## Perspective Projection Transformation

Where does a point of a scene appear in an image?


Transformation in 3 steps:

1. scene coordinates => camera coordinates
2. projection of camera coordinates into image plane
3. camera coordinates => image coordinates

Perspective projection equations are essential for Computer Graphics. For Image Understanding we will need the inverse: What are possible scene coordinates of a point visible in the image? This will follow later.

## Perspective Projection in Independent Coordinate Systems

It is often useful to describe real-world points, camera geometry and image points in separate coordinate systems. The formal description of projection involves transformations between these coordinate systems.


## 3D Coordinate Transformation (1)

The new coordinate system is specified by a translation and rotation with respect to the old coordinate system:

$\underline{v^{\prime}}=\mathbf{R}\left(\underline{v}-\underline{v}_{0}\right) \quad$| $\mathbf{v}_{0}$ is displacement vector |
| :--- |
| $\mathbf{R}$ | is rotation matrix $\quad$| Note that these matrices <br> describe coo transforms |
| :--- |
| for positive rotations of |
| the coo system. |

If rotations are performed in the above order:

1) $\gamma=$ rotation angle about $z$-axis
2) $\beta=$ rotation angle about (new) $y$-axis
3) $\alpha=$ rotation angle about (new) $x$-axis
("tilt angle", "pan angle", and "nick angle" for the camera coordinate

$$
\mathbf{R}_{\mathbf{y}}=\left[\begin{array}{ccc}
\cos \beta & 0 & -\sin \beta \\
0 & 1 & 0 \\
\sin \beta & 0 & \cos \beta
\end{array}\right]
$$

assignment shown before)

$$
\mathbf{R}_{\mathbf{z}}=\left[\begin{array}{ccc}
\cos \gamma & \sin \gamma & 0 \\
-\sin \gamma & \cos \gamma & 0 \\
0 & 0 & 1
\end{array}\right]
$$

## 3D Coordinate Transformation (2)

By multiplying the 3 matrices $\mathbf{R}_{\mathbf{x}}, \mathrm{R}_{\mathbf{y}}$ and $\mathrm{R}_{\mathbf{z}}$, one gets
$R=\left[\begin{array}{lcl}\cos \beta \cos \gamma & \cos \beta \sin \gamma & -\sin \beta \\ \sin \alpha \sin \beta \cos \gamma-\cos \alpha \sin \gamma & \sin \alpha \sin \beta \sin \gamma+\cos \alpha \cos \gamma & \sin \alpha \cos \beta \\ \cos \alpha \sin \beta \cos \gamma+\sin \alpha \sin \gamma & \cos \alpha \sin \beta \sin \gamma-\sin \alpha \cos \gamma & \cos \alpha \cos \beta\end{array}\right]$

For formula manipulations, one tries to avoid the trigonometric functions and takes
$\mathbf{R}=\left[\begin{array}{lll}\mathbf{r}_{11} & \mathbf{r}_{12} & \mathbf{r}_{13} \\ \mathbf{r}_{21} & \mathbf{r}_{22} & \mathbf{r}_{23} \\ \mathbf{r}_{31} & \mathbf{r}_{32} & \mathbf{r}_{33}\end{array}\right]$
Note that the coefficients of $\mathbf{R}$ are constrained:
A rotation matrix is orthonormal:
$R R^{\top}=I$ (unit matrix)

## Example for Coordinate Transformation


camera coo system:

- displacement by $\mathbf{v}_{0}$
- rotation by pan angle $\beta=-30^{\circ}$
- rotation by nick angle $\alpha=45^{\circ}$

$$
\underline{v}^{\prime}=R\left(\underline{v}-\underline{v}_{0}\right) \text { with } R=R x R y
$$

$$
\mathbf{R}_{\mathbf{x}}=\left[\begin{array}{ccc}
1 & 0 & 0 \\
0 & \frac{1}{2} \sqrt{2} & \frac{1}{2} \sqrt{2} \\
0 & -\frac{1}{2} \sqrt{2} & \frac{1}{2} \sqrt{2}
\end{array}\right] \quad \mathbf{R}_{\mathbf{y}}=\left[\begin{array}{ccc}
\frac{1}{2} \sqrt{3} & 0 & \frac{1}{2} \\
0 & 1 & 0 \\
-\frac{1}{2} & 0 & \frac{1}{2} \sqrt{3}
\end{array}\right]
$$

## Perspective Projection Geometry

Projective geometry relates the coordinates of a point in a scene to the coordinates of its projection onto an image plane.

