| Parage Data Andys: 5 - IDA |
|---|
| Lossan One |
| In remove, sensing exergy radiating from the earth's surface is measured using sensors Minted and Jan aircraft, satelietic, or other sixborne object. |
| The measurements orallow one to construct an image of the land scape/carth's Sulface |
| Basic Diagram Transmitter Sentor Transmitter January Transmitter January Transmitter |
| Seafor Tristrumentation of Tronsmitter |
| 17 11/2 b 1 1 1 to 00 YALA |
| Signal processing Data in mage bornet |
| 1//////// Hap. Mag. |
| |
| The energy collected by the Sensor can come from a number of sourced, giving rise to two types of sonsor: |
| - Energy variated from the earth as a result of) the son's radiation Energy radiated from the earth as a result of Passive sensing |
| - Evergy sadjected from the earth as a versult of Passive sensing . it's own temperature (Bluebody) - Energy sadjected due to illumination by an } Active sensing . artifical source (loser/rador/etc.) . Active sensing . digital format. |
| The most important parameter involved in remote sensing is, ultimatedy, the wavelength of the radiation. |
| For Passive sensors: 1 = 0.4 µm -> 12 µm (visible/Ive) |
| l = 30n -> 300mm (Microunve) |
| |
| For active sensors: Excitation, 1 = 355 nm (= UV) Measurement, 2 = 0.4 mm -> 12 mm (visible/IR) } Laser Floro sensors |
| SAR (Synthetic Aperluse Madar) } \(\lambda = \{ Ka, K, Ku, X, C, S, L. Bands \} \) /SLAR (Side Looking Airbourne Radar) \(\lambda = \{ Radar Sensors} \) |
| |

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Resson One-Introduction to Data Sources in Kemore Introduction to Data Sources in Komore sensing Intro Po Data Sources.
Characteristics of Digital Image Data.
Remote Sensing Data Products and Processing
Wavelengths of Totorest, Active, and Passive Suprems
Technical Characteristics of Digital Image Data.
Technical Characteristics of Digital Image Data. Eq. Landsot 7 Darports (1999-Present) Eq. Lanset 5 Images, (1984-2013) Spectral Ranges Commonly used in Kernote Sensing Radiction Engry (som Differtly) Hot bodies /Black Bodies

E.g.: OLI and TIKS Sonsors on Mondent 8. (2013-Present) of Landset 9. (2021

Electromagnetic Rediction Scattering Annotation

Note Sensing Platforms. Electromagnetic Rediction Scattering Remote Sensing Platforms Different kinds of platforms and Applications Sur-synchronous Low Earth Orbits. Satellite and Aircraft Imaging Technologies. Spectral Kesolution: Convential vs Spectrometric Imag Image Data Sources in the Micromove Region Side Looking Airbourne Madlar - SLAR. Synthetic Aperture Radar - SAR. Spatial Data Sources in General Types of Spatial Data Data Formats . Use of Colour in Remote Sensing. Images . Multispectial Roster Image Formats. LESSON TWO - Error Correction and Registration of Image Error Correction and Registration of Image Data Image Data Acquisition Sources of Error in the (data driven) I mage Analysis Pipeline Sources of Error in the Remote Sensing Data Radiometric Distortion in Remote Sensing Types of Radiometric Distoction The Effect of Atmosphere on Radiation.

The Effect of Atmosphere on Radiation.
Radiometric Quantity Definitions
Absorption and Scattering by the Atmosphere.
The Effect of Atmosphere on Radiation Cont.
Further Radiometric Quantity Definitions
Atmospheric Effects on Remote Sensing Imagery.
In strumentation Errors
Striping (or Barding) Artifacts (Aling Swith Direction)

Radiometric Distortion Correction of Almispheric Effects
Corrections of Instrumentation Errors
Sources of Exponetric Distortion in Remote Sensing Sources of Geologye Distoction Sources of Geologic Distortion

Diasic Image Formation Geometry.

Earth Patation Effects A

Panoranic Distortion Distortion

Earth Curvature

Scan Time Skew

Variations in Platform Altitude, Whicity, and Attitude

Aspect Ratio Distortion

Sensor Scan Mon-Linearities

Non-Systemic and Systematic Distortion

Distortion Correction of Geometric Distortion CTEMERAL Aspects USE of Mapping Polynomials for I mage correction Mapping Polynomials and Ground Control Points Resampling Interpolation. Choice of Control Points Eq. : Map- Lankat MSS Registica Mathematical modelling . Aspect . Ratio. Correction Earth Notation Skew Correction . Image . Orientation to North-South . Correction of Panoramic Effects. Combining the corrections

I mage Registration

Geo. Ceferencing and Geo. Coding. Image to Image Registration.

. Example of Im. - Im Registration (dutomated Feature Based)

. Incop. to I mage (Affine) Registration in the Frequency Domain. Control Point Registration las Automated Feature Extraction and Matching

LASSON Three - Multispectral Transformations of Image Data Latiodiction

The Principle Components Transformation

The Man Vector and Covariance Matrix

An example

A zero Correlation, Retational Transform

An example (cont...) An example ((on...)

