## Result Text Detection

## **Detected Texts**

deeplearningbook.org/contents/rep https://www.deeplearningbook.org/contents/representation.h 120% Greedy layer-wise unsupervised pretraining relies on a sing tation learning algorithm such as an RBM, a single-layer au coding model, or another model that learns latent represent pretrained using unsupervised learning, taking the output o and producing as output a new representation of the data, w its relation to other variables, such as categories to predic See algorithm 15.1 for a formal description. Greedy layer-wise training procedures based on unsupervise been used to sidestep the difficulty of deeplearningbook.org/contents/rep https://www.deeplearningbook.org/contents/representation.h 120% Greedy layer-wise unsupervised pretraining relies o n a

single-layer

```
represen-
tation
learning
algorithm
such
a s
a n
RBM,
a
single-layer
autoencoder,
a
sparse
coding
model,
o r
another
model
that
learns
latent
representations.
Each
layer
is
pretrained
using
unsupervised
learning,
taking
t h e
output
o f
```

```
t h e
previous
layer
a n d
producing
a s
output
a
n e w
representation
o f
t h e
data,
whose
distribution
(or
its
relation
t o
other
variables,
such
a s
categories
t o
predict)
is
hopefully
simpler.
See
algorithm
15.1
for
```

```
a
formal
description.
Greedy
layer-wise
training
procedures
based
o n
unsupervised
criteria
have
long
been
used
t o
sidestep
t h e
difficulty
o f
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
Greedy layer-wise unsupervised pretraining relies on a sing
```

tation learning algorithm such as an RBM, a single-layer autoding model, or another model that learns latent represent pretrained using unsupervised learning, taking the output oand producing as output a new representation of the data, wits relation to other variables, such as categories to predic See algorithm 15.1 for a formal description.

```
Greedy layer-wise training procedures based on unsupervise
been used to sidestep the difficulty of
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
Greedy
layer-wise
unsupervised
pretraining
relies
o n
a
single-layer
represen-
tation
learning
algorithm
such
a s
a n
RBM,
single-layer
autoencoder,
sparse
coding
model,
o r
another
```

```
model
that
learns
latent
representations.
Each
layer
is
pretrained
using
unsupervised
learning,
taking
t h e
output
o f
t h e
previous
layer
a n d
producing
a s
output
n e w
representation
o f
t h e
data,
whose
distribution
(or
its
```

```
relation
t o
other
variables,
such
a s
categories
t o
predict)
is
hopefully
simpler.
See
algorithm
15.1
for
a
formal
description.
Greedy
layer-wise
training
procedures
based
o n
unsupervised
criteria
have
long
been
used
t o
sidestep
```

```
t h e
difficulty
o f
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
Greedy layer-wise unsupervised pretraining relies on a sing
tation learning algorithm such as an RBM, a single-layer au
coding model, or another model that learns latent represent
pretrained using unsupervised learning, taking the output o
and producing as output a new representation of the data, w
its relation to other variables, such as categories to predic
See algorithm 15.1 for a formal description.
Greedy layer-wise training procedures based on unsupervise
been used to sidestep the difficulty of
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
Greedy
layer-wise
unsupervised
pretraining
relies
o n
a
single-layer
```

represen-

```
tation
learning
algorithm
such
a s
a n
RBM,
single-layer
autoencoder,
a
sparse
coding
model,
o r
another
model
that
learns
latent
representations.
Each
layer
is
pretrained
using
unsupervised
learning,
taking
t h e
output
o f
t h e
```

```
previous
layer
a n d
producing
a s
output
n e w
representation
o f
t h e
data,
whose
distribution
(or
its
relation
t o
other
variables,
such
a s
categories
t o
predict)
is
hopefully
simpler.
See
algorithm
15.1
for
a
```

```
description.
Greedy
layer-wise
training
procedures
based
o n
unsupervised
criteria
have
long
been
used
t o
sidestep
t h e
difficulty
o f
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
Greedy layer-wise unsupervised pretraining relies on a sing
tation learning algorithm such as an RBM, a single-layer au
```

formal

tation learning algorithm such as an RBM, a single-layer autoding model, or another model that learns latent represent pretrained using unsupervised learning, taking the output oand producing as output a new representation of the data, with the relation to other variables, such as categories to prediculate algorithm 15.1 for a formal description.

