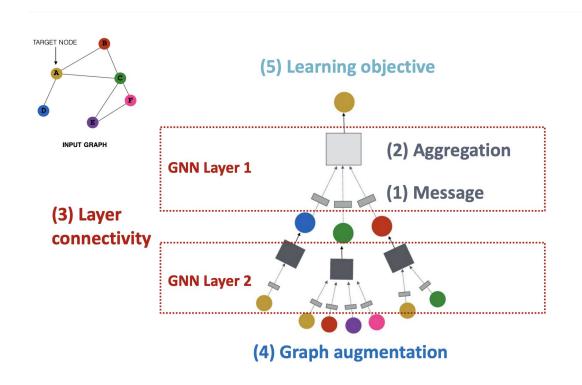
# Глубинное обучение в анализе графовых данных

7. Аугментации и обучение

в предыдущих сериях...

### Части GNN

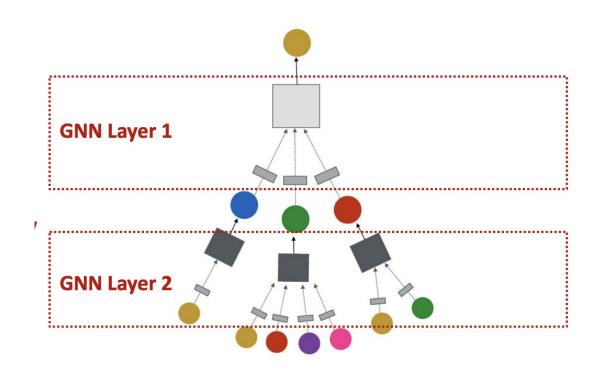


# Различные архитектуры

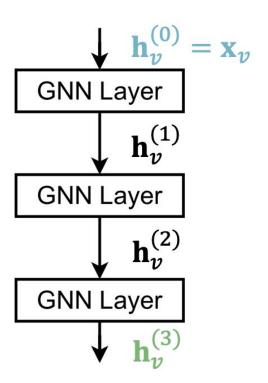
- GCN
- GraphSage
- GAT

# Связь слоев

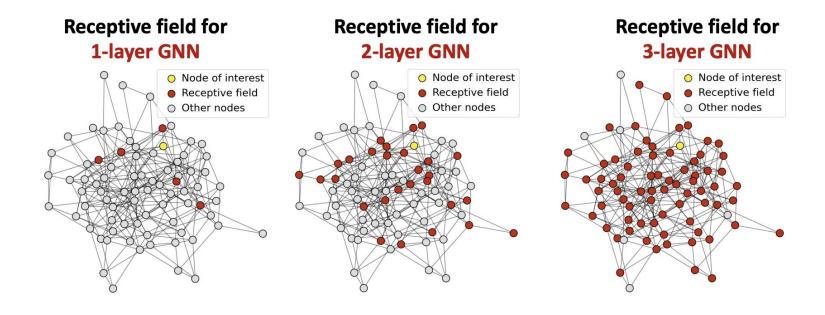
### Связь слоев



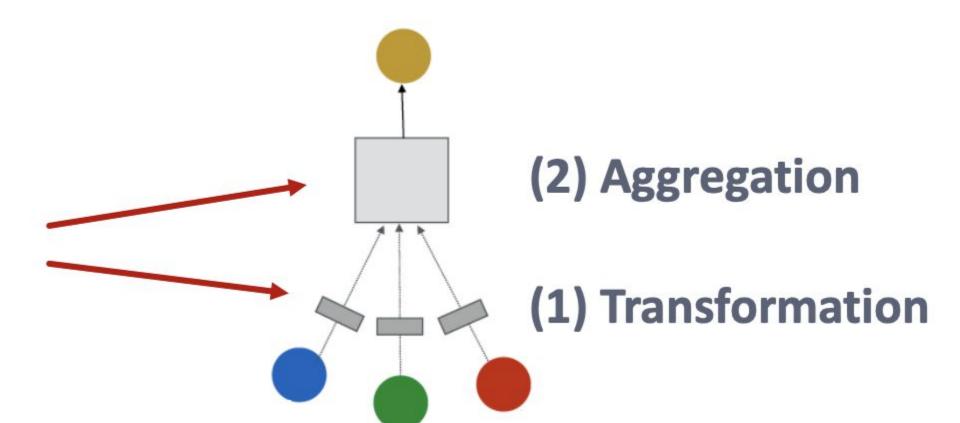
### Последовательно



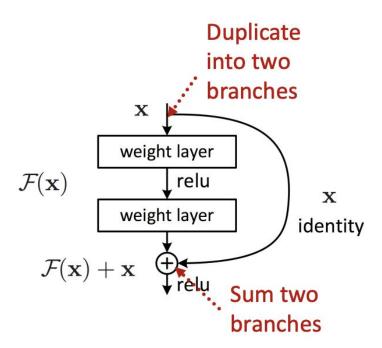
### Рецептивное поле



### Решение 1



# Skip-connections



## Пример

### A standard GCN layer

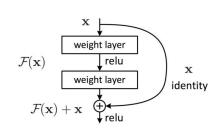
$$\mathbf{h}_{v}^{(l)} = \sigma\left(\sum_{u \in N(v)} \mathbf{W}^{(l)} \frac{\mathbf{h}_{u}^{(l-1)}}{|N(v)|}\right)$$

