

The logo for TI Developer Conference, featuring the letters 'TI' in a stylized font followed by the words 'Developer Conference' in a bold, sans-serif font.

TI Developer Conference

February 28-March 2, 2008 • Dallas, TX

Silhouettes of three people standing and talking, with a large white curved line passing through them. The background is a green circuit board pattern with various icons like a printer, a car, and a laptop.

DLP Driven, Learning, Optical Neural Networks

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SPRP506

Technology for Innovators™

 **TEXAS INSTRUMENTS**

Neural Network Applications

- ◆ **Stock Prediction:** Currency, Bonds, S&P 500, Natural Gas
- ◆ **Business:** Direct mail, Credit Scoring, Appraisal, Summoning Juries
- ◆ **Medical:** Breast Cancer, Heart Attack Diagnosis, ER Test Ordering
- ◆ **Sports:** Horse and Dog Racing
- ◆ **Science:** Solar Flares, Protein Sequencing, Mosquito ID, Weather
- ◆ **Manufacturing:** Welding Quality, Plastics or Concrete Testing
- ◆ **Pattern Recognition:** Speech, Article Class., Chem. Drawings
- ◆ **No Optical Applications:** Starting with Boolean

Most from www.calsci.com/Applications.html

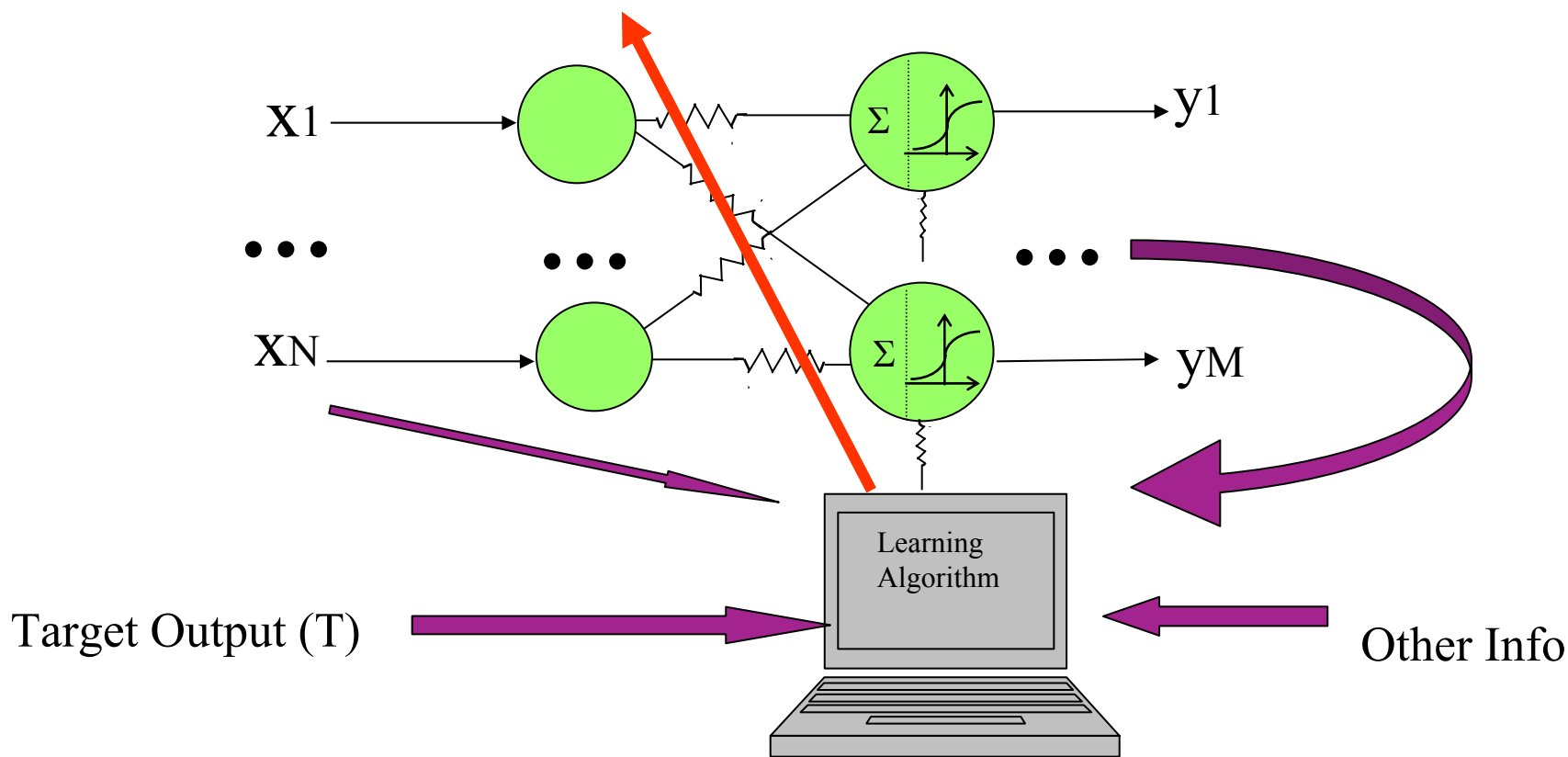
- ◆ **Optical Parallel Processing Gives Speed**
 - Lenslet's Enlight 256—8 Giga Multiply and Accumulate per second
 - Order 10^{11} connections per second possible with holographic attenuators
- ◆ **Neural Networks**
 - Parallel versus Serial
 - Learn versus Program
 - Solutions beyond Programming
 - Deal with Ambiguous Inputs
 - Solve Non-Linear problems
 - Thinking versus Constrained Results