Perspective projection is an adequate model for most cameras.


## Perspective and Orthographic Projection

Within the camera coordinate system the perspective projection of a scene point onto the image plane is described by
$x_{p}^{\prime}=\frac{x^{\prime} f}{z^{\prime}} \quad y_{p}^{\prime}=\frac{y^{\prime} f}{z^{\prime}} \quad z_{p}^{\prime}=f \quad(f=$ focal distance $)$

- nonlinear transformation
- loss of information

If all objects are far away (large $z^{\prime}$ ), $f / z^{\prime}$ is approximately constant => orthographic projection
$x_{p}{ }^{\prime}=s x^{\prime} \quad y_{p}{ }^{\prime}=s y^{\prime} \quad(s=$ scaling factor $)$
Orthographic projection can be viewed as projection with parallel rays + scaling

## From Camera Coordinates to Image Coordinates

Transform may be necessary because

- optical axis may not penetrate image plane at origin of desired coordinate system
- transition to discrete coordinates may require scaling
$x_{p}{ }^{\prime \prime}=\left(x_{p}{ }^{\prime}-x_{p 0}\right) a \quad a, b$ scaling parameters
$y_{p}{ }^{\prime \prime}=\left(y_{p}^{\prime}-y_{p 0^{\prime}}\right) b \quad x_{p 0^{\prime}}, y_{p 0^{\prime}}$ origin of image coordinate system

Discrete image coordinates:
$x^{\prime \prime}=0 . .511 y^{\prime \prime}=0 . .575$

Transformation parameters:
$\mathrm{x}_{\mathrm{p} 0}{ }^{\prime}=\mathrm{c} 1 \quad \mathrm{y}_{\mathrm{p} 0}{ }^{\prime}=\mathrm{d} 1 \quad \mathrm{a}=512 /(\mathrm{c} 2-\mathrm{c} 1) \quad \mathrm{b}=576 /(\mathrm{d} 2-\mathrm{d} 1)$

## Complete Perspective Projection Equation

We combine the 3 transformation steps:

1. scene coordinates $\Rightarrow$ camera coordinates
2. projection of camera coordinates into image plane
3. camera coordinates $=>$ image coordinates
$x_{p}{ }^{\prime \prime}=\left\{f / z^{\prime}\left[\cos \beta \cos \gamma\left(x-x_{0}\right)+\cos \beta \sin \gamma\left(y-y_{0}\right)+\sin \beta\left(z-z_{0}\right)\right]-x_{p 0}\right\} a$
$y_{p}{ }^{\prime \prime}=\left\{\mathrm{f} / \mathrm{z}^{\prime}\left[(-\sin \alpha \sin \beta \cos \gamma-\cos \alpha \sin \gamma)\left(x-x_{0}\right)+\right.\right.$ $(-\sin \alpha \sin \beta \sin \gamma+\cos \alpha \cos \gamma)\left(y-y_{0}\right)+$ $\left.\left.\sin \alpha \cos \beta\left(z-z_{0}\right)\right]-y_{p 0}\right\} b$
with $z^{\prime}=(-\cos \alpha \sin \beta \cos \gamma+\sin \alpha \sin \gamma)\left(x-x_{0}\right)+$ $(-\cos \alpha \sin \beta \sin \gamma-\sin \alpha \cos \gamma)\left(y-y_{0}\right)+$ $\cos \alpha \cos \beta\left(z-z_{0}\right)$

## Homogeneous Coordinates (1)

4D notation for 3D coordinates which allows to express nonlinear 3D transformations as linear 4D transformations.

