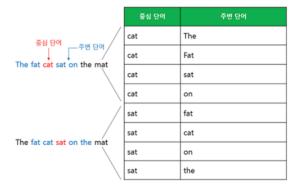
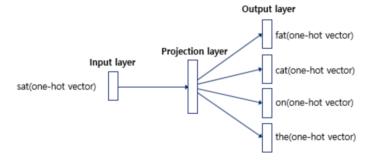


Skit-gram





Skit-gram _{학습과정}



Skip-Gram with Negative Sampling

Negative Sampling

입력과 레이블의 변화

입력1	입력2	레이블
cat	The	1
cat	fat	1
cat	sat	1
cat	on	1



입력1	입력2	레이블
cat	The	1
cat	fat	1
cat	pizza	0
cat	computer	0
cat	sat	1
cat	on	1

집한에서 랜덤으로



입력 레이블 중심 단어 ──주변 단어 cat The fat cat The fat cat sat on the mat cat sat cat on fat sat sat cat The fat cat sat on the mat sat on the sat ...

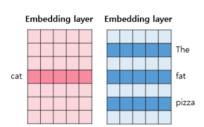
입력과 레이블의 변화

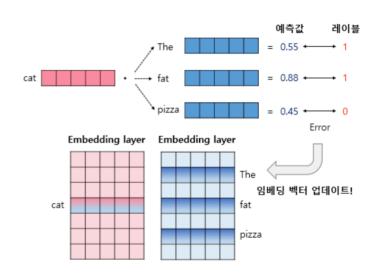
입력1	입력2	레이블
cat	The	1
cat	fat	1
cat	sat	1
cat	on	1
sat	fat	1
sat	cat	1
sat	on	1
sat	the	1

입력1	입력2	레이블
cat	The	1
cat	fat	1
cat	pizza	0
cat	computer	0
cat	sat	1
cat	on	1
cat	cute	1
cat	mighty	0

tongues(3115), tongues(3115) -> 1
church(352), armentrout(25793) -> 0

네거티브 샘플링 학습과정





```
from tensorflow.keras.models import Sequential, Model
from tensorflow.keras.layers import Embedding, Reshape, Activation, Input, Dot
from tensorflow.keras.utils import plot_model

embed_size = 50

def word2vec():
    target_inputs = Input(shape=(1,), dtype='int32')
    target_embedding = Embedding(vocab_size, embed_size)(target_inputs)

context inputs = Input(shape=(1,), dtype='int32')
```

context_embedding = Embedding(vocab_size, embed_size)(context_inputs)
dot_product = Dot(axes=2)([target_embedding, context_embedding])
dot product = Reshape((1,), input shape=(1,1))(dot product)

model = Model(inputs=[target inputs, context inputs], outputs=output)

model.compile(loss='binary crossentropy', optimizer='adam')

output = Activation('sigmoid')(dot product)

return model

```
for epoch in range(1,11):
       loss = 0
        for _, elem in enumerate(skip grams):
            first elem = np.array(list(zip(*elem[0]))[0], dtype='int32')
            second elem = np.array(list(zip(*elem[0]))[1], dtype='int32')
           labels = np.array(elem[1], dtype='int32')
            X = [first elem, second elem] # 학습데이터
           Y = labels # 정답
```

```
loss += model.train on batch(X,Y) # 한번의 epoch에 여러 번 트레이닝
print('Epoch : ', epoch, 'Loss : ', loss)
```

□ Epoch : 1 Loss : 1230.8024683147669 Epoch: 2 Loss: 913.956863515079 Epoch: 3 Loss: 804.8944275230169 Epoch: 4 Loss: 752.8729493767023 Epoch: 5 Loss: 721.9261713698506 Epoch: 6 Loss: 698.8518658801913 Epoch: 7 Loss: 677.1577545925975 Epoch: 8 Loss: 653.6212713383138 Epoch: 9 Loss: 626.5351788364351 Epoch: 10 Loss: 595.3925204891711