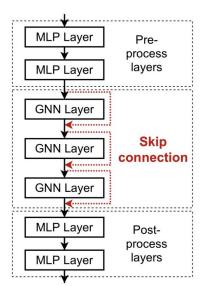
This is our F(x)

### A GCN layer with skip connection

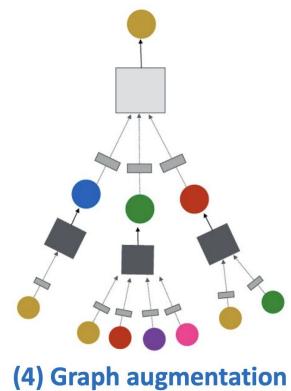
$$\mathbf{h}_{v}^{(l)} = \sigma \left( \sum_{u \in N(v)} \mathbf{W}^{(l)} \frac{\mathbf{h}_{u}^{(l-1)}}{|N(v)|} + \mathbf{h}_{v}^{(l-1)} \right)$$

$$F(\mathbf{x}) + \mathbf{x}$$





# Аугментации



# Проблемы

- нет фичей
- sparse
- dense
- large

### Решения

- нет фичей аугментации
- sparse добавить связей
- dense использовать сэмплирование
- large сэмплировать подграфы

# Нет фичей

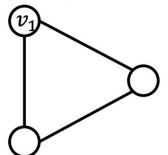
- 1) добавить всем константные фичи
- 2) добавить айдишники

# Сравнение

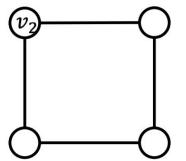
	Constant node feature	One-hot node feature
Expressive power	Medium. All the nodes are identical, but GNN can still learn from the graph structure	High. Each node has a unique ID, so node-specific information can be stored
Inductive learning (Generalize to unseen nodes)	High. Simple to generalize to new nodes: we assign constant feature to them, then apply our GNN	Low. Cannot generalize to new nodes: new nodes introduce new IDs, GNN doesn't know how to embed unseen IDs
Computational cost	Low. Only 1 dimensional feature	<b>High</b> . $O( V )$ dimensional feature, cannot apply to large graphs
Use cases	Any graph, inductive settings (generalize to new nodes)	Small graph, transductive settings (no new nodes)

# Циклы

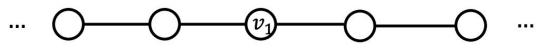
 $v_1$  resides in a cycle with length 3



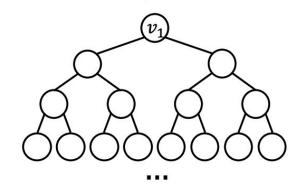
 $v_1$  resides in a cycle with length 4



 $v_1$  resides in a cycle with infinite length



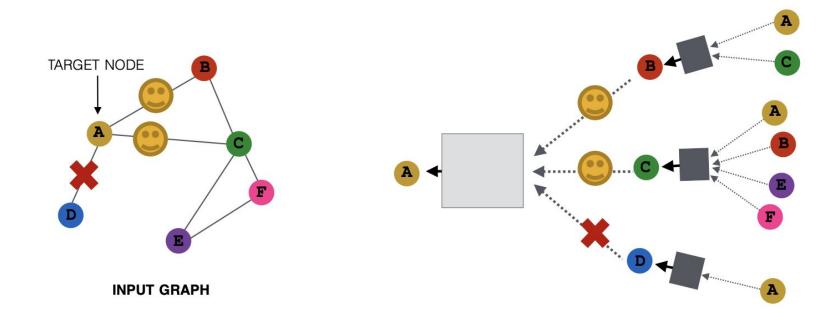
The computational graphs for node  $v_1$  are always the same



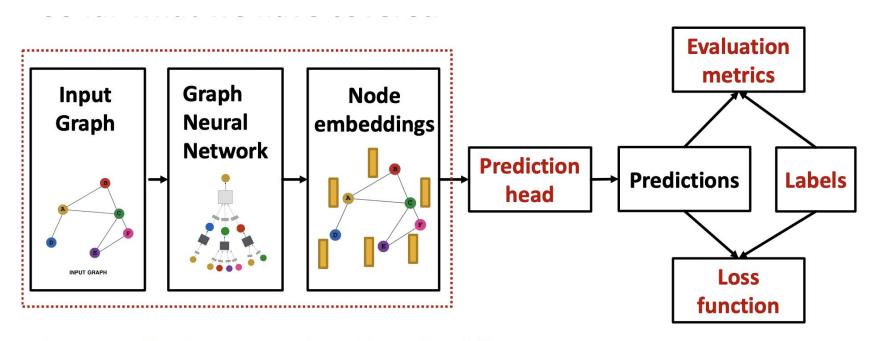
# Добавление вершин/ребер

• сокращение дистанции

# Сэмплирование соседей



# Обучение



**Output of a GNN: set of node embeddings** 

Иерархический global pooling

$$\{-1, -2, 0, 1, 2\}$$
  
 $\{-10, -20, 0, 10, 20\}$ 

# **Splitting**