Greedy layer-wise training procedures based on unsupervise been used to sidestep the difficulty of jointly

```
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
Greedy
layer-wise
unsupervised
pretraining
relies
o n
a
single-layer
represen-
tation
learning
algorithm
such
a s
a n
RBM,
single-layer
autoencoder,
sparse
coding
model,
o r
another
model
that
learns
```

```
latent
representations.
Each
layer
is
pretrained
using
unsupervised
learning,
taking
t h e
output
o f
t h e
previous
layer
a n d
producing
a s
output
a
n e w
representation
o f
t h e
data,
whose
distribution
(or
its
relation
t o
other
```

```
variables,
such
a s
categories
t o
predict)
is
hopefully
simpler.
See
algorithm
15.1
for
а
formal
description.
Greedy
layer-wise
training
procedures
based
o n
unsupervised
criteria
have
long
been
used
t o
sidestep
t h e
difficulty
o f
```

```
jointly
deeplearningbook.org/contents/rep
120%
that greedy layer-wise pretraining is not required to train f
architectures, but the unsupervised pretraining approach wa
succeed.
Greedy layer-wise pretraining is called greedy because it is
CHAPTER 15. REPRESENTATION LEARNING
rithm, meaning that it optimizes each piece of the solution
piece at a time, rather than jointly optimizing all pieces. It
because these independent pieces are the layers of the netw
greedy
deeplearningbook.org/contents/rep
120%
that
greedy
layer-wise
pretraining
is
n o t
required
t o
train
fully
connected
deep
```

```
architectures,
but
t h e
unsupervised
pretraining
approach
was
t h e
first
m e t h o d
t o
succeed.
Greedy
layer-wise
pretraining
is
called
greedy
because
i t
is
a
greedy
algo-
526
CHAPTER
15.
REPRESENTATION
LEARNING
rithm,
meaning
that
i t
```

```
optimizes
each
piece
o f
t h e
solution
independently,
o n e
piece
a t
a
time,
rather
than
jointly
optimizin g
all
pieces.
Ιt
is
called
layer-wise
because
these
in dependent
pieces
are
t h e
layers
o f
t h e
network.
Specifically,
```

```
greedy
deeplearningbook.org/contents/rep
120%
Greedy layer-wise pretraining is called greedy because it is
526
CHAPTER 15. REPRESENTATION LEARNING
rithm, meaning that it optimizes each piece of the solution
piece at a time, rather than jointly optimizing all pieces. It
because these independent pieces are the layers of the netv
greedy
layer-wise pretraining proceeds one layer at a time, training
keeping the previous ones fixed. In particular, the lower lay
first) are not adapted after the upper layers are introduced.
deeplearningbook.org/contents/rep
120%
Greedy
layer-wise
pretraining
i s
called
greedy
because
i t
is
a
greedy
algo-
526
CHAPTER
```

```
15.
REPRESENTATION
LEARNING
rithm,
meaning
that
i t
optimizes
each
piece
o f
t h e
solution
independently,
o n e
piece
a t
a
time,
rather
than
jointly
optimizing
all
pieces.
l t
is
called
layer-wise
because
these
in dependent
pieces
```

```
are
t h e
layers
o f
t h e
network.
Specifically,
greedy
layer-wise
pretraining
proceeds
o n e
layer
a t
a
time,
training
t h e
k-th
layer
while
keeping
t h e
previous
o n e s
fixed.
l n
particular,
t h e
lower
layers
(which
are
```

```
trained
first)
are
n o t
adapted
after
t h e
upper
layers
are
introduced.
is
called
unsuper-
deeplearningbook.org/contents/rep
120%
Greedy layer-wise pretraining is called greedy because it is
526
CHAPTER 15. REPRESENTATION LEARNING
rithm, meaning that it optimizes each piece of the solution
piece at a time, rather than jointly optimizing all pieces. It
because these independent pieces are the layers of the netw
greedy
layer-wise pretraining proceeds one layer at a time, training
keeping the previous ones fixed. In particular, the lower lay
```

first) are not adapted after the upper layers are introduced.