Normal: $\underline{v}^{\prime}=\mathbf{R}\left(\underline{\mathbf{v}}-\underline{v}_{0}\right)$
Homogeneous coordinates: $v^{\prime}=A v \quad$ (note italics for
$A=R T=\left[\begin{array}{cccc}r_{11} & r_{12} & r_{13} & 0 \\ r_{21} & r_{22} & r_{23} & 0 \\ r_{11} & r_{32} & r_{33} & 0 \\ 0 & 0 & 0 & 1\end{array}\right]\left[\begin{array}{cccc}1 & 0 & 0 & -x_{0} \\ 0 & 1 & 0 & -y_{0} \\ 0 & 0 & 1 & -z_{0} \\ 0 & 0 & 0 & 1\end{array}\right]$

Transition to homogeneous coordinates:
$\underline{v}^{\top}=\left[\begin{array}{lll}\mathrm{x} & \mathrm{z}\end{array}\right] \Rightarrow \underline{v}^{\top}=[\mathbf{w x} w y \mathrm{wz} w] \quad \mathbf{w} \neq 0$ is arbitrary constant

Return to normal coordinates:

1. Divide components 1-3 by 4th component
2. Omit 4th component

## Homogeneous Coordinates (2)

Perspective projection in homogeneous coordinates:
$\underline{v}_{p}^{\prime}=P \underline{v}^{\prime}$ with $P=\left[\begin{array}{llll}1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 1 / f & 0\end{array}\right]$ and $\underline{v}^{\prime}=\left[\begin{array}{l}w x \\ w y \\ w z \\ w\end{array}\right]$ gives $\underline{v}_{p}{ }^{\prime}=\left[\begin{array}{l}w x \\ w y \\ w z \\ w z / f\end{array}\right]$
Returning to normal coordinates gives $\underline{v}_{p}{ }^{\prime}=\left[\begin{array}{c}x f / z \\ \mathrm{yf} / \mathrm{z} \\ \mathrm{f}\end{array}\right]$
compare with earlier slide

Transformation from camera into image coordinates:
$\underline{v}_{p}{ }^{\prime \prime}=B \underline{v}_{p}{ }^{\prime}$ with $B=\left[\begin{array}{cccc}a & 0 & 0 & -x_{0} a \\ 0 & b & 0 & -y_{0} b \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1\end{array}\right]$ and $\underline{v}_{p}{ }^{\prime}=\left[\begin{array}{l}w x_{p} \\ w y_{p} \\ 0 \\ w\end{array}\right]$ gives $\underline{v}_{p}{ }^{\prime \prime}=\left[\begin{array}{l}w a\left(x_{p^{\prime}}-x_{0}\right) \\ w b\left(y_{p}-y_{0}\right) \\ 0 \\ w\end{array}\right]$

## Homogeneous Coordinates (3)

Perspective projection can be completely described in terms of a linear transformation in homogeneous coordinates:

$$
\underline{v}_{p}{ }^{\prime \prime}=B P R T \underline{v}
$$

$B P R T$ may be combined into a single $4 \times 4$ matrix $C$ :

$$
v_{p}{ }^{\prime \prime}=C \underline{v}
$$

In the literature the parameters of these equations may vary because of different choices of coordinate systems, different order of translation and rotation, different camera models, etc.

## Inverse Perspective Equations

Which points in a scene correspond to a point in the image?


Each image point defines a projection ray as the locus of possible scene points (for simplicity in camera coordinates):
$\underline{v}_{p}{ }^{\prime}=\underline{v}_{\lambda}{ }^{\prime}=\lambda \underline{v}_{p}{ }^{\prime}$
origin

$\underline{v}=\underline{v}_{0}+R^{\top} \lambda \underline{v}_{p}{ }^{\prime}$

3 equations with the 4 unknowns $\mathbf{x}, \mathbf{y}, \mathbf{z}, \lambda$ and camera parameters $R$ and $\underline{v}_{0}$
Applications of inverse perspective mapping for e.g.

- distance measurements
- binocular stereo
- camera calibration
- motion stereo


## Binocular Stereo (1)



| $\underline{I}_{1}, \underline{I}_{2}$ | camera positions (optical center) |
| :--- | :--- |
| $\underline{b}$ | stereo base |
| $\underline{o}_{1}, \underline{o}_{2}$ | camera orientations (unit vectors) |
| $\mathbf{f}_{1}, f_{2}$ | focal distances |
| $\underline{v}$ | scene point |
| $\underline{u}_{1}, \underline{u}_{2}$ | projection rays of scene point (unit vectors) |