Optical Neural Networks

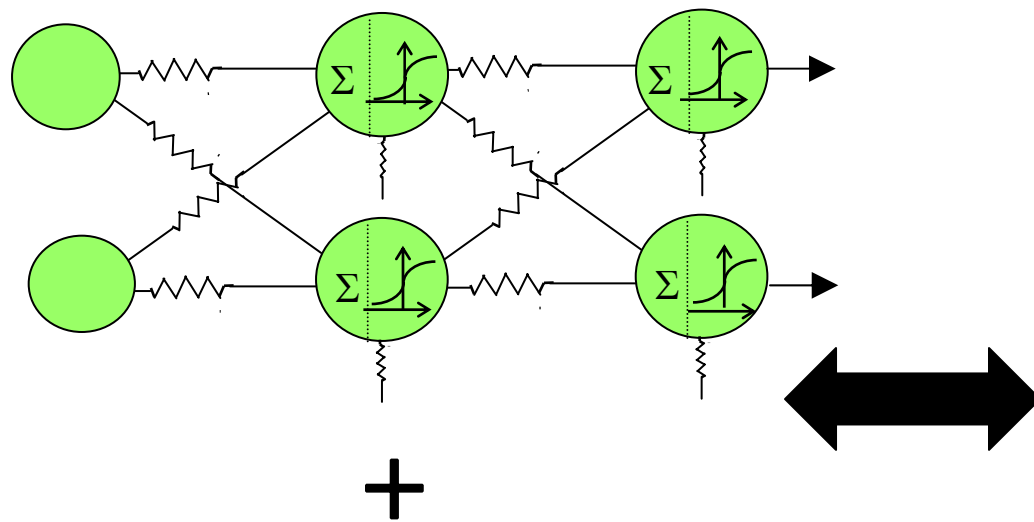
- ◆ **Sources are modulated light beams (pulse or amplitude)**
- ◆ **Synaptic Multiplications are due to attenuation of light passing through an optical medium**
- ◆ **Geometric or Holographic**
- ◆ **Target neurons sum signals from many source neurons.**
- ◆ **Squashing by operational-amps or nonlinear optics**

Standard Neural Net Learning

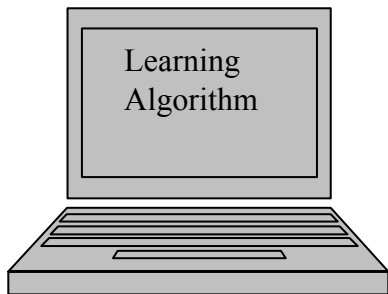
- ◆ We use a Training or Learning algorithm to adjust the weights, usually in an iterative manner.