Practical Considerations 1,70% A Second Keal Franch A Second Keal Example Conformatic Mapper
Remorks about the Principle Components Transform. The Effects of an Origin Shift & Applications of Principal Components in Image Enhancement and Display. The Taylor Method of Contrast Enhancement (Decorrelation Stretch). Decorrelation Stretch Example One Decorrelation Stretch Example Two
Other Applications of Principle Components Analysis. Lecture Four - The Interpretation of Digital Image Data Apploaches to Interpretation. The Classic View of Complementarity Pinel-Wise Classification . Pixel- Wise Classification in the Pattern Space Multi-spectral Pattern Space: Information vs Spectral

Object-wise Detection / Classification The Current "Interpretability". Perspective Definition and Types of Machine Learning. Machine Learning Overview. . Applications of Machine Learning . .

Classification and Beyond. I'm age. Interpretation Domains: Beyond Remote Sensing . Un Supervised Classification

. Imaging and Beyond

Supervised Classification. Steps in Supervised Classification Measuring the Success and Performance for Classificati Typical Issues in Machine Learning.

Overfitting and Underfitting Generalisation and Capacity Model Capacity Bias and Variance. Overall Picture.

Supervised (Statistical) Classification Techniques Maximon Likelihood Classification Bayes' Classification
The Maximum's Likelihood Decision Rule
Multivariate Normal Class Models
Decision Suffaces
Thresholds
Number of Training Rhals Required for Each Class
A Simple Illustration In
Gaussian Mixture Models
Minimum Dixance and Parallel-Pikel Classification The Case of Limited Training Data Minimum Distance Classifier: The Discriminate Function

Minimum Distance Classification Cons Remarks on minimum Distance Classification Decision Suffaces - Example Remarks on Minimum Distance Classification Parallel-Piped Classification. Classification Time Comparison of the classifiers Context Classification The Concept of Spatial Context Context Classification by Image Pre-Processing Post Classification Filtering Probabilistic Label Relaxation. The Basic Algo Cithm. Determining the Compatibility Coefficients. The Final Step Stopping the Process Propagation Control Erangle one. Example Two Handling spatial Context by Markov Random Fields Comparative Example Lisbon Six - Supervisied Non-Parametric Classification: Geometric Approaches Introduction The KNN (Nearest Neighbour) Classifier Non-Parametric Methods from a Geometric Basis: Linear Discriminal Concept of a Weight Vector Testing Class Membership. . Training . a . Linear . Classifier . in the . weight . Space . . Setting the . Correct ion . Increment

Support Vector Classifiers Lineray Seperable Classes

Overlappition Classes - The Use of Slack Variables

Linear Engrand Functions

Example

Popular Kersels

Of

Binary in Multicategory Classification Binary Classification - The Pophershold Logic Unit
Mulli category Classification
Networks of Classifiers - Solutions of Non-Linear Problems
SVM Application: Kernel Selection Mand Multicategory Stategy
SVM Application: Examples
Networks of Classifiers: The Neural Retorne Approach The Hewal Metrorx Aproach The Processing Element.
Training the Newral Networks - Backpropagation
Chopsing the Network Parameters Committees of Classifiers: Ensemble Classification Problem Statement and Decision Logics Bagging. Aduboost. Committees of Classifiers: Mandow Forests The Idea of Ensembling Multistage Classifiers Decision Tree Classifiers. CART - Classification And Regression: Random Forests Lesson Seven - Christering and Unsupervised Classification Back to the General Problem of Delineation of Spectral Classes Similarity Metrics and Chostering Criteria Up many Clusters? - Similarity Metrics and Clustering Criteria The Iterative optimisation (migrating means, K-means) Clustering Algorithm The Basic Algorithm Tsodata - Margings, Deletiony, and Sphitting Elongated Chasters Choice of Instal Chaster Centres Clustering Co

Onspervised Classification and Cluster Maps - Examples Unsigning Example A Single Pesso, Clustering Technique Single Pass Algoration Adventages and Limiters Skip Generation Parameter Variations on the Single Bass Algorithm Example Agglomerative Hierarchical Clustering Municol Clustering by Histogram Peak selections, Digital Example. From Shallow to Deep Merral Methorics Main Resources - Related Courses Knowledge Base vs Representation Learning Food-Formard Neural Networks The Mannalian Visual Cortex is Hierarchical Multilayer Perception (MLP) Representation (Fedure) Learning Perspective of MLP Activation (Non-Linear) Functions, g() Training an MLP (and a Function) MLP Expressiveness Number of Newroxs Example One Expressiveness Example Two: Binary Case and Complexity Universal Approximation Theorems Approximation Capability (Expressiveness): The W; ath or Depth Dilemma Outputs and Loss Functions LOSS Function Task Dependency of Output and Loss Functions I mage Classification Binary Classification - Bernoulli Distribution - BCE Multi Class : Classification - Categorical Distribution - CE. Loss Multicless Classification - Categorical Distribution - Softmax Output Summary: Classification . Deep . Neural Networks Activation Functions Sigmoid Rectified Linear Unit (Re LU) Leaky ReLU. Exponential Linear Units (ELU)

Bosson Nine - Convolutional Neural Networks Interdig tion

Notationing

Fully Connected vs Convolutional Layers

Downsamplify

Pooling

Strided Convolution

Receptive Field Algerthm

Fully Connected Layers

Convolutional Network Extensible

Upsampling.

Max Unpooling

Dilated Convolution

Greidding Effects

Architetures

Some Remarks, about Chals

Lesson Ten - AI and Deep Leafning for Medical Image Malysis Intro-Main Issues/Challenges Data Driven Approach Evaluation Metrics Applications and Case Studies Medical Image Classification . Medical Image Sequentation and Detect Higher Dimensionality Data. Interpretability, fairness, Ethics