## Binocular Stereo (2)

Determine distance to $\underline{v}$ by measuring $\underline{u}_{1}$ and $\underline{u}_{2}$
Formally: $\quad \alpha \underline{u}_{1}=\underline{b}+\beta \underline{u}_{2} \quad \Rightarrow \quad \underline{v}=\alpha \underline{u}_{1}+\underline{l}_{1}$
$\alpha$ and $\beta$ are overconstrained by the vector equation. In practice,
measurements are inexact, no exact solution exists (rays do not intersect).
Better approach: Solve for the point of closest approximation of both rays:
$\underline{v}=\frac{\alpha_{0} \underline{u}_{1}+\left(\underline{b}+\beta_{0} \underline{u}_{2}\right)}{2}+\underline{l}_{1} \Rightarrow \quad$ minimize $\left\|\alpha \underline{u}_{1}-\left(\underline{b}+\beta \underline{u}_{2}\right)\right\|^{2}$
Solution: $\quad \alpha_{0}=\frac{\underline{u}_{1}^{\top} \underline{b}-\left(\underline{u}_{1}^{\top} \underline{u}_{2}\right)\left(\underline{u}_{2}^{\top} \underline{b}\right)}{1-\left(\underline{u}_{1}^{\top} \underline{u}_{2}\right)^{2}}$

$$
\beta_{0}=\frac{\left(\underline{u}_{1}^{\top} \underline{u}_{2}\right)\left(\underline{u}_{1}^{\top} \underline{b}\right)-\left(\underline{u}_{2}^{\top} \underline{b}\right)}{1-\left(\underline{u}_{1}^{\top} \underline{u}_{2}\right)^{2}}
$$

## Distance in Digital Images

Intuitive concepts of continuous images do not always carry over to digital images.
Several methods for measuring distance between pixels:
Eucledian distance
$\left.D_{E}(i, j),(h, k)\right)=\sqrt{(i-h)^{2}+(j-k)^{2}}$
costly computation of square root, can be avoided for distance comparisons

City-block distance
$D_{4}((i, j)(h, k))=|i-h|+|j-k|$
number of horizontal and vertical steps in a rectangular grid

Chessboard distance
$\left.D_{8}(\mathbf{i}, j)(h, k)\right)=\max \{|i-h|,|j-k|\}$
number of steps in a rectangular grid if diagonal steps are allowed (number of moves of a king on a chessboard)

## Connectivity in Digital Images

Connectivity is an important property of subsets of pixels. It is based on adjacency (or neighbourhood):

Pixels are 4-neighbours
if their distance is $D_{4}=1$

Pixels are 8-neighbours
if their distance is $D_{8}=1$

all 4-neighbours of center pixel
all 8-neighbours of center pixel

A path from pixel $P$ to pixel $Q$ is a sequence of pixels beginning at $Q$ and ending at $P$, where consecutive pixels are neighbours.

In a set of pixels, two pixels $P$ and $Q$ are connected, if there is a path between $P$ and $Q$ with pixels belonging to the set.
A region is a set of pixels where each pair of pixels is connected.

## Closed Curve Paradoxon


line 1
a similar paradoxon
arises if
4-neighbourhoods are used

line 2

solid lines if 8-neighbourhood is used
line 2 does not intersect line 1 although it crosses from the outside to the inside

## Geometric Transformations

Various applications:

- change of view point
- elimination of geometric distortions from image capturing
- registration of corresponding images
- artificial distortions, Computer Graphics applications

Step 1: Determine mapping $T(x, y)$ from old to new coordinate system
Step 2: Compute new coordinates ( $x^{\prime}, y^{\prime}$ ) for ( $x, y$ )
Step 3: Interpolate greyvalues at grid positions from greyvalues at transformed positions


## Polynomial Coordinate Transformations

General format of transformation:

$$
\begin{aligned}
& x^{\prime}=\sum_{r=0}^{m} \sum_{k=0}^{m-r} a_{r k} x^{r} y^{k} \\
& y^{\prime}=\sum_{r=0}^{m} \sum_{k=0}^{m-r} b_{r k} x^{r} y^{k}
\end{aligned}
$$

- Assume polynomial mapping between ( $\mathbf{x}, \mathrm{y}$ ) and ( $\mathbf{x}^{\prime}, y^{\prime}$ ) of degree $m$
- Determine corresponding points
- a) Solve linear equations for $a_{r k}, b_{r k}(r, k=1 \ldots m)$
b) Minimize mean square error (MSE) for point correspondences