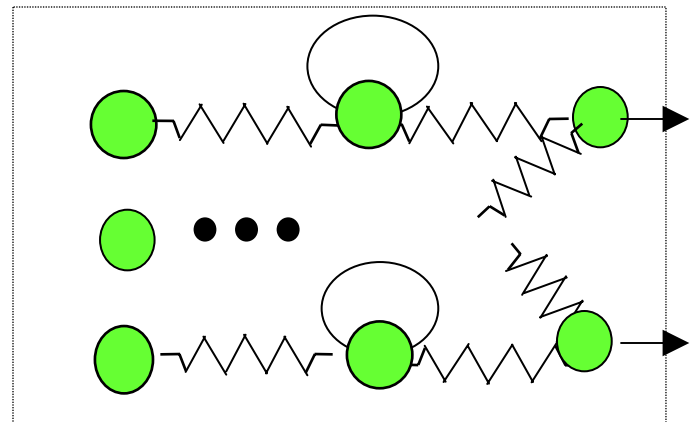
FWL-NN Is Equivalent to a Standard Neural Network + Learning Algorithm



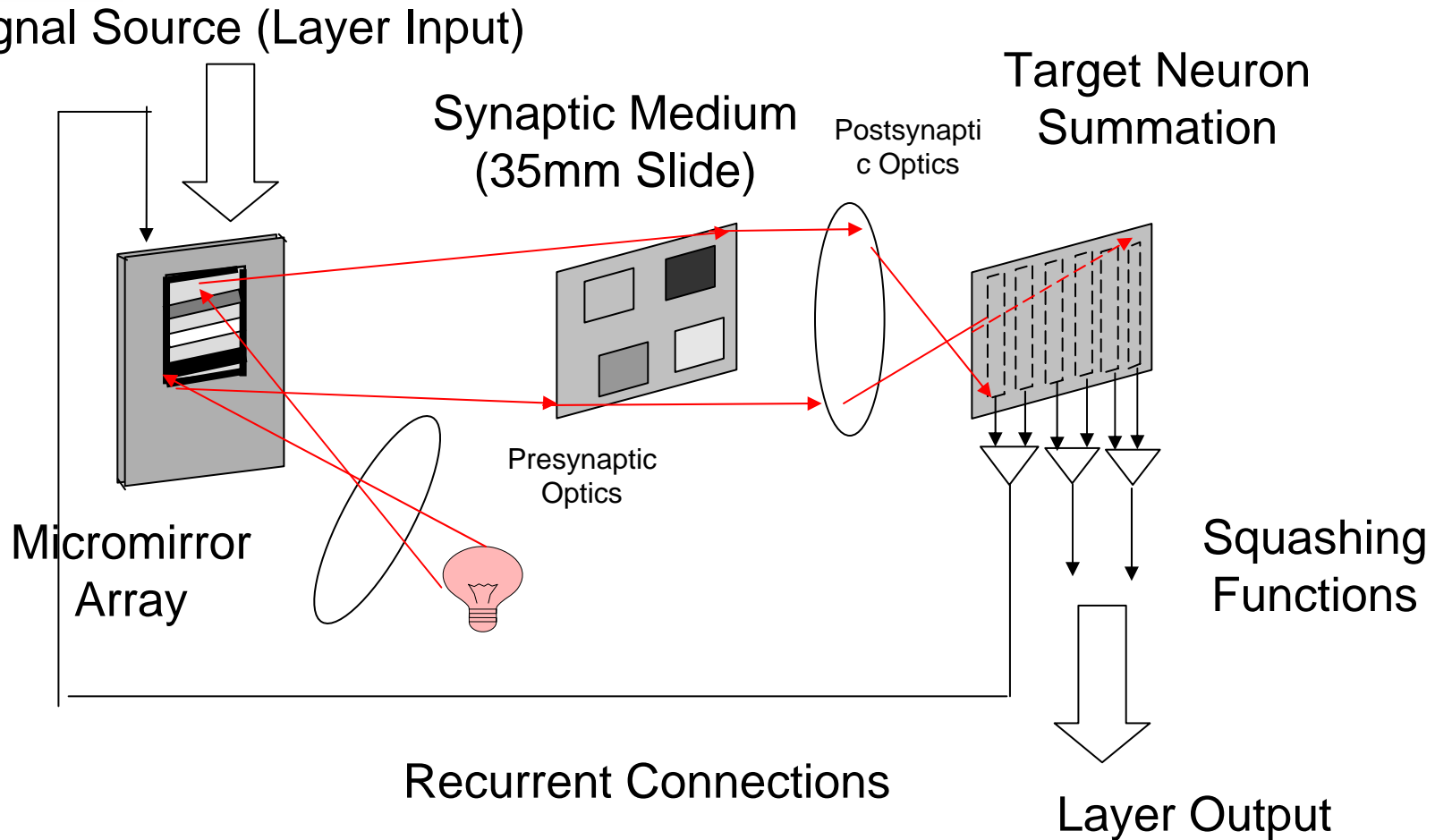
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FWL-NN

