrates\_to}, \text{River\_B})$
- ▶ **Advantage:** Multiple tails can exist on the complex rim



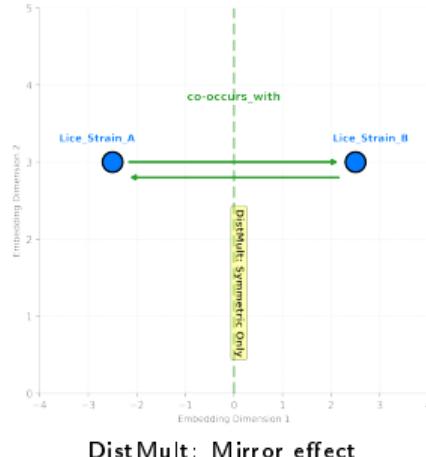
RotatE: Rotation in complex space

# DistMult: High Speed, Low Detail

**Definition:**  $h^T \text{diag}(r)t$

- ▶ Matrix factorization approach
- ▶ **Strength:** Extremely efficient
- ▶ **Flaw:** Intrinsically symmetric

$(A, r, B)$  is the same as  $(B, r, A)$



DistMult: Mirror effect

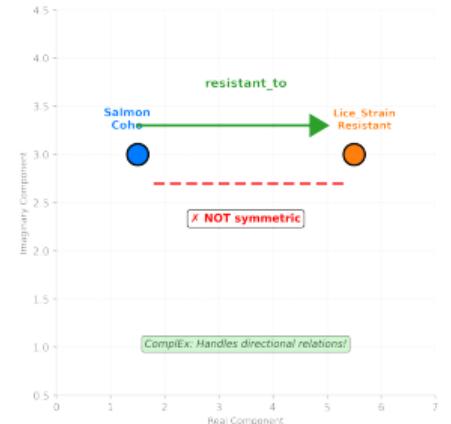
# ComplEx: The Winner for Directional Data

## Why it works:

- ▶ Extends DistMult to complex space
- ▶ Hermitian dot product allows asymmetry
- ▶ Success: Correctly models  $A \rightarrow B \neq B \rightarrow A$

## Example:

- ▶ (*Salmon*, *resistant\_to*, *Lice\_A*)
- ▶ Lice is **not** resistant to Salmon!



ComplEx: Captures flow and direction

# The Problem: What is ID 42?

## Traditional PyKeen Training:

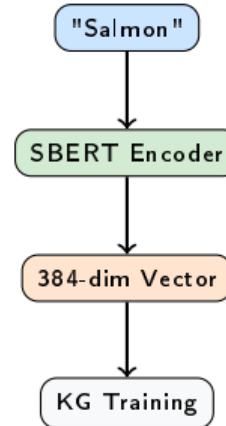
- ▶ Entities: "Atlantic\_Salmon" → ID 42
- ▶ Relations: "treats\_with" → ID 7
- ▶ **Loss:** The model knows nothing about biology at  $t = 0$

*"A Salmon is just more similar to a Trout than to a Chemical."*

# The Solution: SBERT Pre-training

## Semantic Seeding:

- ▶ Convert labels to 384-dim vectors
- ▶ Initialize embeddings with these vectors
- ▶ **Result:** Faster convergence, better zero-shot performance



# Evaluation: The Link Prediction Task

- ▶ **Objective:** Predict missing treatments for specific farm conditions
- ▶ **Dataset:** Salmon-Lice-Treatment KG (30k+ triples)
- ▶ **Key Metrics:**
  - ▶ **MRR:** Mean Reciprocal Rank (How high is the correct treatment ranked?)
  - ▶ **Hits@10:** Is the correct treatment in the Top 10 recommendations?

*Testing the models against real-world asymmetric constraints*

## Results: The Model Leaderboard

Model	Hits@5 ( $\uparrow$ )	Hits@10 ( $\uparrow$ )	MRR ( $\uparrow$ )	MR ( $\downarrow$ )
<b>SSL-SSM (Ours)</b>	0.3380	0.4000	0.2636	597.8
TransE	0.3040	0.3780	0.2169	285.8
RotatE	<b>0.4160</b>	<b>0.4800</b>	<b>0.3078</b>	<b>259.2</b>
DistMult	0.2780	0.3720	0.2075	<b>223.9</b>
AutoSF	0.0140	0.0220	0.0128	4353.6
ComplEx	0.0100	0.0200	0.0127	4073.0
Transformer	0.0820	0.1380	0.0544	1570.7

Analysis:

- ▶ **SSL-SSM** outperforms translational and bilinear baselines in Hits metrics.
- ▶ **RotatE** shows strong performance in complex space modeling for this dataset.
- ▶ **AutoSF/ComplEx** struggle with current initialization, highlighting the need for SSL.

# Future Work: From Lab to Farm

- ▶ **Temporal Dynamics:** Tracking efficacy over seasons
- ▶ **Sensory Data:** Integrating water quality and sensor logs
- ▶ **Deployment:** Real-time recommendation app for farmers



Moving towards autonomous aquaculture management

# Questions?

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Thank you for supporting sustainable aquaculture!