## Approximation by biquadratic transformation:

$$
\begin{array}{ll}
x^{\prime}=a_{00}+a_{10} x+a_{01} y+a_{11} x y+a_{20} x^{2}+a_{02} y^{2} & \text { at least } 6 \text { corresponding } \\
y^{\prime}=b_{00}+b_{10} x+b_{01} y+b_{11} x y+b_{20} x^{2}+b_{02} y^{2} & \text { pairs needed }
\end{array}
$$

Approximation by affine transformation:

$$
\begin{array}{ll}
x=a_{00}+a_{10} x+a_{01} y & \text { at least } 3 \text { corresponding } \\
y^{\prime}=b_{00}+b_{10} x+b_{01} y & \text { pairs needed }
\end{array}
$$

## Translation, Rotation, Scaling, Skewing

Translation by vector t :
$\underline{v}^{\prime}=\underline{v}+\underline{t} \quad$ with $\quad \underline{v}^{\prime}=\left[\begin{array}{l}x^{\prime} \\ y^{\prime}\end{array}\right] \quad \underline{v}=\left[\begin{array}{l}x \\ y\end{array}\right] \quad \underline{t}=\left[\begin{array}{l}t_{x} \\ t_{y}\end{array}\right]$
Rotation of image coordinates by angle $\alpha$ :
$\mathbf{v}^{\prime}=\mathbf{R} \mathbf{v}$
with $\quad \mathbf{R}=\left[\begin{array}{cc}\cos \alpha & \sin \alpha \\ -\sin \alpha & \cos \alpha\end{array}\right]$

Scaling by factor a in $\mathbf{x}$-direction and factor $\mathbf{b}$ in $\mathbf{y}$-direction:
$\underline{v}^{\prime}=S \underline{v}$
with $S=\left[\begin{array}{ll}a & 0 \\ 0 & b\end{array}\right]$
Skewing by angle $\beta$ :


## Example of Geometry Correction by Scaling

Distortions of electron-tube cameras may be
1-2\% => more than 5 lines for TV images


Correction procedure may be based on

- fiducial marks engraved into optical system
- a test image with regularly spaced marks

Ideal mark positions:
$x_{m n}=a+m b, y_{m n}=c+n d$
Actual mark positions:
$\mathrm{m}=0$... $\mathrm{M}-1$
$\mathbf{x}_{\mathrm{mn}}, \mathrm{y}^{\prime} \mathrm{mn}$
$\mathrm{n}=0$... $\mathrm{N}-1$

Determine a, b, c, d such that MSE (mean

square error) of deviations is minimized

## Minimizing the MSE

Minimize $\quad E=\sum_{m=0}^{M-1} \sum_{n=0}^{N-1}\left(x_{m n}-x_{m n}^{\prime}\right)^{2}+\left(y_{m n}-y_{m n}^{\prime}\right)^{2}$

$$
=\sum_{m=0}^{M-1} \sum_{n=0}^{N-1}\left(a+m b-x_{m n}^{\prime}\right)^{2}+\left(c+n d-y_{m n}^{\prime}\right)^{2}
$$

From $d E / d a=d E / d b=d E / d c=d E / d d=0$ we get:
$a=\frac{2}{M N(M+1)} \sum_{m} \sum_{n}(2 M-1-3 m) x_{m n}^{\prime}$
$b=\frac{6}{M N\left(M^{2}-1\right)} \sum_{m} \sum_{n}(2 m-M+1) x_{m n}^{\prime}$
$c=\frac{2}{M N(N+1)} \sum_{m} \sum_{n}(2 N-1-3 n) y_{m n}^{\prime}$
$d=\frac{6}{M N\left(N^{2}-1\right)} \sum_{m} \sum_{n}(2 n-N+1) y_{m n}^{\prime}$
Special case $M=N=2:$
$a=1 / 2\left(x^{\prime}{ }_{00}+x^{\prime}{ }_{01}\right)$
$b=1 / 2\left(x_{10}^{\prime}-x_{00}^{\prime}+x_{11}^{\prime}-x^{\prime} 01\right.$
$c=1 / 2\left(y^{\prime}{ }_{00}+y^{\prime}{ }_{01}\right)$
$d=1 / 2\left(y_{01}^{\prime}-y^{\prime}{ }_{00}+y^{\prime}{ }_{11}-y^{\prime}{ }_{10}\right)$