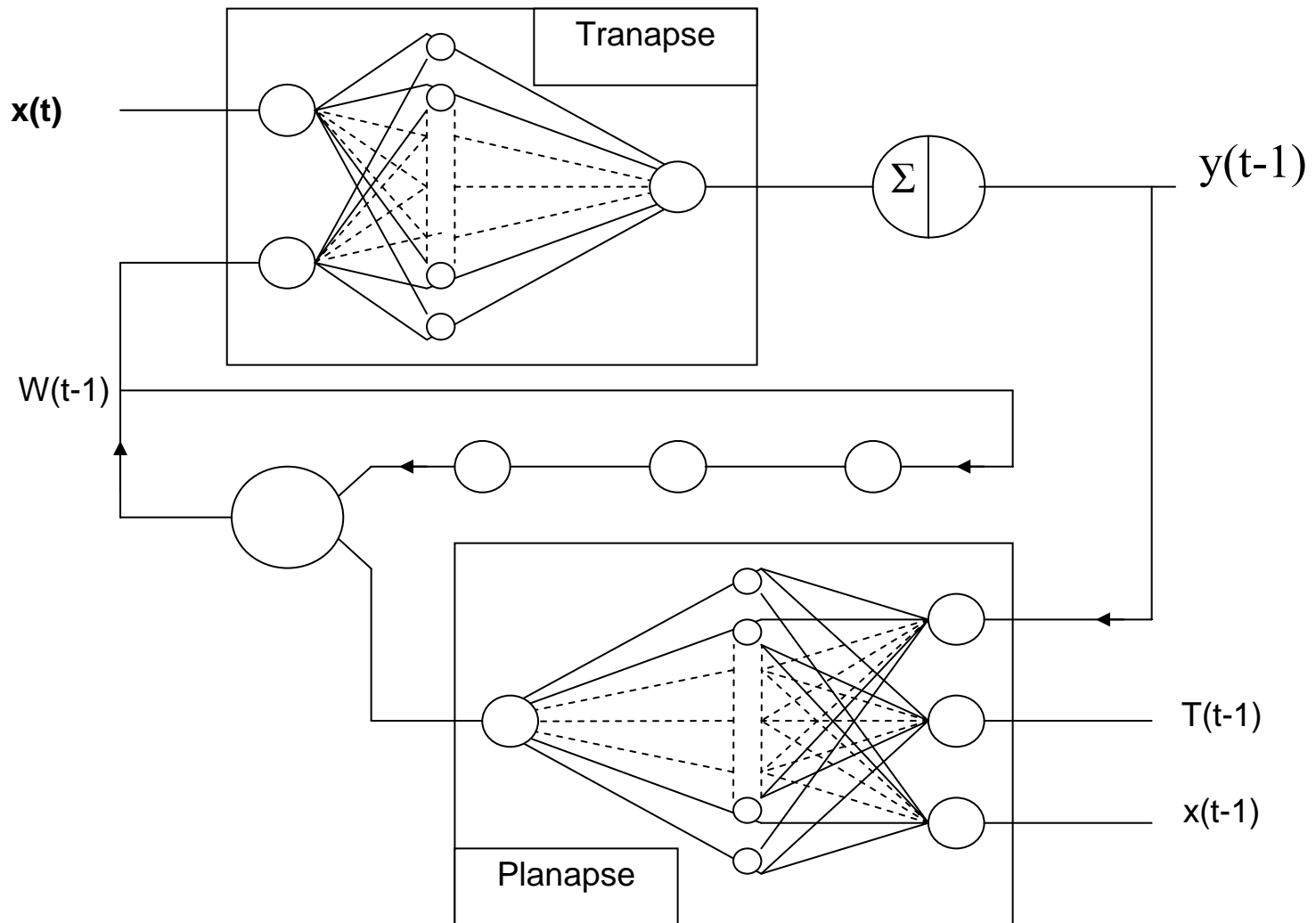


Optical Recurrent Neural Network



A Single Layer of an Optical Recurrent Neural Network. Only four synapses are shown. Actual networks will have a large number of synapses. A multi-layer network has several consecutive layers.