deeplearningbook.org/contents/rep

+ 1 2 0 %

+

```
Greedy
layer-wise
pretraining
is
called
greedy
because
i t
is
a
greedy
algo-
526
CHAPTER
15.
REPRESENTATION
LEARNING
rithm,
meaning
that
i t
optimizes
each
piece
o f
t h e
solution
independently,
o n e
piece
a t
a
time,
```

```
rather
than
jointly
optimizing
all
pieces.
Ιt
is
called
layer-wise
because
these
independent
pieces
are
t h e
layers
o f
t h e
network.
Specifically,
greedy
layer-wise
pretraining
proceeds
o n e
layer
a t
a
time,
training
t h e
k-th
```

```
layer
while
keeping
t h e
previous
o n e s
fixed.
l n
particular,
t h e
lower
layers
(which
are
trained
first)
are
n o t
adapted
after
t h e
upper
layers
are
introduced.
l t
is
called
unsuper-
deeplearningbook.org/contents/rep
+
120%
```

## CHAPTER 15. REPRESENTATION LEARNING

rithm, meaning that it optimizes each piece of the solution piece at a time, rather than jointly optimizing all pieces. It because these independent pieces are the layers of the network greedy

layer-wise pretraining proceeds one layer at a time, training keeping the previous ones fixed. In particular, the lower lay first) are not adapted after the upper layers are introduced. vised because each layer is trained with an unsupervised realgorithm. However, it

deeplearningbook.org/contents/rep

+

120%

+

CHAPTER

15.

REPRESENTATION

LEARNING

rithm,

meaning

that

i t

optimizes

each

piece

o f

t h e

solution

independently,

o n e

piece

a t

a

```
time,
rather
than
jointly
optimizing
all
pieces.
Ιt
is
called
layer-wise
because
these
in dependent
pieces
are
t h e
layers
o f
t h e
network.
Specifically,
greedy
layer-wise
pretraining
proceeds
o n e
layer
a t
a
time,
training
t h e
```

```
k-th
layer
while
keeping
t h e
previous
o n e s
fixed.
l n
particular,
t h e
lower
layers
(which
are
trained
first)
are
not
adapted
after
t h e
upper
layers
are
introduced.
Ιt
is
called
unsuper-
vised
because
each
```

```
layer
is
trained
with
a n
unsupervised
representation
learning
algorithm.
However,
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
CHAPTER 15. LEARNING
rithm, meaning that it optimizes each piece of the solution
piece at a time, rather than jointly optimizing all pieces. It
because these independent pieces are the layers of the netw
greedy
layer-wise pretraining proceeds one layer at a time, training
keeping the previous ones fixed. In particular, the lower lay
first) are not adapted after the upper layers are introduced.
vised because each layer is trained with an unsupervised re
algorithm. However, it
deeplearningbook.org/contents/rep
https://www.deeplearningbook.org/contents/representation.h
120%
CHAPTER
```

15.

```
LEARNING
rithm,
meaning
that
i t
optimizes
each
piece
o f
t h e
solution
independently,
o n e
piece
a t
a
time,
rather
than
jointly
optimizing
all
pieces.
Ιt
is
called
layer-wise
because
these
in dependent
pieces
are
t h e
```

```
layers
o f
t h e
network.
Specifically,
greedy
layer-wise
pretraining
proceeds
o n e
layer
a t
a
time,
training
t h e
k-th
layer
while
keeping
t h e
previous
o n e s
fixed.
l n
particular,
t h e
lower
layers
(which
are
trained
first)
```

```
are
n o t
adapted
after
t h e
upper
layers
are
introduced.
Ιt
is
called
unsuper-
vised
because
each
layer
```

is

trained