## Principle of Greyvalue Interpolation

Greyvalue interpolation = computation of unknown greyvalues at locations ( $u^{\prime} v{ }^{\prime}$ ) from known greyvalues at locations ( $x^{\prime} y^{\prime}$ )


Two ways of viewing interpolation in the context of geometric transformations:

A Greyvalues at grid locations ( $x y$ ) in old image are placed at corresponding locations ( $x^{\prime} y^{\prime}$ ) in new image: $g\left(x^{\prime} y^{\prime}\right)=g(T(x y))$ => interpolation in new image
B Grid locations ( $u^{\prime} v^{\prime}$ ) in new image are transformed into corresponding locations (uv) in old image: $g(u v)=g\left(T^{-1}\left(u^{\prime} v^{\prime}\right)\right)$ => interpolation in old image
We will take view B:
Compute greyvalues between grid from greyvalues at grid locations.

## Nearest Neighbour Greyvalue Interpolation

Assign to (xy) greyvalue of nearest grid location
$\left(x_{i} y_{j}\right)\left(x_{i+1} y_{j}\right)\left(x_{i} y_{j+1}\right)\left(x_{i+1} y_{j+1}\right) \quad$ grid locations
( $\mathrm{x} y$ )
location between grid with $\mathrm{x}_{\mathrm{i}} \leq \mathrm{x} \leq \mathrm{x}_{\mathrm{i}+1}, \mathrm{y}_{\mathrm{j}} \leq \mathrm{y} \leq \mathrm{y}_{\mathrm{j}+1}$


Each grid location represents the greyvalues in a rectangle centered around this location:

Straight lines or edges may appear step-like after this transformation:


## Bilinear Greyvalue Interpolation

The greyvalue at location ( $x y$ ) between 4 grid points $\left(x_{i} y_{j}\right)\left(x_{i+1} y_{j}\right)$
$\left(x_{i} y_{j+1}\right)\left(x_{i+1} y_{j+1}\right)$ is computed by linear interpolation in both directions:
$g(x, y)=\frac{1}{\left(x_{i+1}-x_{i}\right)\left(y_{i+1}-y_{i}\right)}\left\{\left(x_{i+1}-x\right)\left(y_{j+1}-y\right) g\left(x_{i} y_{j}\right)+\left(x-x_{i}\right)\left(y_{j+1}-y\right) g\left(x_{i+1} y_{j}\right)+\right.$ $\left.\left(x_{i+1}-x\right)\left(y-y_{j}\right) g\left(x_{i} y_{j+1}\right)+\left(x-x_{i}\right)\left(y-y_{j}\right) g\left(x_{i+1} y_{j+1}\right)\right\}$

Simple idea behind long formula:

1. Compute $g_{12}=$ linear interpolation of $g_{1}$ and $g_{2}$
2. Compute $g_{34}=$ linear interpolation of $g_{3}$ and $g_{4}$
3. Compute $\mathbf{g}=$ linear interpolation of $\mathbf{g}_{12}$ and $g_{34}$

The step-like boundary effect is reduced.
But bilear interpolation may blur sharp edges.

## Bicubic Interpolation

Each greyvalue at a grid point is taken to represent the center value of a local bicubic interpolation surface with cross section $h_{3}$.
$h_{3}= \begin{cases}1-2|x|^{2}+|x|^{3} & \text { for } 0<|x|<1 \\ 4-8|x|+5|x|^{2}-|x|^{3} & \text { for } 1<|x|<2 \\ 0 & \text { otherwise }\end{cases}$


The greyvalue at an arbitrary point [ $u, v$ ] (black dot in figure) can be computed by

- 4 horizontal interpolations to obtain greyvalues at points $[\mathrm{u}, \mathrm{j}-1] \ldots[\mathrm{u}, \mathrm{j}+2]$ (red dots), followed by - 1 vertical interpolation (between red dots) to obtain greyvalue at [ $\mathrm{u}, \mathrm{v}$ ].

Note:
For an image with constant geyvalues $g_{0}$ the interpolated greyvalues at all points between
cross section of interpolation kernel
 the grid lines are also $\mathrm{g}_{0}$.