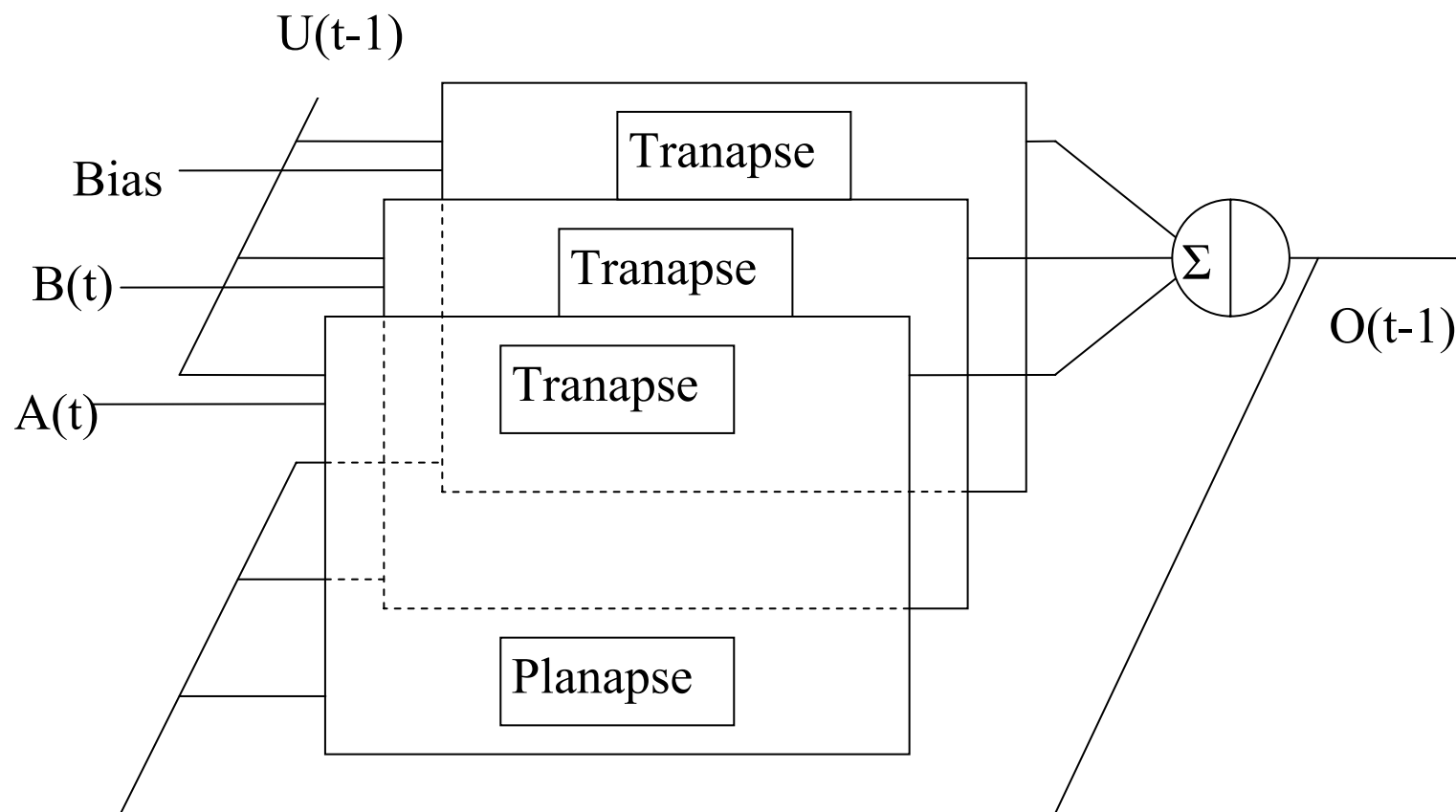
Optical Fixed-Weight Learning Synapse



Definitions

- ◆ **Fixed-Weight Learning Neural Network (FWL-NN)**
 - A recurrent network that learns without changing synaptic weights
- ◆ **Potency – A weight signal**
- ◆ **Tranapse – A Potency modulated synapse**
- ◆ **Planapse – Supplies Potency error signal**
- ◆ **Recurron – A recurrent neuron**
- ◆ **Recurral Network – A network of Recurrons**

Page Representation of a Recurron



Optical Neural Network Constraints

- ◆ **Finite Range Unipolar Signals [0,+1]**
- ◆ **Finite Range Bipolar Attenuation[-1,+1]**
- ◆ **Excitatory/Inhibitory handled separately**
- ◆ **Limited Resolution Signal**
- ◆ **Limited Resolution Synaptic Weights**
- ◆ **Alignment and Calibration Issues**

Proposed Design

- ◆ **Digital Micromirror Device**
- ◆ **35 mm slide Synaptic Media**
- ◆ **CCD Camera**
- ◆ **Synaptic Weights - Positionally Encoded
- Digital Attenuation**
- ◆ **Planapse, Tranapse trained using
Standard Backpropagation**
- ◆ **Allows flexibility for evaluation.**

◆ Demonstration

- Left Images (time order)
 - Gray-scale Source Neuron Signal
 - Equivalent Pulse Modulated Source Neuron Signal
 - Gray-scale Source Neuron Signal – Reprise
 - Blank during Target Summation and Squashing
- Middle Image – Fixed Weights on Attenuator Slide (static)
- Right Images (time order)
 - Blank during First Source Gray-scale
 - Accumulating (Integrating Pulses) Target Neuron Input
 - Gray-scale during Summation and Squashing
- Phases and Layers
 - Synapses Connect Two Layers (N and N+1)
 - Only Source Neurons (layer N) Illuminate
 - Only Target Neurons (layer N+1) Accumulate

Hard and Soft Optical Alignment

- ◆ **Hardware Alignment**
 - Best for Gross Alignment
 - Precision Stages Required for Precise Alignment
 - Hard to Correct for Distortions in Optics
- ◆ **Software Alignment**
 - Fine Image Displacements
 - Image Rotation
 - Integrating Regions of Arbitrary Shape and Size
 - Diffraction Effects Filtered and Ignored
 - Correct for Optical Path Changes

Photonic Circuit Elements

◆ NEAR

- Optical Fiber
- Fiber Amplifiers
- Splitters

◆ MEDIUM

- Cylindrical Lenses before and after Synaptic Media (1-D systems, sparse media)
- Non-Linear Crystals for Squashing

◆ FAR

- Integrated Emitters, Attenuator, Detectors/Limiters
- Holographic Attenuators (destructive interference)
- Stackable



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